



STATISTICS OF INVERSE PROBLEMS

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Table of contents

1	Statistical inverse problems	1
§01	Noisy image and known operator	1
§02	Noisy image and noisy operator	23
2	Regularisation of inverse problems	39
§03	Ill-posed inverse problems	39
§04	Regularisation by orthogonal projection	41
§05	(Generalised) linear Galerkin approach	47
§06	Spectral regularisation	54
3	Regularised estimation	61
§07	Orthogonal projection estimator	61
§08	(Generalised) Galerkin estimator	88
§09	Spectral regularisation estimator	118
4	Minimax optimal estimation	123
§10	Minimax theory: a general approach	123
§11	Deriving a lower bound	127

Chapter 1

Statistical inverse problems

The observable signal $g = T\theta$ corrupted with an additive noise is first formalised in this chapter and secondly the noisy observation of the operator.

Overview

§01	Noisy image and known operator	1
§01 01	Stochastic process	2
§01 02	Noisy image	5
§01 02 01	Examples of empirical mean models	6
§01 02 02	Extension to complex-valued models	7
§01 03	Statistical direct problem	11
§01 04	Diagonal statistical inverse problem	12
§01 04 01	Examples of diagonal inverse empirical mean models	14
§01 05	Non-diagonal statistical inverse problem	19
§01 05 01	Examples of non-diagonal inverse empirical mean models	21
§02	Noisy image and noisy operator	23
§02 01	Noisy non-diagonal operator	23
§02 01 01	Examples of empirical mean models	24
§02 02	Non-diagonal statistical inverse problem with noisy operator	26
§02 02 01	Examples of non-diagonal inverse empirical mean models with noisy operator	28
§02 03	Noisy diagonal operator	31
§02 03 01	Examples of empirical mean models	31
§02 04	Diagonal statistical inverse problem with noisy operator	33
§02 04 01	Examples of diagonal inverse empirical mean models with noisy operator	35

§01 Noisy image and known operator

Let $(\mathbb{H}, \langle \cdot, \cdot \rangle_{\mathbb{H}})$ and $(\mathbb{G}, \langle \cdot, \cdot \rangle_{\mathbb{G}})$ be separable real Hilbert spaces and let $T : \mathbb{H} \rightarrow \mathbb{G}$ be a known linear, bounded operator, $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ in short. We are interested in the reconstruction of $\theta \in \mathbb{H}$ from a noisy version of $g = T\theta$, which we formalise first in this section by introducing stochastic processes.

§01|00.01 **Notation.** For $x, y \in \mathbb{R}$ we agree on the following notations $\lfloor x \rfloor := \max \{k \in \mathbb{Z} : k \in (-\infty, x]\}$ (integer part), $x \vee y = \max(x, y)$ (maximum), $x \wedge y = \min(x, y)$ (minimum), $\{x\}_+ = \max(x, 0)$ (positive part), $\{x\}_- = \max(-x, 0)$ (negative part) and $|x| = \{x\}_+ + \{x\}_-$ (modulus).

(a) For $c \in \mathbb{R}$ and $\mathbb{A} \subseteq \overline{\mathbb{R}} := \mathbb{R} \cup \{\pm\infty\} = [-\infty, \infty]$ we set $\mathbb{A}_{\geq c} := \mathbb{A} \cap [c, \infty]$, $\mathbb{A}_{< c} := \mathbb{A} \cap (-\infty, c]$, $\mathbb{A}_{> c} := \mathbb{A} \cap (c, \infty]$, $\mathbb{A}_{< c} := \mathbb{A} \cap [-\infty, c)$, $\mathbb{A}_{\leq c} := \mathbb{A} \cap (-\infty, c]$, $\mathbb{A}_{\geq c} := \mathbb{A} \cap [c, \infty)$, $\mathbb{A}_{\neq c} := \mathbb{A} \setminus \{c\}$, and $\overline{\mathbb{A}} := \mathbb{A} \cup \{\pm\infty\}$.

- (b) For $b \in \overline{\mathbb{R}}$ and $a \in \overline{\mathbb{R}}_{<b}$, we write $\llbracket a, b \rrbracket := [a, b] \cap \overline{\mathbb{Z}}$, $\llbracket a, b \rrbracket := [a, b] \cap \overline{\mathbb{Z}}$, $\langle a, b \rangle := (a, b) \cap \overline{\mathbb{Z}}$, and $\langle a, b \rangle := (a, b) \cap \overline{\mathbb{Z}}$. Moreover, let $\llbracket n \rrbracket := \llbracket 1, n \rrbracket$ and $\langle n \rangle := \langle 1, n \rangle$ for $n \in \mathbb{N} = \mathbb{Z}_{>0}$.
- (c) For a σ -algebra \mathcal{A} we denote by $\mathcal{A}_{\mathbb{A}} := \mathcal{A} \cap \mathbb{A}$ the trace of \mathcal{A} over a set \mathbb{A} which is for $\mathbb{A} \in \mathcal{A}$ a σ -algebra too. For $c \in \mathbb{R}$ we set $\mathcal{A}_{\geq c} := \mathcal{A} \cap [c, \infty]$, $\mathcal{A}_{> c} := \mathcal{A} \cap (c, \infty]$, $\mathcal{A}_{\leq c} := \mathcal{A} \cap [-\infty, c]$, and $\mathcal{A}_{< c} := \mathcal{A} \cap [-\infty, c)$. We denote by $\overline{\mathcal{B}} := \overline{\mathcal{B}_{\mathbb{R}}}$ the Borel- σ -algebra over the compactified real line $\overline{\mathbb{R}}$, where the sets $\{-\infty\}$, $\{\infty\}$ and \mathbb{R} are in $\overline{\mathbb{R}}$ closed and open, respectively, and hence Borel-measurable. Note that $\mathcal{B} := \mathcal{B}_{\mathbb{R}}$ is the Borel- σ -algebra over \mathbb{R} .
- (d) Given two measurable space (Ω, \mathcal{A}) and $(\Omega_2, \mathcal{A}_2)$ we denote by $\mathcal{M}(\mathcal{A}, \mathcal{A}_2)$ the set of all \mathcal{A} - \mathcal{A}_2 measurable functions mapping Ω into Ω_2 . We call $f \in \mathcal{M}(\mathcal{A}) := \mathcal{M}(\mathcal{A}, \mathcal{B})$ and $f \in \overline{\mathcal{M}}(\mathcal{A}) := \overline{\mathcal{M}(\mathcal{A}, \mathcal{B})}$ *real* and *numerical*, respectively. Similarly, $f \in \mathcal{M}_{\geq 0}(\mathcal{A}) := \mathcal{M}(\mathcal{A}, \mathcal{B}_{\geq 0})$ (or $f \in \overline{\mathcal{M}}_{\geq 0}(\mathcal{A}) := \overline{\mathcal{M}(\mathcal{A}, \mathcal{B}_{\geq 0})}$) and $f \in \mathcal{M}_{> 0}(\mathcal{A}) := \mathcal{M}(\mathcal{A}, \mathcal{B}_{> 0})$ (or $f \in \overline{\mathcal{M}}_{> 0}(\mathcal{A}) := \overline{\mathcal{M}(\mathcal{A}, \mathcal{B}_{> 0})}$) is called *positiv* and *strictly positiv*. If $\mathcal{A} = \mathcal{B}$ then we write $\mathcal{M}_{\geq 0} := \mathcal{M}_{\geq 0}(\mathcal{B})$, $\mathcal{M}_{> 0} := \mathcal{M}_{> 0}(\mathcal{B})$, $\overline{\mathcal{M}}_{\geq 0} := \overline{\mathcal{M}}_{\geq 0}(\mathcal{B})$, and $\overline{\mathcal{M}}_{> 0} := \overline{\mathcal{M}}_{> 0}(\mathcal{B})$ for short. \square

§01|01 Stochastic process

§01|01.01 **Notation.** Here and subsequently, a non-empty and generally non-finite subset \mathcal{J} of \mathbb{N} , \mathbb{Z} or \mathbb{R} and a subset \mathcal{U} of \mathcal{J} denote an index set. We consider the product spaces $\mathbb{R}^{\mathcal{J}} = \prod_{j \in \mathcal{J}} \mathbb{R}$ and $\mathbb{R}^{\mathcal{U}} = \prod_{u \in \mathcal{U}} \mathbb{R}$, where we identify the family $\mathbf{y} = (y_j)_{j \in \mathcal{J}} \in \mathbb{R}^{\mathcal{J}}$ and the map $y : \mathcal{J} \rightarrow \mathbb{R}$ with $j \mapsto y_j$. Eventually, we define arithmetic operations on elements of $\mathbb{R}^{\mathcal{J}}$ coordinate-wise, for example meaning $a \cdot b = (a_j b_j)_{j \in \mathcal{J}}$ and $ra = (ra_j)_{j \in \mathcal{J}}$ for $a, b \in \mathbb{R}^{\mathcal{J}}$ and $r \in \mathbb{R}$. Let us further introduce $\mathbf{0} := (0)_{j \in \mathcal{J}}$ and $\mathbf{1} := (1)_{j \in \mathcal{J}}$. The map $\Pi_{\mathcal{U}} : \mathbb{R}^{\mathcal{J}} \rightarrow \mathbb{R}^{\mathcal{U}}$ given by $\mathbf{y} = (y_j)_{j \in \mathcal{J}} \mapsto (y_j)_{j \in \mathcal{U}} =: \Pi_{\mathcal{U}} \mathbf{y}$ is called *canonical projection*. In particular, for each $j \in \mathcal{J}$ the *coordinate map* $\Pi_j := \Pi_{\{j\}} : \mathbb{R}^{\mathcal{J}} \rightarrow \mathbb{R}$ is given by $\mathbf{y} = (y_j)_{j \in \mathcal{J}} \mapsto y_j =: \Pi_j \mathbf{y}$. Moreover, $\mathbb{R}^{\mathcal{J}}$ is equipped with the product Borel- σ -algebra $\mathcal{B}^{\otimes \mathcal{J}} := \bigotimes_{j \in \mathcal{J}} \mathcal{B}$. Recall that $\mathcal{B}^{\otimes \mathcal{J}}$ equals the smallest σ -algebra on $\mathbb{R}^{\mathcal{J}}$ such that all coordinate maps Π_j , $j \in \mathcal{J}$ are measurable. i.e., $\mathcal{B}^{\otimes \mathcal{J}} = \sigma(\Pi_j, j \in \mathcal{J})$. Moreover, let $(\mathcal{J}, \mathcal{I}, \nu)$ be a measure space with σ -algebra \mathcal{I} over \mathcal{J} containing all elementary events $\{j\}$, $j \in \mathcal{J}$, and σ -finite measure $\nu \in \mathcal{M}_\sigma(\mathcal{I})$. We denote by $\mathcal{L}_2(\nu) := \mathcal{L}_2(\mathcal{I}, \nu) := \mathcal{L}_2(\mathcal{J}, \mathcal{I}, \nu) \subseteq \overline{\mathcal{M}}(\mathcal{I})$ the usual set of square integrable numerical functions defined on $(\mathcal{J}, \mathcal{I}, \nu)$. Define the set of equivalence classes $\mathbb{J} := \mathbb{L}_2(\nu) := \mathbb{L}_2(\mathcal{J}, \mathcal{I}, \nu)$, which forms a Hilbert space endowed with usual inner product $\langle \cdot, \cdot \rangle_{\mathbb{J}} := \langle \cdot, \cdot \rangle_{\mathbb{L}_2(\nu)}$ and induced norm $\|\cdot\|_{\mathbb{J}} := \|\cdot\|_{\mathbb{L}_2(\nu)}$. \square

§01|01.02 **Comment.** Given a measurable space $(\Omega, \mathcal{A}, \mu)$, $s \in \overline{\mathbb{R}}_{\geq 0}$ and the usual space $\mathcal{L}_s(\Omega, \mathcal{A}, \mu)$ of $\mathcal{L}_s(\mu)$ -integrable functions introduce for each $h \in \overline{\mathcal{M}}(\mathcal{A})$, the μ -equivalence class $\{h\}_\mu := \{h_o \in \overline{\mathcal{M}}(\mathcal{A}) : h = h_o \text{ } \mu\text{-a.e.}\}$. Define the set of equivalence classes $\mathbb{L}_s(\mu) := \mathbb{L}_s(\mathcal{A}, \mu) := \mathbb{L}_s(\Omega, \mathcal{A}, \mu) := \{\{h\}_\mu : h \in \mathcal{L}_s(\mathcal{A}, \mu)\}$ and $\|\{h\}_\mu\|_{\mathbb{L}_s(\mu)} := \|h\|_{\mathcal{L}_s(\mu)}$ for $\{h\}_\mu \in \mathbb{L}_s(\mu)$. For $s \in \overline{\mathbb{R}}_{\geq 1}$, $(\mathbb{L}_s(\mu), \|\cdot\|_{\mathbb{L}_s(\mu)})$ is a complete normed vector space, i.e. a Banach space. Formally, we denote by $\{\bullet\}_\mu : \mathcal{L}_s(\mu) \rightarrow \mathbb{L}_s(\mu)$ the natural injection $h \mapsto \{h\}_\mu$. In case $s = 2$ the norm $\|\{h\}_\mu\|_{\mathbb{L}_2(\mu)} := \|h\|_{\mathcal{L}_2(\mu)} = (\mu(|h|^2))^{1/2}$ is induced by the inner product $(\{h\}_\mu, \{h_o\}_\mu) \mapsto \langle \{h\}_\mu, \{h_o\}_\mu \rangle_{\mathbb{L}_2(\mu)} := \mu(h h_o)$, and hence $(\mathbb{L}_2(\mu), \langle \cdot, \cdot \rangle_{\mathbb{L}_2(\mu)})$ is a Hilbert space. As usual we identify the equivalence class $\{h\}_\mu$ with its representative h , and write $h \in \mathbb{L}_2(\mu)$ for short. If $\lambda = \mu$ is the Lebesgue-measure then we write shortly $(\mathbb{L}_2, \langle \cdot, \cdot \rangle_{\mathbb{L}_2})$ and $\{\bullet\} : \mathcal{L}_2 \rightarrow \mathbb{L}_2$. \square

§01|01.03 **Stochastic process.** Let $(Y_j)_{j \in \mathcal{J}}$ be a family of real-valued random variables on a common probability space $(\Omega, \mathcal{A}, \mathbb{P})$, that is, $Y_j \in \mathcal{A}$ for each $j \in \mathcal{J}$. Consider the $\mathbb{R}^{\mathcal{J}}$ -valued random

variable $Y_\cdot := (Y_j)_{j \in \mathcal{J}} \in \mathcal{M}(\mathcal{A}, \mathcal{B}^{\otimes \mathcal{J}})$, i.e. $Y_\cdot : \Omega \rightarrow \mathbb{R}^{\mathcal{J}}$ is a \mathcal{A} - $\mathcal{B}^{\otimes \mathcal{J}}$ -measurable map given by $\omega \mapsto (Y_j(\omega))_{j \in \mathcal{J}} =: Y_\cdot(\omega)$. Y_\cdot is called a *stochastic process*. Its *distribution* $\mathbb{P}^Y := \mathbb{P} \circ Y_\cdot^{-1}$ is the image probability measure of \mathbb{P} under the map Y_\cdot , i.e. $Y_\cdot \sim \mathbb{P}^Y$ or $\mathbb{P}^Y \in \mathcal{W}(\mathcal{B}^{\otimes \mathcal{J}})$ for short. Further, denote by $\mathbb{P}^Y_u = \mathbb{P} \circ Y_u^{-1} = \mathbb{P}^Y \circ \Pi_u^{-1}$ the distribution of the stochastic process $Y_u := \Pi_u Y_\cdot = (Y_u)_{u \in \mathcal{U}}$ on $\mathcal{U} \subseteq \mathcal{J}$. The family $(\mathbb{P}^Y_u)_{u \subseteq \mathcal{J} \text{ finite}}$ is called *family of finite-dimensional distributions* of Y_\cdot or \mathbb{P}^Y . In particular, $\mathbb{P}^Y_j = \mathbb{P}^{\Pi_j Y_\cdot} = \mathbb{P}^Y \circ \Pi_j^{-1} \in \mathcal{W}(\mathcal{B})$ denotes the distribution of $Y_j = \Pi_j Y_\cdot$. Furthermore, for $j, j_0 \in \mathbb{H}$ we write $\mathbb{P}(Y_j) = \mathbb{P}^Y(\Pi_j)$ and $\text{Cov}(Y_j, Y_{j_0}) := \mathbb{P}(Y_j Y_{j_0}) - \mathbb{P}(Y_j)\mathbb{P}(Y_{j_0}) = \mathbb{P}^Y(\Pi_j \Pi_{j_0}) - \mathbb{P}^Y(\Pi_j)\mathbb{P}^Y(\Pi_{j_0})$, if it exists, for the expectation of Y_j and the covariance of Y_j and Y_{j_0} with respect to \mathbb{P} . \square

§01101.04 **Assumption.** The stochastic process $Y_\cdot = (Y_j)_{j \in \mathcal{J}}$ on a measurable space (Ω, \mathcal{A}) as a function $\Omega \times \mathcal{J} \rightarrow \mathbb{R}$ with $(\omega, j) \mapsto Y_j(\omega)$ is $\mathcal{A} \otimes \mathcal{J}$ - \mathcal{B} -measurable, $Y_\cdot \in \mathcal{M}(\mathcal{A} \otimes \mathcal{J})$ for short. \square

§01101.05 **Definition.** Let $Y_\cdot = (Y_j)_{j \in \mathcal{J}} \sim \mathbb{P}^Y$ be a stochastic process satisfying Assumption §01101.04. If $\mathbb{P}(|Y_j|) \in \mathbb{R}_{>0}$, i.e. $Y_j \in \mathcal{L}_1(\mathbb{P})$ or $\Pi_j \in \mathcal{L}_1(\mathbb{P}^Y)$ in equal, for each $j \in \mathcal{J}$, then $\mathbf{m}_\cdot := (\mathbf{m}_j := \mathbb{P}(Y_j))_{j \in \mathcal{J}} \in \mathbb{R}^{\mathcal{J}}$ is called *mean function* of Y_\cdot where $\mathbf{m}_\cdot \in \mathcal{M}(\mathcal{J})$ due to Assumption §01101.04. If in addition $\nu(\mathbf{m}_\cdot^2) \in \mathbb{R}_{>0}$, hence $\mathbf{m}_\cdot \in \mathbb{J}$, then \mathbf{m}_\cdot is called (\mathbb{J} -)mean. If $\mathbb{P}(|Y_j|^2) < \infty$, i.e., $Y_j \in \mathcal{L}_2(\mathbb{P})$ or $\Pi_j \in \mathcal{L}_2(\mathbb{P}^Y)$ in equal, for each $j \in \mathcal{J}$, then $\text{cov}_\cdot := (\text{cov}_{j,j_0} := \text{Cov}(Y_j, Y_{j_0}))_{j, j_0 \in \mathcal{J}} \in \mathbb{R}^{\mathcal{J}^2}$ is called *covariance function* of Y_\cdot , where $\text{cov}_\cdot \in \mathcal{M}(\mathcal{J}^2)$ due to Assumption §01101.04. A linear and bounded (continuous) operator from \mathbb{J} into itself, $\Gamma \in \mathbb{L}(\mathbb{J})$ for short, satisfying $\langle \Gamma x_\cdot, y_\cdot \rangle_{\mathbb{J}} = \int_{\mathcal{J}} \int_{\mathcal{J}} y_j \text{cov}_{j,j_0} x_{j_0} \nu(dj) \nu(dj_0)$ for all $y_\cdot, x_\cdot \in \mathbb{J} = \mathbb{L}_2(\nu)$ is called *covariance operator* of Y_\cdot or \mathbb{P}^Y . If Y_\cdot admits a mean function $\mathbf{m}_\cdot \in \mathcal{M}(\mathcal{J})$ (respectively mean $\mathbf{m}_\cdot \in \mathbb{J}$) and a covariance function $\text{cov}_\cdot \in \mathcal{M}(\mathcal{J}^2)$ (respectively covariance operator $\Gamma \in \mathbb{L}(\mathbb{J})$) then we write shortly $Y_\cdot \sim P_{(\mathbf{m}_\cdot, \text{cov}_\cdot)}$ (respectively $Y_\cdot \sim P_{(\mathbf{m}_\cdot, \Gamma)}$). \square

§01101.06 **Notation.** For notional convenience we eventually identify Y_j and Π_j , i.e. $Y_\cdot \sim \mathbb{P}$ for short. We denote by $\mathcal{W}(\mathcal{B})$ the set of all probability measures on $(\mathbb{R}, \mathcal{B})$, by $\mathcal{W}_2(\mathcal{B}) \subseteq \mathcal{W}(\mathcal{B})$ the subset of all probability measures with finite second moment, by $P_{(\mu, \sigma^2)} \in \mathcal{W}_2(\mathcal{B})$ a probability measure with mean $\mu \in \mathbb{R}$ and variance $\sigma^2 \in \mathbb{R}_{>0}$, and by $P_{\{0\} \times \mathbb{R}_{>0}} = \{P_{(0, \sigma^2)} \in \mathcal{W}_2(\mathcal{B}) : \sigma \in \mathbb{R}_{>0}\}$ the subset of all probability distributions with finite second moment and mean zero. For $P_{(\mu, \sigma^2)} \in \mathcal{W}_2(\mathcal{B})$, $j \in \mathbb{N}$, we denote by $\otimes_{j \in \mathbb{N}} P_{(\mu, \sigma^2)}$ the associated product measure on $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}})$. \square

§01101.07 **Remark.** A covariance operator $\Gamma \in \mathbb{L}(\mathbb{J})$ associated with a stochastic process $Y_\cdot \sim \mathbb{P}^Y$ is self-adjoint and non-negative definite, $\Gamma \in \mathbb{L}^+(\mathbb{J})$ for short. If

$$\sup \{ \mathbb{P}(|\nu(y \cdot Y_\cdot)|^2) : y_\cdot \in \mathbb{J} = \mathbb{L}_2(\nu), \|y_\cdot\|_{\mathbb{J}} \leq 1 \} \in \mathbb{R}_{>0},$$

which holds for example if $\mathbb{P}(\|Y_\cdot\|_{\mathbb{J}}^2) \in \mathbb{R}_{>0}$ or in equal $\|Y_\cdot\|_{\mathbb{J}} \in \mathcal{L}_2(\mathbb{P})$ (implying $Y_\cdot \in \mathbb{J}$ \mathbb{P} -a.s.), then there exists a covariance operator $\Gamma \in \mathbb{L}^+(\mathbb{J})$ satisfying $\langle \Gamma x_\cdot, y_\cdot \rangle_{\mathbb{J}} = \text{Cov}(\nu(x \cdot Y_\cdot), \nu(y \cdot Y_\cdot))$ for all $x_\cdot, y_\cdot \in \mathbb{J}$. Observe that $\|Y_\cdot\|_{\mathbb{J}}^2 = \sup \{ |\nu(y \cdot Y_\cdot)|^2 : y_\cdot \in \mathbb{J}, \|y_\cdot\|_{\mathbb{J}} \leq 1 \}$. Note that $\|Y_\cdot\|_{\mathbb{J}} \in \mathcal{L}_2(\mathbb{P})$ is a sufficient condition for the existence of a covariance operator, but it is not a necessary condition (see Lemma §01101.18). \square

§01101.08 **Lemma.** Let $Y_\cdot = (Y_j)_{j \in \mathcal{J}} \sim \mathbb{P}^Y$ be a stochastic process satisfying Assumption §01101.04 and $Y_j \in \mathcal{L}_2(\mathbb{P})$ for each $j \in \mathcal{J}$, and let $\mathfrak{v} \in \mathbb{R}_{>1}$.

(i) If for all $h_\cdot \in \mathbb{J}$

$$\mathbb{P}(|\nu(h \cdot Y_\cdot)|^2) \leq \mathfrak{v} \|h_\cdot\|_{\mathbb{J}}^2 \tag{01.01}$$

then Y_\cdot admits a covariance operator $\Gamma \in \mathbb{L}^+(\mathbb{J})$ satisfying $\|\Gamma\|_{\mathbb{L}(\mathbb{J})} \leq \mathfrak{v}$.

(ii) If for all $h_\bullet \in \mathbb{J}$ in addition to (01.01) we have also

$$\mathbb{P}(|\nu(h_\bullet Y_\bullet)|^2) - |\mathbb{P}(\nu(h_\bullet Y_\bullet))|^2 \geq \mathfrak{v}^{-1} \|h_\bullet\|_{\mathbb{J}}^2 \quad (01.02)$$

then $\Gamma \in \mathbb{L}(\mathbb{J})$ is invertible with inverse $\Gamma^{-1} \in \mathbb{L}(\mathbb{J})$ where $\|\Gamma^{-1}\|_{\mathbb{L}(\mathbb{J})} \leq \mathfrak{v}$.

Consequently, if (01.01) and (01.01) are satisfied for all $h_\bullet \in \mathbb{J}$ then we have

$$\mathfrak{v}^{-1} \|h_\bullet\|_{\mathbb{J}}^2 \leq \|h_\bullet\|_{\Gamma}^2 = \langle \Gamma h_\bullet, h_\bullet \rangle_{\mathbb{J}} \leq \mathfrak{v} \|h_\bullet\|_{\ell_2}^2 \quad \forall h_\bullet \in \mathbb{J}. \quad (01.03)$$

§01101.09 **Proof of Lemma §01101.08.** Given in the lecture. \square

§01101.10 **Empirical mean function.** Assume a probability space $(\mathcal{Z}, \mathcal{Z}, \mathbb{P})$ and a stochastic process $\psi_\bullet = (\psi_j)_{j \in \mathcal{J}} \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$, i.e. $\mathcal{Z} \times \mathcal{J} \ni (z, j) \mapsto \psi_j(z) \in \mathbb{R}$ is $\mathcal{Z} \otimes \mathcal{J}$ - \mathcal{B} -measurable, satisfying in addition $\psi_j \in \mathcal{L}_1(\mathbb{P}) := \mathcal{L}_1(\mathcal{Z}, \mathcal{Z}, \mathbb{P})$ for each $j \in \mathcal{J}$. Consider the product probability space $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, \mathbb{P}^{\otimes n})$ and $Y_\bullet = (Y_j)_{j \in \mathcal{J}}$ with $Y_j := \widehat{\mathbb{P}}_n(\psi_j) \in \mathcal{Z}^{\otimes n}$ where $z = (z_i)_{i \in [n]} \mapsto Y_j(z) = (\widehat{\mathbb{P}}_n(\psi_j))(z) = \frac{1}{n} \sum_{i \in [n]} \psi_j(z_i)$ for each $j \in \mathcal{J}$. By construction $\mathfrak{m}_\bullet = (\mathfrak{m}_j = \mathbb{P}(\psi_j))_{j \in \mathcal{J}} = \mathbb{P}(\psi_\bullet) \in \mathcal{M}(\mathcal{J})$ is the mean function of Y_\bullet . The statistic $\dot{\epsilon}_j := n^{1/2}(\widehat{\mathbb{P}}_n(\psi_j) - \mathbb{P}(\psi_j)) \in \mathcal{M}(\mathcal{Z}^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathcal{L}_1(\mathbb{P}^{\otimes n})$ with $\mathbb{P}^{\otimes n}(\dot{\epsilon}_j) = 0$, and we have

$$\dot{\epsilon}_\bullet = (\dot{\epsilon}_j)_{j \in \mathcal{J}} = n^{1/2}(\widehat{\mathbb{P}}_n - \mathbb{P})(\psi_\bullet) = n^{1/2}(\widehat{\mathbb{P}}_n(\psi_\bullet) - \mathbb{P}(\psi_\bullet)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J}).$$

exploiting $\psi_\bullet \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$. Since $Y_j = \mathfrak{m}_j + n^{-1/2} \dot{\epsilon}_j$ for each $j \in \mathcal{J}$ by construction we write shortly $Y_\bullet = \mathfrak{m}_\bullet + n^{-1/2} \dot{\epsilon}_\bullet$ and call Y_\bullet *empirical mean function*. If for each $j \in \mathcal{J}$ in addition we assume $\psi_j \in \mathcal{L}_2(\mathbb{P})$ then we obtain $Y_j = \widehat{\mathbb{P}}_n(\psi_j) \in \mathcal{L}_2(\mathbb{P}^{\otimes n})$ and, hence $\dot{\epsilon}_j \in \mathcal{L}_2(\mathbb{P}^{\otimes n})$ by construction. By exploiting $\psi_\bullet \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ the *covariance function* $\text{cov}_{\bullet\bullet} \in \mathcal{M}(\mathcal{J}^2)$ of $\dot{\epsilon}_\bullet = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ is given for each $j, j_\circ \in \mathcal{J}$ by

$$\text{cov}_{j,j_\circ} = \text{Cov}(\dot{\epsilon}_j, \dot{\epsilon}_{j_\circ}) = \mathbb{P}(\psi_j \psi_{j_\circ}) - \mathbb{P}(\psi_j) \mathbb{P}(\psi_{j_\circ}) = n \text{Cov}(Y_j, Y_{j_\circ}).$$

Consequently, we have $\dot{\epsilon}_\bullet \sim \mathbb{P}_{(0, \text{cov}_{\bullet\bullet})}$ and $Y_\bullet = \mathfrak{m}_\bullet + n^{-1/2} \dot{\epsilon}_\bullet \sim \mathbb{P}_{(\mathfrak{m}_\bullet, n^{-1} \text{cov}_{\bullet\bullet})}$. There exists a covariance operator $\Gamma \in \mathbb{L}(\mathbb{J})$, if in addition $\sup \{ \mathbb{P}(|\nu(a_\bullet \psi_\bullet)|^2) : a_\bullet \in \mathbb{J} = \mathbb{L}_2(\nu), \|a_\bullet\|_{\mathbb{J}} \leq 1 \} \in \mathbb{R}_{\geq 0}$, which holds whenever $\|\psi_\bullet\|_{\mathbb{J}} \in \mathcal{L}_2(\mathbb{P})$ or in equal $\mathbb{P}(\|\psi_\bullet\|_{\mathbb{J}}^2) \in \mathbb{R}_{\geq 0}$. Observe that $\|\psi_\bullet\|_{\mathbb{J}}^2 = \sup \{ |\nu(a_\bullet \psi_\bullet)|^2 : a_\bullet \in \mathbb{J}, \|a_\bullet\|_{\mathbb{J}} \leq 1 \}$. Note that $\|\psi_\bullet\|_{\mathbb{J}} \in \mathcal{L}_2(\mathbb{P})$ is a sufficient condition for the existence of a covariance operator, but it is not necessary. \square

§01101.11 **White noise process.** A stochastic process $\dot{W}_\bullet = (\dot{W}_j)_{j \in \mathcal{J}}$ is called *white noise process*, if $(\dot{W}_j)_{j \in \mathcal{J}}$ is a family of independent and identically $\mathbb{P}_{(0,1)}$ -distributed real random variables, where each \dot{W}_j has zero mean and variance one, $\dot{W}_j \sim \mathbb{P}_{(0,1)}$ and $\dot{W}_\bullet \sim \mathbb{P}_{(0,1)}^{\otimes \mathcal{J}}$ in short. \square

§01101.12 **Notation.** In other words, the distribution $\mathbb{P}^{\dot{W}_\bullet}$ of a white noise process $\dot{W}_\bullet = (\dot{W}_j)_{j \in \mathcal{J}} \sim \mathbb{P}^{\dot{W}_\bullet}$ equals the product of its marginal $\mathbb{P}_{(0,1)}$ -distributions, i.e. $\mathbb{P}^{\dot{W}_\bullet} = \otimes_{j \in \mathcal{J}} \mathbb{P}^{\dot{W}_j} = \otimes_{j \in \mathcal{J}} \mathbb{P}_{(0,1)} = \mathbb{P}_{(0,1)}^{\otimes \mathcal{J}}$. \square

§01101.13 **Remark.** The centred stochastic process $\dot{\epsilon}_\bullet := (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ of error terms considered in an Empirical mean function §01101.10 is in general not a white noise process. \square

§01101.14 **Notation.** We denote by $\ell_2 := \mathbb{L}_2(\nu_{\mathbb{N}}) = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ the space of all square-summable real-valued sequences endowed with counting measure $\nu_{\mathbb{N}} := \sum_{j \in \mathbb{N}} \delta_{\{j\}}$ over the index set \mathbb{N} .

§01101.15 **Property.** Let $\dot{W}_\bullet := (\dot{W}_j)_{j \in \mathbb{N}} \sim \mathbb{P}_{(0,1)}^{\otimes \mathbb{N}}$ be a white noise process. By assumption \dot{W}_\bullet admits $\mathbb{Q}_\bullet := (0)_{j \in \mathbb{N}}$ as ℓ_2 -mean and $\Gamma = \text{id}_{\ell_2} \in \mathbb{L}(\ell_2)$ as covariance operator, i.e. $\dot{W}_\bullet \sim \mathbb{P}_{(\mathbb{Q}_\bullet, \text{id}_{\ell_2})}$, since $\langle a_\bullet, b_\bullet \rangle_{\ell_2} = \sum_{j \in \mathbb{N}} a_j b_j = \sum_{j \in \mathbb{N}} a_j \sum_{j_\circ \in \mathbb{N}} \text{cov}_{j,j_\circ} b_{j_\circ} = \langle \Gamma a_\bullet, b_\bullet \rangle_{\ell_2}$ for all $a_\bullet, b_\bullet \in \ell_2$. \square

§01101.16 **Gaussian process.** A stochastic process $Y = (Y_j)_{j \in \mathcal{J}} \sim P_{(m, \text{cov})}$ satisfying Assumption §01101.04 with mean function $m \in \mathcal{M}(\mathcal{J})$ and covariance function $\text{cov} \in \mathcal{M}(\mathcal{J}^2)$ is called a *Gaussian process*, if the family of finite-dimensional distributions $(P^Y_u)_{u \subseteq \mathcal{J} \text{ finite}}$ consists of normal distributions, that is, $Y_u = (Y_j)_{j \in u}$ is normally distributed with mean vector $(m_j)_{j \in u}$ and covariance matrix $(\text{cov}_{j,j'})_{j,j' \in u}$. We write shortly $Y \sim N_{(m, \text{cov})}$ or $Y \sim N_{(m, \Gamma)}$, if in addition there exist a covariance operator $\Gamma \in \mathbb{L}(\mathbb{J})$ associated with Y . The Gaussian process $\dot{B} \sim N_{(0, \text{id}_{\mathbb{J}})}$ with \mathbb{J} -mean zero and covariance operator $\text{id}_{\mathbb{J}}$ is called *iso-Gaussian process* or *Gaussian white noise process*, which equals $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ in the particular case $\mathbb{J} = \mathbb{L}_2(\mathcal{U}_{\mathbb{N}}) = \ell_2$. \square

§01101.17 **Definition Random function.** Let $(\mathbb{H}, \langle \cdot, \cdot \rangle_{\mathbb{H}})$ be an Hilbert space equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{H}}$, which is induced by its topology. A random variable $Y \in \mathcal{M}(\mathcal{A}, \mathcal{B}_{\mathbb{H}})$, i.e. an \mathcal{A} - $\mathcal{B}_{\mathbb{H}}$ -measurable map $Y : (\Omega, \mathcal{A}) \rightarrow (\mathbb{H}, \mathcal{B}_{\mathbb{H}})$, is called an \mathbb{H} -valued random variable or a *random function* in \mathbb{H} . \square

§01101.18 **Lemma.** Consider $(\ell_2, \langle \cdot, \cdot \rangle_{\ell_2})$. There does not exist a non-zero random function $Y = (Y_j)_{j \in \mathbb{N}}$ in ℓ_2 which is a Gaussian white noise process.

§01101.19 **Proof of Lemma §01101.18.** Given in the lecture. \square

§01|02 Noisy image

§01102.01 **Assumption.** The Hilbert space $\mathbb{J} = \mathbb{L}_2(\mathcal{J}, \mathcal{J}, \nu)$ with σ -finite measure $\nu \in \mathcal{M}_{\sigma}(\mathcal{J})$, σ -algebra \mathcal{J} over \mathcal{J} containing all elementary events $\{j\}$, $j \in \mathcal{J}$, and the surjective partial isometry $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$, i.e. $VV^* = \text{id}_{\mathbb{J}}$, are *fixed* and presumed to be *known in advance*. \square

§01102.02 **Notation.** Come back to the reconstruction of $\theta \in \mathbb{H}$ from a noisy version of $g = T\theta \in \mathbb{G}$. Under Assumption §01102.01 setting $A := VT \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $g = (g_j)_{j \in \mathcal{J}} := Vg \in \mathbb{J}$ we write $g = A\theta$. Keep in mind, that we identify the equivalence class and its representative g . \square

§01102.03 **Noisy image.** Let $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ be a stochastic process satisfying Assumption §01101.04 with mean zero and let $n \in \mathbb{N}$ be a sample size. The stochastic process $\hat{g} = g + n^{-1/2}\dot{\epsilon}$ with \mathbb{J} -mean g is called a *noisy version* of the image $g = Vg \in \mathbb{J}$, or *noisy image* for short. We denote by \mathbb{P}_g^n the distribution of \hat{g} . If $\dot{\epsilon}$ admits (possibly depending on g) a covariance function, say $\text{cov}_{\dot{\epsilon}}^g \in \mathcal{M}(\mathcal{J}^2)$, or a covariance operator, say $\Gamma_g \in \mathbb{L}(\mathbb{J})$, then we eventually write $\dot{\epsilon} \sim P_{(0, \text{cov}_{\dot{\epsilon}}^g)}$ and $\hat{g} \sim P_{(g, n^{-1}\text{cov}_{\dot{\epsilon}}^g)}$ or $\dot{\epsilon} \sim P_{(0, \Gamma)}$ and $\hat{g} \sim P_{(g, n^{-1}\Gamma)}$ for short. \square

§01102.04 **Empirical mean model.** For each $g \in \mathbb{G}$ let $\mathbb{P}_g \in \mathcal{W}(\mathcal{Z})$ be a probability measure on a measurable space $(\mathcal{Z}, \mathcal{Z})$. Similar to an Empirical mean function §01101.10 consider a stochastic process $\psi = (\psi_j)_{j \in \mathcal{J}} \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ which in addition for all $g \in \mathbb{G}$ satisfies $\psi_j \in \mathcal{L}_1(\mathbb{P}_g)$ for each $j \in \mathcal{J}$ and $\mathbb{P}_g(\psi) = (g_j = \mathbb{P}_g(\psi_j))_{j \in \mathcal{J}} = g = Vg$. Considering a statistical product experiment $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, \mathbb{P}_g^{\otimes n} = (\mathbb{P}_g^{\otimes})_{g \in \mathbb{G}})$ as in an Empirical mean function §01101.10 we define $\hat{g} = (\hat{g}_j := \hat{\mathbb{P}}_n(\psi_j))_{j \in \mathcal{J}} = \hat{\mathbb{P}}_n(\psi) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J})$. For $g \in \mathbb{G}$ assuming a $\mathbb{P}_g^{\otimes n}$ -sample the \mathbb{J} -mean of \hat{g} is by construction $\mathbb{P}_g(\hat{g}) = g = Vg \in \mathbb{J}$. Moreover, the stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathcal{J}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_g)(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{P}_g(\psi)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J}).$$

is centred, i.e. $\dot{\epsilon}_j \in \mathcal{L}_1(\mathbb{P}_g^{\otimes n}) = \mathcal{L}_1(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, \mathbb{P}_g^{\otimes n})$ with $\mathbb{P}_g^{\otimes n}(\dot{\epsilon}_j) = 0$ for each $j \in \mathcal{J}$, and exploiting $\psi_j \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ it satisfies Assumption §01101.04. Since $\hat{g}_j = g_j + n^{-1/2}\dot{\epsilon}_j$ for each $j \in \mathcal{J}$ the stochastic process $\hat{g} = g + n^{-1/2}\dot{\epsilon}$ is a noisy version of the image $g = Vg \in \mathbb{J}$. \square

§01102.05 **Sequence model.** Consider $\mathbb{J} = \ell_2 = \mathbb{L}_2(\mathcal{U}_{\mathbb{N}})$ as in §01101.14. Let $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}}$ be a real-valued stochastic process (satisfying always Assumption §01101.04) with mean zero and let $n \in \mathbb{N}$ be

a sample size. The observable noisy version $\widehat{g} = g + n^{-1/2}\dot{\epsilon} \sim \mathbb{P}_g^n$ with ℓ_2 -mean $g \in \ell_2$ as in §01101.14 takes the form of a *sequence model*

$$\widehat{g}_j = g_j + n^{-1/2}\dot{\epsilon}_j, \quad j \in \mathbb{N}. \quad (01.04)$$

If $\dot{\epsilon}$ admits a covariance function (possibly depending on g), say $\text{cov}^g \in \mathbb{R}^{\mathbb{N}^2}$, then we eventually write $\widehat{g} \sim \mathbb{P}_{(g, n^{-1}\text{cov}^g)}$ for short. If in addition $\dot{\epsilon}$ admits a covariance operator $\Gamma_g \in \mathbb{L}(\ell_2)$ (an infinite matrix) then we write $\widehat{g} \sim \mathbb{P}_{(g, n^{-1}\Gamma_g)}$. \square

§01102.06 **Gaussian sequence model.** Let $\dot{B} := (\dot{B}_j)_{j \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ be a Gaussian white noise process. The observable noisy version $\widehat{g} = g + n^{-1/2}\dot{B}$ with ℓ_2 -mean $g \in \ell_2$ takes the form of a *Gaussian sequence model*

$$\widehat{g}_j = g_j + n^{-1/2}\dot{B}_j, \quad j \in \mathbb{N} \quad \text{with} \quad (\dot{B}_j)_{j \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}} \quad (01.05)$$

and we denote by N_g^n the distribution of the stochastic process \widehat{g} . \square

§01|02|01 Examples of empirical mean models

§01102.07 **Notation.** Consider over $\mathcal{D} \in \mathcal{B}$ the measure space $(\mathcal{D}, \mathcal{B}_{\mathcal{D}}, \lambda_{\mathcal{D}})$ where $\lambda_{\mathcal{D}} \in \mathcal{M}_{\sigma}(\mathcal{B}_{\mathcal{D}})$ denotes the restriction of the Lebesgue measure $\lambda \in \mathcal{M}_{\sigma}(\mathcal{B})$ to the Borel- σ -algebra $\mathcal{B}_{\mathcal{D}} = \mathcal{B} \cap \mathcal{D}$, and the Hilbert space $\mathbb{L}_2(\lambda_{\mathcal{D}}) := \mathbb{L}_2(\mathcal{D}, \mathcal{B}_{\mathcal{D}}, \lambda_{\mathcal{D}}) =: \mathbb{G}$. Let $(v_j)_{j \in \mathbb{N}}$ be an *orthonormal system* in $\mathbb{L}_2(\lambda_{\mathcal{D}})$. The linear operator $V : \mathbb{L}_2(\lambda_{\mathcal{D}}) \rightarrow \ell_2$ with $g \mapsto Vg := g = (g_j := \langle g, v_j \rangle_{\mathbb{L}_2(\lambda_{\mathcal{D}})})_{j \in \mathbb{N}}$ is a surjective partial isometry $V \in \mathbb{L}(\mathbb{L}_2(\lambda_{\mathcal{D}}), \ell_2)$. Its adjoint operator $V^* \in \mathbb{L}(\ell_2, \mathbb{L}_2(\lambda_{\mathcal{D}}))$ satisfies $V^*a = \sum_{j \in \mathbb{N}} a_j v_j = v_{\mathbb{N}}(a, v_{\cdot}) \in \mathbb{L}_2(\lambda_{\mathcal{D}})$ for all $a \in \ell_2$ (the limit is taken in ℓ_2). We call $g = (g_j)_{j \in \mathbb{N}}$ (*generalised*) *Fourier coefficients* and V (*generalised*) *Fourier series transform*. \square

§01102.08 **Density estimation on \mathcal{D} .** Let \mathbb{D}_2 be a set of square-integrable Lebesgue densities on $(\mathcal{D}, \mathcal{B}_{\mathcal{D}})$, and hence $\mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{\mathcal{D}}) =: \mathbb{G}$. We denote for each density $g \in \mathbb{D}_2$ by $\mathbb{P}_g := g\lambda_{\mathcal{D}} \in \mathcal{W}(\mathcal{B}_{\mathcal{D}})$ the associated probability measure. Assuming an iid. sample $(X_i)_{i \in [n]}$ of size $n \in \mathbb{N}$ we consider the statistical product experiment $(\mathcal{D}^n, \mathcal{B}_{\mathcal{D}}^{\otimes n}, \mathbb{P}_{\mathbb{D}_2}^{\otimes n} := (\mathbb{P}_g^{\otimes n})_{g \in \mathbb{D}_2})$. Let $V \in \mathbb{L}(\mathbb{L}_2(\lambda_{\mathcal{D}}), \ell_2)$ be a generalised Fourier series transform (see **Notation** §01102.07) which is fixed and known in advanced. Evidently, for each density $g \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{\mathcal{D}})$ the generalised Fourier coefficients $g = (g_j)_{j \in \mathbb{N}} = Vg$ for each $j \in \mathbb{N}$ satisfy

$$g_j = \langle g, v_j \rangle_{\mathbb{L}_2(\lambda_{\mathcal{D}})} = \lambda_{\mathcal{D}}(gv_j) = g\lambda_{\mathcal{D}}(v_j) = \mathbb{P}_g(v_j),$$

i.e. $v_j \in \mathbb{L}_1(\mathcal{D}, \mathcal{B}_{\mathcal{D}}, \mathbb{P}_g) =: \mathbb{L}_1(\mathbb{P}_g)$. Moreover, the stochastic process $v_{\cdot} = (v_j)_{j \in \mathbb{N}}$ on $(\mathcal{D}, \mathcal{B}_{\mathcal{D}}, \mathbb{P}_g)$ is $\mathcal{B}_{\mathcal{D}} \otimes 2^{\mathbb{N}}$ - \mathcal{B} -measurable, i.e. $v_{\cdot} \in \mathcal{M}(\mathcal{B}_{\mathcal{D}} \otimes 2^{\mathbb{N}})$. Similar to an Empirical mean model §01102.04 we define $\widehat{g} = (\widehat{g}_j := \widehat{\mathbb{P}}_n(v_j))_{j \in \mathbb{N}} = \widehat{\mathbb{P}}_n(v_{\cdot}) \in \mathcal{M}(\mathcal{B}_{\mathcal{D}}^{\otimes n} \otimes 2^{\mathbb{N}})$ where for each $j \in \mathbb{N}$

$$x = (x_i)_{i \in [n]} \mapsto \widehat{g}_j(x) = (\widehat{\mathbb{P}}_n(v_j))(x) = n^{-1} \sum_{i \in [n]} v_j(x_i).$$

By construction $g = (g_j = \mathbb{P}_g(v_j))_{j \in \mathbb{N}} = \mathbb{P}_g(v_{\cdot}) \in \mathcal{M}(2^{\mathbb{N}})$ is the ℓ_2 -mean of \widehat{g} . For each $j \in \mathbb{N}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\widehat{\mathbb{P}}_n(v_j) - \mathbb{P}_g(v_j)) \in \mathcal{M}(\mathcal{B}_{\mathcal{D}}^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(\mathcal{D}^n, \mathcal{B}_{\mathcal{D}}^{\otimes n}, \mathbb{P}_g^{\otimes n}) =: \mathbb{L}_1(\mathbb{P}_g^{\otimes n})$ with $\mathbb{P}_g^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $v_{\cdot} \in \mathcal{M}(\mathcal{B}_{\mathcal{D}} \otimes 2^{\mathbb{N}})$ the stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}} = n^{1/2}(\widehat{\mathbb{P}}_n - \mathbb{P}_g)(v_{\cdot}) = n^{1/2}(\widehat{\mathbb{P}}_n(v_{\cdot}) - \mathbb{P}_g(v_{\cdot})) \in \mathcal{M}(\mathcal{B}_{\mathcal{D}}^{\otimes n} \otimes 2^{\mathbb{N}})$$

satisfies Assumption §01101.04. Since $\widehat{g}_j = g_j + n^{-1/2}\dot{\epsilon}_j$ for each $j \in \mathbb{N}$ by construction $\widehat{g} = g + n^{-1/2}\dot{\epsilon}$ is a noisy version of g . \square

§01102.09 **Regression with uniform design.** Consider the measure space $([0, 1], \mathcal{B}_{[0,1]}, \lambda_{[0,1]})$ where $\lambda_{[0,1]}$ denotes the restriction of the Lebesgue measure to the Borel- σ -algebra $\mathcal{B}_{[0,1]}$ over $[0, 1]$, and the Hilbert space $\mathbb{L}_2(\lambda_{[0,1]}) := \mathbb{L}_2([0, 1], \mathcal{B}_{[0,1]}, \lambda_{[0,1]})$ of square Lebesgue-integrable functions. Let (X, Y) be a $[0, 1] \times \mathbb{R}$ -valued random vector. We denote by $\mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{[0,1]})$ the marginal distribution of X , by $\mathbb{P}^{Y|X}$ a regular conditional distribution of Y given X , and by $\mathbb{P}^{X,Y} = \mathbb{P}^X \odot \mathbb{P}^{Y|X} \in \mathcal{W}(\mathcal{B}_{[0,1]} \otimes \mathcal{B})$ the joint distribution of (X, Y) . We tactically identify X and Y with the coordinate map $\Pi_{[0,1]}$ and $\Pi_{\mathbb{R}}$, respectively, and thus (X, Y) with the identity $\text{id}_{[0,1] \times \mathbb{R}}$ such that $\mathbb{P} = \mathbb{P}^{X,Y} \in \mathcal{W}(\mathcal{B}_{[0,1]} \otimes \mathcal{B})$. If in addition $Y \in \mathbb{L}_1(\mathbb{P}) = \mathbb{L}_1([0, 1] \times \mathbb{R}, \mathcal{B}_{[0,1]} \otimes \mathcal{B}, \mathbb{P})$ then $\mathbb{P}^{Y|X}(\text{id}_{\mathbb{R}}) = \mathbb{P}(Y|X) =: g \in \mathcal{M}(\mathcal{B}_{[0,1]})$ is unique up to \mathbb{P}^X -a.s. equality. Moreover, we have $g \in \mathbb{L}_1(\mathbb{P}^X) = \mathbb{L}_1([0, 1], \mathcal{B}_{[0,1]}, \mathbb{P}^X)$ and the error term $\xi := Y - g(X)$ satisfies $\xi \in \mathbb{L}_1(\mathbb{P})$ with $\mathbb{P}(\xi) = 0$. Let us denote in this situation by $\mathbb{P}_g^{Y|X}$ and $\mathbb{P}_g := \mathbb{P}^X \odot \mathbb{P}_g^{Y|X} \in \mathcal{W}(\mathcal{B}_{[0,1]} \otimes \mathcal{B})$, respectively, a regular conditional distribution of Y given X and the joint distribution of (X, Y) . Keep however in mind, that even if $g \in \mathbb{L}_1(\mathbb{P}^X)$ is fixed the conditional distribution $\mathbb{P}_g^{Y|X}$ is still not fully specified. We assume in what follows that the regressor X is uniformly distributed on the interval $[0, 1]$, i.e. $X \sim U_{[0,1]} = \lambda_{[0,1]} = \mathbb{P}^X$ and that $g \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\mathbb{P}^X) = \mathbb{L}_2(\lambda_{[0,1]}) =: \mathbb{G}$ identifying again equivalence classes and their representatives. Denote by $U_g := U_{[0,1]} \odot \mathbb{P}_g^{Y|X}$ the joint distribution of (X, Y) without fully specifying the conditional distribution $\mathbb{P}_g^{Y|X}$. For $g, h \in \mathbb{L}_2(\mathbb{P}^X) \subseteq \mathbb{L}_1(\mathbb{P}^X)$ we have $gh \in \mathbb{L}_1(\mathbb{P}^X)$ and thus $\mathbb{P}^X(gh) \in \mathbb{R}$. Keep in mind that X and Y equals the coordinate map $\Pi_{[0,1]}$ and $\Pi_{\mathbb{R}}$, respectively. Consequently, if $Y \in \mathbb{L}_2(U_g)$ and $h \in \mathbb{L}_2(\mathbb{P}^X) = \mathbb{L}_2(\lambda_{[0,1]})$, hence $h(X) \in \mathbb{L}_2(U_g)$, then we obtain $Yh(X) \in \mathbb{L}_1(U_g)$ and

$$U_g(Yh(X)) = \mathbb{P}^X(\mathbb{P}_g^{Y|X}(Y)h) = \mathbb{P}^X(gh) = \lambda_{[0,1]}(gh) = \langle g, h \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} \in \mathbb{R}$$

identifying again equivalence classes and their representatives. We consider the statistical product experiment $(([0, 1] \times \mathbb{R})^n, (\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n}, U_g^{\otimes n} := (U_g^{\otimes n})_{g \in \mathbb{F}_2})$ of size $n \in \mathbb{N}$ and for $g \in \mathbb{F}_2$ we denote by $((X_i, Y_i))_{i \in [n]} \sim U_g^{\otimes n}$ an iid. sample of $(X, Y) \sim U_g = U_{[0,1]} \odot \mathbb{P}_g^{Y|X}$. Let $V \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2)$ be a generalised Fourier series transform as in **Notation** §01102.07 which is fixed and known in advanced. Evidently, for each $g \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]})$ the generalised Fourier coefficients $g = (g_j)_{j \in \mathbb{N}} = Vg$ for each $j \in \mathbb{N}$ satisfy

$$g_j = \langle g, v_j \rangle_{\mathbb{G}} = \lambda_{[0,1]}(g v_j) = U_g(Y v_j(X)).$$

Therefore the stochastic process $\psi_g = (\psi_j(X, Y) := Y v_j(X))_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}})$ fulfils Assumption §01101.04 and $g = U_g(\psi_g)$. Similar to an Empirical mean model §01102.04 we define $\hat{g} = (\hat{g}_j := \hat{\mathbb{P}}_n(\psi_j))_{j \in \mathbb{N}} = \hat{\mathbb{P}}_n(\psi_g) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}})$. By construction $g = U_g(\psi_g) \in \mathcal{M}(2^{\mathbb{N}})$ is the ℓ_2 -mean of \hat{g} . For each $j \in \mathbb{N}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\hat{\mathbb{P}}_n(\psi_j) - U_g(\psi_j)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(U_g^{\otimes n})$ with $U_g^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $\psi_g \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}})$ the stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}} = n^{1/2}(\hat{\mathbb{P}}_n - U_g)(\psi_g) = n^{1/2}(\hat{\mathbb{P}}_n(\psi_g) - U_g(\psi_g)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}})$$

satisfies Assumption §01101.04. Since $\hat{g}_j = g_j + n^{-1/2}\dot{\epsilon}_j$ for each $j \in \mathbb{N}$ by construction $\hat{g} = g + n^{-1/2}\dot{\epsilon}$ is a noisy version of g . \square

§01|02|02 Extension to complex-valued models

§01102.10 **Notation Reminder.** Given a non-empty and generally non-finite subset \mathcal{J} of \mathbb{N} , \mathbb{Z} or \mathbb{R} and a subset \mathcal{U} of \mathcal{J} as an index set consider the complex product spaces $\mathbb{C}^{\mathcal{J}} = \prod_{j \in \mathcal{J}} \mathbb{C}$ and $\mathbb{C}^{\mathcal{U}} = \prod_{u \in \mathcal{U}} \mathbb{C}$, where we identify the family $y = (y_j)_{j \in \mathcal{J}} \in \mathbb{C}^{\mathcal{J}}$ and the map $y : \mathcal{J} \rightarrow \mathbb{C}$ with $j \mapsto y_j$. Eventually, we define arithmetic operations on elements of $\mathbb{C}^{\mathcal{J}}$ coordinate-wise, for example meaning $a.b = (a_j b_j)_{j \in \mathcal{J}}$ and $ra = (ra_j)_{j \in \mathcal{J}}$ for $a, b \in \mathbb{C}^{\mathcal{J}}$ and $r \in \mathbb{C}$. Let us further

introduce $\mathbf{0} := (0)_{j \in \mathcal{J}}$, $\mathbf{1} := (1)_{j \in \mathcal{J}}$, and the imaginary unit i . The map $\Pi_u : \mathbb{C}^{\mathcal{J}} \rightarrow \mathbb{C}^{\mathcal{U}}$ given by $y_u = (y_j)_{j \in \mathcal{J}} \mapsto (y_j)_{j \in \mathcal{U}} =: \Pi_u y_u$ is called *canonical projection*. In particular, for each $j \in \mathcal{J}$ the *coordinate map* $\Pi_j := \Pi_{\{j\}} : \mathbb{C}^{\mathcal{J}} \rightarrow \mathbb{C}$ is given by $y_u = (y_{j'})_{j' \in \mathcal{J}} \mapsto y_j =: \Pi_j y_u$. Let \mathcal{B} denote the Borel- σ -algebra over \mathbb{C} (with a slight abuse of notation). Moreover, $\mathbb{C}^{\mathcal{J}}$ is equipped with the product Borel- σ -algebra $\mathcal{B}^{\otimes \mathcal{J}} := \bigotimes_{j \in \mathcal{J}} \mathcal{B}$. Recall that $\mathcal{B}^{\otimes \mathcal{J}}$ equals the smallest σ -algebra on $\mathbb{C}^{\mathcal{J}}$ such that all coordinate maps Π_j , $j \in \mathcal{J}$ are measurable. i.e., $\mathcal{B}^{\otimes \mathcal{J}} = \sigma(\Pi_j, j \in \mathcal{J})$. Moreover, let $(\mathcal{J}, \mathcal{I}, \nu)$ be a measure space with σ -finite measure $\nu \in \mathcal{M}_\sigma(\mathcal{J})$. We write for each \mathcal{I} - \mathcal{B} -measurable $h : \mathcal{J} \rightarrow \mathbb{C}$ shortly $h \in \mathcal{M}(\mathcal{I})$ with a slight abuse of notation. For $s \in \overline{\mathbb{R}}_{\geq 1} = [1, \infty]$ we introduce the usual space $\mathcal{L}_s(\nu) := \mathcal{L}_s(\mathcal{J}, \mathcal{I}, \nu)$ of $\mathcal{L}_s(\nu)$ -integrable complex-valued functions. Define further the set of equivalence classes $\mathbb{L}_s(\nu) := \mathbb{L}_s(\mathcal{J}, \mathcal{I}, \nu) := \{\{h\}_\nu : h \in \mathcal{L}_s(\nu)\}$ (see **Comment** §01101.02). In case $s = 2$ the norm $\|\{h\}_\mu\|_{\mathbb{L}_2(\nu)} := \|h\|_{\mathcal{L}_2(\nu)} = (\nu(|h|^2))^{1/2}$ is induced by the inner product $(\{h\}_\nu, \{h_o\}_\nu) \mapsto \langle \{h\}_\nu, \{h_o\}_\nu \rangle_{\mathbb{L}_2(\nu)} := \nu(h \overline{h_o})$ (denoting by \bar{z} the complex conjugate of $z \in \mathbb{C}$), and hence $(\mathbb{L}_2(\nu), \langle \cdot, \cdot \rangle_{\mathbb{L}_2(\nu)})$ is a complex Hilbert space. As usual we identify the equivalence class $\{h\}_\nu$ with its representative h , and write $h \in \mathbb{L}_2(\nu)$ for short. If $\lambda = \nu$ is the Lebesgue-measure then we write also shortly $(\mathbb{L}_s, \|\cdot\|_{\mathbb{L}_s})$ and $(\mathbb{L}_2, \langle \cdot, \cdot \rangle_{\mathbb{L}_2})$. Let $(Y_j)_{j \in \mathcal{J}}$ be a family of complex-valued random variables on a common probability space $(\Omega, \mathcal{A}, \mathbb{P})$, that is, $Y_j \in \mathcal{M}(\mathcal{A})$ for each $j \in \mathcal{J}$. Consider the $\mathbb{C}^{\mathcal{J}}$ -valued random variable $Y_\cdot := (Y_j)_{j \in \mathcal{J}} \in \mathcal{M}(\mathcal{A}, \mathcal{B}^{\otimes \mathcal{J}})$ where $Y_\cdot : \Omega \rightarrow \mathbb{C}^{\mathcal{J}}$ is a \mathcal{A} - $\mathcal{B}^{\otimes \mathcal{J}}$ -measurable map given by $\omega \mapsto (Y_j(\omega))_{j \in \mathcal{J}} =: Y_\cdot(\omega)$. Y_\cdot is called a (complex-valued) *stochastic process*. Its *distribution* $\mathbb{P}^{Y_\cdot} := \mathbb{P} \circ Y_\cdot^{-1}$ is the image probability measure of \mathbb{P} under the map Y_\cdot , i.e. $Y_\cdot \sim \mathbb{P}^{Y_\cdot}$ or $\mathbb{P}^{Y_\cdot} \in \mathcal{W}(\mathcal{B}^{\otimes \mathcal{J}})$ for short. Further, denote by $\mathbb{P}^{Y_u} = \mathbb{P} \circ Y_u^{-1} = \mathbb{P}^{Y_\cdot} \circ \Pi_u^{-1}$ the distribution of the stochastic process $Y_u := \Pi_u Y_\cdot = (Y_u)_{u \in \mathcal{U}}$ on $\mathcal{U} \subseteq \mathcal{J}$. The family $(\mathbb{P}^{Y_u})_{\mathcal{U} \subseteq \mathcal{J} \text{ finite}}$ is called *family of finite-dimensional distributions* of Y_\cdot or \mathbb{P}^{Y_\cdot} . In particular, $\mathbb{P}^{Y_j} = \mathbb{P}^{\Pi_j Y_\cdot} = \mathbb{P}^{Y_\cdot} \circ \Pi_j^{-1} \in \mathcal{W}(\mathcal{B})$ denotes the distribution of $Y_j = \Pi_j Y_\cdot$. Furthermore, for $j, j_o \in \mathcal{J}$ we write $\mathbb{P}(Y_j) = \mathbb{P}^{Y_j}(\Pi_j)$ and $\text{Cov}(Y_j, Y_{j_o}) := \mathbb{P}(Y_j \overline{Y_{j_o}}) - \mathbb{P}(Y_j) \mathbb{P}(\overline{Y_{j_o}})$, if it exists, for the expectation of Y_j and the covariance of Y_j and Y_{j_o} with respect to \mathbb{P} . \square

§01102.11 **Assumption.** The complex-valued stochastic process $Y_\cdot = (Y_j)_{j \in \mathcal{J}}$ on a common measurable space (Ω, \mathcal{A}) as a function $\Omega \times \mathcal{J} \rightarrow \mathbb{C}$ with $(\omega, j) \mapsto Y_j(\omega)$ is $\mathcal{A} \otimes \mathcal{I}$ - \mathcal{B} -measurable, $Y_\cdot \in \mathcal{M}(\mathcal{A} \otimes \mathcal{I})$ for short. \square

§01102.12 **Notation.** Consider the *complex* Hilbert spaces $\mathbb{L}_2(\lambda_{[0,1]}) := \mathbb{L}_2([0, 1], \mathcal{B}_{[0,1]}, \lambda_{[0,1]})$ and $\mathbb{J} := \ell_2(\mathbb{Z}) = \mathbb{L}_2(\nu_{\mathbb{Z}}) = \mathbb{L}_2(\mathbb{Z}, 2^{\mathbb{Z}}, \nu_{\mathbb{Z}})$ where the latter is the space of all square-summable *complex-valued* sequences endowed with counting measure $\nu_{\mathbb{Z}} := \sum_{j \in \mathbb{Z}} \delta_{\{j\}}$ over the index set \mathbb{Z} . For each $j \in \mathbb{Z}$ introduce the exponential $e_j \in \mathcal{M}(\mathcal{B}_{[0,1]})$ with $e_j(x) := \exp(-i2\pi x j)$ for $x \in [0, 1)$ forming together the exponential basis $(e_j)_{j \in \mathbb{Z}}$ in $\mathbb{L}_2(\lambda_{[0,1]})$. Moreover, the complex-valued stochastic process $e_\cdot = (e_j)_{j \in \mathbb{Z}}$ on $([0, 1), \mathcal{B}_{[0,1]})$ is $\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}}$ - \mathcal{B} -measurable, i.e. satisfies Assumption §01102.11. The linear operator $F : \mathbb{L}_2(\lambda_{[0,1]}) \rightarrow \ell_2(\mathbb{Z})$ with $g \mapsto Fg := g_\cdot = (g_j := \langle g, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})})_{j \in \mathbb{Z}} = \lambda_{[0,1]}(g \overline{e_\cdot})$ is a bijective isometry (unitary) $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2(\mathbb{Z}))$. Its adjoint operator $F^* \in \mathbb{L}(\ell_2(\mathbb{Z}), \mathbb{L}_2(\lambda_{[0,1]}))$ satisfies

$$\nu_{\mathbb{Z}}(\lambda_{[0,1]}(g \overline{e_\cdot}) \overline{a_\cdot}) = \nu_{\mathbb{Z}}((Fg) \overline{a_\cdot}) = \langle Fg, a_\cdot \rangle_{\ell_2} = \langle g, F^* a_\cdot \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \lambda_{[0,1]}(g \overline{F^* a_\cdot}) = \lambda_{[0,1]}(g \nu_{\mathbb{Z}}(\overline{a_\cdot} e_\cdot))$$

and hence $F^* a_\cdot = \sum_{j \in \mathbb{Z}} a_j e_j = \nu_{\mathbb{Z}}(a_\cdot e_\cdot) \in \mathbb{L}_2(\lambda_{[0,1]})$ for all $a_\cdot \in \ell_2(\mathbb{Z})$ (the limit is taken in $\mathbb{L}_2(\lambda_{[0,1]})$). We call $g_\cdot = (g_j)_{j \in \mathbb{Z}}$ *Fourier coefficients* and F *Fourier-series transform*. \square

§01102.13 **Density estimation on** $[0, 1)$. Let \mathbb{D}_2 be a set of square-integrable Lebesgue densities on $([0, 1), \mathcal{B}_{[0,1]})$, and hence $\mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]})$ (by the usual embedding of real-valued functions) as in **Notation** §01102.10. We denote for each Lebesgue density g on $([0, 1), \mathcal{B}_{[0,1]})$ by $\mathbb{P}_g := g \lambda_{[0,1]} \in \mathcal{W}(\mathcal{B}_{[0,1]})$ the associated probability measure. We consider the statistical product experiment $([0, 1)^n, \mathcal{B}_{[0,1]}^{\otimes n}, \mathbb{P}_{\mathbb{D}_2}^{\otimes n} :=$

$(\mathbb{P}_g^{\otimes n})_{g \in \mathbb{D}_2}$). Let $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2(\mathbb{Z}))$ be the Fourier-series transform (see [Notation §01102.12](#)). Evidently, for each $g \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]}) \subseteq \mathbb{L}_1(\lambda_{[0,1]})$ its Fourier-series $g = (g_j)_{j \in \mathbb{Z}} = Fg$ for each $j \in \mathbb{Z}$ satisfy

$$g_j = \langle g, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \lambda_{[0,1]}(g \bar{e}_j) = \mathbb{P}_g(\bar{e}_j).$$

The complex-valued stochastic process $\bar{e} = (\bar{e}_j)_{j \in \mathbb{Z}}$ on $([0, 1], \mathcal{B}_{[0,1]})$ is $(\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}})$ - \mathcal{B} -measurable, i.e. $\bar{e} \in \mathcal{M}(\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}})$ for short. We define $\hat{g} = (\hat{g}_j := \hat{\mathbb{P}}_n(\bar{e}_j))_{j \in \mathbb{Z}} = \hat{\mathbb{P}}_n(\bar{e}) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes n} \otimes 2^{\mathbb{Z}})$ similar to an Empirical mean model [§01102.04](#) where for each $j \in \mathbb{Z}$

$$x = (x_i)_{i \in [n]} \mapsto \hat{g}_j(x) = (\hat{\mathbb{P}}_n(\bar{e}_j))(x) = n^{-1} \sum_{i \in [n]} \bar{e}_j(x_i) = n^{-1} \sum_{i \in [n]} \exp(i2\pi j x_i).$$

By construction $g = (g_j = \mathbb{P}_g(\bar{e}_j))_{j \in \mathbb{Z}} = \mathbb{P}_g(\bar{e}) \in \mathcal{M}(2^{\mathbb{Z}})$ is the $\ell_2(\mathbb{Z})$ -mean of \hat{g} . For each $j \in \mathbb{Z}$ the statistic $\dot{e}_j := n^{1/2}(\hat{\mathbb{P}}_n(\bar{e}_j) - \mathbb{P}_g(\bar{e}_j)) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes n})$ is centred, i.e. $\dot{e}_j \in \mathbb{L}_1(\mathbb{P}_g^{\otimes n})$ with $\mathbb{P}_g^{\otimes n}(\dot{e}_j) = 0$, and exploiting $\bar{e} = (\bar{e}_j)_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}})$ the complex valued stochastic process

$$\dot{e} = (\dot{e}_j)_{j \in \mathbb{Z}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_g)(\bar{e}) = n^{1/2}(\hat{\mathbb{P}}_n(\bar{e}) - \mathbb{P}_g(\bar{e})) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes n} \otimes 2^{\mathbb{Z}})$$

satisfies [Assumption §01102.11](#). Since $\hat{g}_j = g_j + n^{-1/2}\dot{e}_j$ for each $j \in \mathbb{Z}$ by construction $\hat{g} = g + n^{-1/2}\dot{e}$ is a noisy version of g . \square

[§01102.14](#) **Regression with uniform design.** Consider the measure space $([0, 1], \mathcal{B}_{[0,1]}, \lambda_{[0,1]})$ and the complex Hilbert space $\mathbb{L}_2(\lambda_{[0,1]})$ as in [Notation §01102.10](#). Let \mathbb{F}_2 be a set of square-integrable real-valued regression function on $([0, 1], \mathcal{B}_{[0,1]})$, and hence $\mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]}) =: \mathbb{G}$ (by the usual embedding of real-valued functions). We consider as in [Regression with uniform design §01102.09](#) the statistical product experiment $(([0, 1] \times \mathbb{R})^n, (\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n}, \mathbb{U}_g^{\otimes n} := (\mathbb{U}_g^{\otimes n})_{g \in \mathbb{F}_2})$ of size $n \in \mathbb{N}$ and for $g \in \mathbb{F}_2$ we denote by $((X_i, Y_i)_{i \in [n]} \sim \mathbb{U}_g^{\otimes n}$ an iid. sample of $(X, Y) \sim \mathbb{U}_g = \mathbb{U}_{[0,1]} \odot \mathbb{P}_g^{Y|X}$. Let $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2(\mathbb{Z}))$ be the Fourier-series transform (see [Notation §01102.12](#)). For each $g \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]})$ the Fourier coefficients $g = (g_j)_{j \in \mathbb{Z}} = Fg$ for each $j \in \mathbb{Z}$ satisfy

$$g_j = \langle g, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \lambda_{[0,1]}(g \bar{e}_j) = \mathbb{U}_g(Y \bar{e}_j(X)).$$

The complex-valued stochastic process $\psi = (\psi_j(X, Y) := Y \bar{e}_j(X))_{j \in \mathbb{Z}} \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{Z}})$ fulfils [Assumption §01102.11](#) and $g = \mathbb{U}_g(\psi)$. Similar to an Empirical mean model [§01102.04](#) we define $\hat{g} = (\hat{g}_j := \hat{\mathbb{P}}_n(\psi_j))_{j \in \mathbb{Z}} = \hat{\mathbb{P}}_n(\psi) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{Z}})$ where for each $j \in \mathbb{Z}$

$$\hat{g}_j = \hat{\mathbb{P}}_n(\psi_j) = n^{-1} \sum_{i \in [n]} Y_i \bar{e}_j(X_i) = n^{-1} \sum_{i \in [n]} Y_i \exp(i2\pi j X_i).$$

By construction $g = \mathbb{U}_g(\psi) \in \mathcal{M}(2^{\mathbb{Z}})$ is the $\ell_2(\mathbb{Z})$ -mean of \hat{g} . For each $j \in \mathbb{Z}$ the statistic $\dot{e}_j := n^{1/2}(\hat{\mathbb{P}}_n(\psi_j) - \mathbb{U}_g(\psi_j)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n})$ is centred, i.e. $\dot{e}_j \in \mathbb{L}_1(\mathbb{U}_g^{\otimes n})$ with $\mathbb{U}_g^{\otimes n}(\dot{e}_j) = 0$, and exploiting $\psi \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{Z}})$ the complex-valued stochastic process

$$\dot{e} = (\dot{e}_j)_{j \in \mathbb{Z}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{U}_g)(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{U}_g(\psi)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{Z}})$$

satisfies [Assumption §01102.11](#). Since $\hat{g}_j = g_j + n^{-1/2}\dot{e}_j$ for each $j \in \mathbb{Z}$ by construction $\hat{g} = g + n^{-1/2}\dot{e}$ is a noisy version of g . \square

[§01102.15](#) **Notation.** Consider the complex Hilbert space $\mathbb{L}_2 := \mathbb{L}_2(\lambda) = \mathbb{L}_2(\mathbb{R}, \mathcal{B}, \lambda)$ as in [Notation §01102.10](#). Let $F \in$ denote the *Fourier-Plancherel transform* satisfying

$$g_j := (Fg)_j = \int_{\mathbb{R}} g(x) \exp(i2\pi x j) \lambda(dx), \quad j \in \mathbb{R}, \quad \forall g \in \mathbb{L}_1 \cap \mathbb{L}_2.$$

Introducing $\mathbf{e}_\bullet = (e_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}^2)$ given by $e_j(x) := \exp(-i2\pi x j)$ for $x, j \in \mathbb{R}$ we evidently have $\mathbf{e}_j, \mathbf{e}_\bullet(x) \in \mathcal{M}(\mathcal{B})$ and (keep for each $j \in \mathbb{R}$ in mind that $\bar{\mathbf{e}}_j \in \mathbb{L}_\infty$ but $\bar{\mathbf{e}}_j \notin \mathbb{L}_2$)

$$g_\bullet = (g_j)_{j \in \mathbb{R}} = Fg = (Fg)_\bullet = ((Fg)_j = \lambda(g\bar{\mathbf{e}}_j))_{j \in \mathbb{R}} = \lambda(g\bar{\mathbf{e}}_\bullet), \quad \forall g \in \mathbb{L}_1 \cap \mathbb{L}_2.$$

Moreover, F is unitary with adjoint $F^* \in \mathbb{L}(\mathbb{L}_2)$ satisfying

$$\lambda(\lambda(g\bar{\mathbf{e}}_\bullet)\bar{h}_\bullet) = \lambda((Fg)_\bullet\bar{h}_\bullet) = \langle Fg, h_\bullet \rangle_{\mathbb{L}_2} = \langle g, F^*h_\bullet \rangle_{\mathbb{L}_2} = \lambda(g\overline{F^*h_\bullet}) = \lambda(g\lambda(\bar{h}_\bullet\bar{\mathbf{e}}_\bullet))$$

and hence $(F^*h_\bullet)(x) = \lambda(h_\bullet\mathbf{e}_\bullet(x))$, $x \in \mathbb{R}$, for all $h_\bullet \in \mathbb{L}_1 \cap \mathbb{L}_2$. For $g \in \mathbb{L}_1 \cap \mathbb{L}_2$ we write $g_\bullet := (g_j := \lambda(g\bar{\mathbf{e}}_j))_{j \in \mathbb{R}} = \lambda(g\bar{\mathbf{e}}_\bullet) = Fg$ such that $g = F^*g_\bullet$ (with a slight abuse of notation). We note that the complex-valued stochastic process $\mathbf{e}_\bullet = (e_j)_{j \in \mathbb{R}}$ on $(\mathbb{R}, \mathcal{B})$ is \mathcal{B}^2 - \mathcal{B} -measurable, i.e. $\mathbf{e}_\bullet = (e_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}^2)$ for short, and thus satisfies Assumption §01102.11. \square

§01102.16 **Density estimation on \mathbb{R} .** Let \mathbb{D}_2 be a set of square-integrable Lebesgue densities on $(\mathbb{R}, \mathcal{B})$, and hence $\mathbb{D}_2 \subseteq \mathbb{L}_2 =: \mathbb{G}$ (by the usual embedding of real-valued functions). We denote for each density $g \in \mathbb{D}_2$ by $\mathbb{P}_g := g\lambda \in \mathcal{W}(\mathcal{B})$ the associated probability measure. We consider the statistical product experiment $(\mathbb{R}^n, \mathcal{B}^{\otimes n}, \mathbb{E}_g^{\otimes n} := (\mathbb{P}_g^{\otimes n})_{g \in \mathbb{D}_2})$. Let $F \in \mathbb{L}(\mathbb{L}_2)$ be the Fourier-Plancherel transform (see **Notation** §01102.15). Evidently, for each $g \in \mathbb{D}_2 \subseteq \mathbb{L}_2$ and hence $g \in \mathbb{L}_1 \cap \mathbb{L}_2$ its Fourier-Plancherel transform $g_\bullet = (g_j)_{j \in \mathbb{R}} = Fg$ for each $j \in \mathbb{R}$ satisfies

$$g_j = \lambda(g\bar{\mathbf{e}}_j) = \mathbb{P}_g(\bar{\mathbf{e}}_j).$$

The stochastic process $(\bar{\mathbf{e}}_j)_{j \in \mathbb{R}}$ on $(\mathbb{R}, \mathcal{B})$ is \mathcal{B}^2 - \mathcal{B} -measurable, i.e. $\bar{\mathbf{e}}_\bullet = (\bar{\mathbf{e}}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}^2)$ for short. Similar to an Empirical mean model §01102.04 we define $\hat{g}_\bullet = (\hat{g}_j := \hat{\mathbb{P}}_n(\bar{\mathbf{e}}_j))_{j \in \mathbb{R}} = \hat{\mathbb{P}}_n(\bar{\mathbf{e}}_\bullet) \in \mathcal{M}(\mathcal{B}^{\otimes n} \otimes \mathcal{B})$ where for each $j \in \mathbb{R}$

$$x = (x_i)_{i \in \llbracket n \rrbracket} \mapsto \hat{g}_j(x) = (\hat{\mathbb{P}}_n(\bar{\mathbf{e}}_j))(x) = n^{-1} \sum_{i \in \llbracket n \rrbracket} \bar{\mathbf{e}}_j(x_i) = n^{-1} \sum_{i \in \llbracket n \rrbracket} \exp(i2\pi j x_i).$$

By construction $g_\bullet = (g_j = \mathbb{P}_g(\bar{\mathbf{e}}_j))_{j \in \mathbb{R}} = \mathbb{P}_g(\bar{\mathbf{e}}_\bullet) \in \mathcal{M}(\mathcal{B})$ is the \mathbb{L}_2 -mean of \hat{g}_\bullet . For each $j \in \mathbb{R}$ the statistic $\dot{\mathbf{e}}_j := n^{1/2}(\hat{\mathbb{P}}_n(\bar{\mathbf{e}}_j) - \mathbb{P}_g(\bar{\mathbf{e}}_j)) \in \mathcal{M}(\mathcal{B}^{\otimes n})$ is centred, i.e. $\dot{\mathbf{e}}_j \in \mathbb{L}_1(\mathbb{P}_g^{\otimes n})$ with $\mathbb{P}_g^{\otimes n}(\dot{\mathbf{e}}_j) = 0$. Since $\bar{\mathbf{e}}_\bullet = (\bar{\mathbf{e}}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}^2)$ the stochastic process

$$\dot{\mathbf{e}}_\bullet = (\dot{\mathbf{e}}_j)_{j \in \mathbb{R}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_g)(\bar{\mathbf{e}}_\bullet) = n^{1/2}(\hat{\mathbb{P}}_n(\bar{\mathbf{e}}_\bullet) - \mathbb{P}_g(\bar{\mathbf{e}}_\bullet)) \in \mathcal{M}(\mathcal{B}^{\otimes n} \otimes \mathcal{B})$$

satisfies Assumption §01102.11 and, by construction $\hat{g}_\bullet = g_\bullet + n^{-1/2}\dot{\mathbf{e}}_\bullet$ is a noisy version of g_\bullet . \square

§01102.17 **Notation.** Consider on the measurable space $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ the restriction $\lambda_{>0} \in \mathcal{M}_\sigma(\mathcal{B}_{>0})$ of the Lebesgue-measure λ on $\mathbb{R}_{>0}$, and for $c \in \mathbb{R}$ the σ -finite measure $x^c \lambda_{>0} \in \mathcal{M}_\sigma(\mathcal{B}_{>0})$ with Lebesgue-density $x^c \in \mathcal{M}(\mathcal{B}_{>0})$ given by $x \mapsto x^c(x) := x^c$. For $s \in \mathbb{R}_{\geq 1}$ introduce the *complex* vector space $\mathbb{L}_s(x^c) := \mathbb{L}_s(x^c \lambda_{>0}) := \mathbb{L}_s(\mathbb{R}_{>0}, \mathcal{B}_{>0}, x^c \lambda_{>0})$ of all *complex-valued $\mathbb{L}_s(x^c \lambda_{>0})$ -integrable functions*. Given the *complex* Hilbert space $\mathbb{L}_2 := \mathbb{L}_2(\lambda) := \mathbb{L}_2(\mathbb{R}, \mathcal{B}, \lambda)$ of all *complex-valued square-Lebesgue-integrable functions* let $M_c \in \mathbb{L}(\mathbb{L}_2(x^{2c-1}), \mathbb{L}_2)$ denote the *Mellin transform* satisfying

$$\begin{aligned} g_j := (M_c g)_j &= \int_{\mathbb{R}_{>0}} x^{c-1+i2\pi j} g(x) \lambda_{>0}(dx) = \int_{\mathbb{R}_{>0}} x^{i2\pi j} g(x) (x^{c-1} \lambda_{>0})(dx) \\ &= \int_{\mathbb{R}_{>0}} x^{-c+i2\pi j} g(x) (x^{2c-1} \lambda_{>0})(dx), \quad j \in \mathbb{R}, \quad \forall g \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1}). \end{aligned}$$

Introducing $\mathbf{x}_\bullet = (x_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ given by $x_j(x) := x^{-i2\pi j}$ for $x \in \mathbb{R}_{>0}$, $j \in \mathbb{R}$ we evidently have $\mathbf{x}_\bullet(x) \in \mathcal{B}$, $\bar{e}_j \circ \log = x^{i2\pi j} = \bar{x}_j \in \mathcal{B}_{>0}$, and

$$\begin{aligned} g_\bullet &= (g_j)_{j \in \mathbb{R}} = M_c g = (M_c g)_\bullet = ((M_c g)_j = x^{c-1} \lambda_{>0}(\bar{x}_j g))_{j \in \mathbb{R}} \\ &= (x^{2c-1} \lambda_{>0}(x^{-c} \bar{x}_j g))_{j \in \mathbb{R}}, \quad \forall g \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1}). \end{aligned}$$

Moreover, M_c is unitary with adjoint $M_c^* \in \mathbb{L}(\mathbb{L}_2, \mathbb{L}_2(x^{2c-1}))$ satisfying

$$\begin{aligned} \lambda((x^{2c-1} \lambda_{>0})(x^{-c} \bar{x}_j g) \bar{h}_\bullet) &= \lambda((M_c g)_\bullet \bar{h}_\bullet) = \langle M_c g, h_\bullet \rangle_{\mathbb{L}_2} = \langle g, M_c^* h_\bullet \rangle_{\mathbb{L}_2(x^{2c-1})} \\ &= (x^{2c-1} \lambda_{>0})(g \bar{M}_c^* h_\bullet) = (x^{2c-1} \lambda_{>0})(g \lambda(\bar{x}_j \bar{h}_\bullet) x^{-c}), \quad \forall g_\bullet \in \mathbb{L}_1 \cap \mathbb{L}_2 \end{aligned}$$

and hence $(M_c^* h_\bullet)(x) = \lambda(\mathbf{x}_\bullet(x) h_\bullet) x^{-c} = (\lambda(\mathbf{x}_\bullet h_\bullet) x^{-c})(x)$, $x \in \mathbb{R}_{>0}$ for all $h_\bullet \in \mathbb{L}_1 \cap \mathbb{L}_2$. For $g \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ we write $g_\bullet := (g_j := x^{2c-1} \lambda_{>0}(x^{-c} \bar{x}_j g) = \lambda_{>0}(x^{c-1} \bar{x}_j g))_{j \in \mathbb{R}} = M_c g$ such that $g = M_c^* g_\bullet$ (with a slight abuse of notation). We note that for each $c \in \mathbb{R}$ the complex-valued stochastic process $x^c \bar{x}_\bullet = (x^c \bar{x}_j)_{j \in \mathbb{R}}$ on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ is $\mathcal{B}_{>0} \otimes \mathcal{B}$ - \mathcal{B} -measurable, i.e. $x^c \bar{x}_\bullet = (x^c \bar{x}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ for short, and thus satisfies Assumption §01102.11. \square

§01102.18 **Density estimation on $\mathbb{R}_{>0}$.** Let $\mathbb{D}_2 \subseteq \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ with $\mathbb{L}_2(x^{2c-1}) =: \mathbb{G}$ (by the usual embedding of real-valued functions) be a set of densities on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ for some $c \in \mathbb{R}$ fixed and presumed to be *known in advance*. We denote for each density $g \in \mathbb{D}_2$ by $\mathbb{P}_g := g \lambda_{>0} \in \mathcal{W}(\mathcal{B}_{>0})$ the associated probability measure. We consider the statistical product experiment $(\mathbb{R}_{>0}^n, \mathcal{B}_{>0}^{\otimes n}, \mathbb{P}_{\mathbb{D}_2}^{\otimes n} := (\mathbb{P}_g^{\otimes n})_{g \in \mathbb{D}_2})$. Let $M_c \in \mathbb{L}(\mathbb{L}_2(x^{2c-1}), \mathbb{L}_2)$ be the Mellin transform (see **Notation** §01102.17). Evidently, for each $g \in \mathbb{D}_2 \subseteq \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ its Mellin transform $g_\bullet = (g_j)_{j \in \mathbb{R}} = M_c g$ for each $j \in \mathbb{R}$ satisfies

$$g_j = \lambda_{>0}(x^{c-1} \bar{x}_j g) = \mathbb{P}_g(x^{c-1} \bar{x}_j).$$

The complex-valued stochastic process $x^{c-1} \bar{x}_\bullet = (x^{c-1} \bar{x}_j)_{j \in \mathbb{R}}$ on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ is $(\mathcal{B}_{>0} \otimes \mathcal{B})$ - \mathcal{B} -measurable, i.e. $x^{c-1} \bar{x}_\bullet = (x^{c-1} \bar{x}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ for short. Similar to an Empirical mean model §01102.04 we define $\hat{g}_\bullet = (\hat{g}_j := \hat{\mathbb{P}}_n(x^{c-1} \bar{x}_j))_{j \in \mathbb{R}} = \hat{\mathbb{P}}_n(x^{c-1} \bar{x}_\bullet) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes n} \otimes \mathcal{B})$ where for each $j \in \mathbb{R}$

$$x = (x_i)_{i \in [n]} \mapsto \hat{g}_j(x) = (\hat{\mathbb{P}}_n(x^{c-1} \bar{x}_j))(x) = n^{-1} \sum_{i \in [n]} x^{c-1}(x_i) \bar{x}_j(x_i) = n^{-1} \sum_{i \in [n]} x_i^{c-1+i2\pi j}.$$

By construction $g_\bullet = (g_j = \mathbb{P}_g(x^{c-1} \bar{x}_j))_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B})$ is the \mathbb{L}_2 -mean of \hat{g}_\bullet . For each $j \in \mathbb{R}$ the statistic $\dot{e}_j := n^{1/2}(\hat{\mathbb{P}}_n(x^{c-1} \bar{x}_j) - \mathbb{P}_g(x^{c-1} \bar{x}_j)) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes n})$ is centred, i.e. $\dot{e}_j \in \mathbb{L}_1(\mathbb{P}_g^{\otimes n})$ with $\mathbb{P}_g^{\otimes n}(\dot{e}_j) = 0$. By exploiting $x^{c-1} \bar{x}_\bullet = (x^{c-1} \bar{x}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ the stochastic process

$$\dot{e}_\bullet = (\dot{e}_j)_{j \in \mathbb{R}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_g)(x^{c-1} \bar{x}_\bullet) = n^{1/2}(\hat{\mathbb{P}}_n(x^{c-1} \bar{x}_\bullet) - \mathbb{P}_g(x^{c-1} \bar{x}_\bullet)) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes n} \otimes \mathcal{B})$$

satisfies Assumption §01102.11 and, by construction $\hat{g}_\bullet = g_\bullet + n^{-1/2} \dot{e}_\bullet$ is a noisy version of g_\bullet . \square

§01|03 Statistical direct problem

§01103.01 **Assumption.** The Hilbert space $\mathbb{J} = \mathbb{L}_2(\mathcal{J}, \mathcal{J}, \nu)$ with σ -finite measure $\nu \in \mathcal{M}_\sigma(\mathcal{J})$ and the surjective partial isometries $\mathbf{V} \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ and $\mathbf{U} := \mathbf{A} = \mathbf{V}\mathbf{T} \in \mathbb{L}(\mathbb{H}, \mathbb{J})$, i.e. $\mathbf{V}\mathbf{V}^* = \text{id}_{\mathbb{J}} = \mathbf{U}\mathbf{U}^*$, are fixed and presumed to be *known in advance*. \square

§01103.02 **Notation.** Under Assumption §01103.01 we consider the reconstruction of $\theta = \mathbf{U}\theta \in \mathbb{J}$ (or in equal $\theta = \mathbf{U}^* \theta \in \mathbb{H}$) from a noisy version of $g = \mathbf{V}g = \mathbf{A}\theta = \mathbf{U}\theta = \theta \in \mathbb{J}$. Keep in mind, that we identify the equivalence class and its representative θ . \square

§01103.03 **Statistical direct problem.** Consider as in Definition §01102.03 a stochastic process $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ satisfying Assumption §01101.04 with mean zero and a sample size $n \in \mathbb{N}$. Under Assumption §01103.01 the observable noisy image has \mathbb{J} -mean $\theta = U\theta \in \mathbb{J}$, takes the form $\widehat{\theta} = \theta + n^{-1/2}\dot{\epsilon}$ and is called a *noisy version* of the parameter $\theta \in \mathbb{H}$, or *noisy parameter* for short. We denote by \mathbb{P}_θ^n the distribution of $\widehat{\theta}$. If $\dot{\epsilon}$ admits (possibly depending on θ) a covariance function, say $\text{cov}_{\dot{\epsilon}}^\theta \in \mathcal{M}(\mathcal{J}^2)$, or a covariance operator, say $\Gamma_\theta \in \mathbb{L}(\mathbb{J})$, then we eventually write $\dot{\epsilon} \sim P_{(\theta, \text{cov}_{\dot{\epsilon}}^\theta)}$ and $\widehat{\theta} \sim P_{(\theta, n^{-1}\text{cov}_{\dot{\epsilon}}^\theta)}$ or $\dot{\epsilon} \sim P_{(\theta, \Gamma)}$ and $\widehat{\theta} \sim P_{(\theta, n^{-1}\Gamma)}$ for short. The reconstruction of $\theta \in \mathbb{J}$ (in equal $\theta = U^*\theta \in \mathbb{H}$) from its noisy version $\widehat{\theta} \sim \mathbb{P}_\theta^n$ is called a *statistical direct problem*. \square

§01103.04 **Direct empirical mean model.** Consider the reconstruction of $\theta \in \mathbb{J}$ (in equal $\theta = U^*\theta \in \mathbb{H}$) in an Empirical mean model as in §01102.04. Under Assumption §01103.01 the observable noisy image has \mathbb{J} -mean $U\theta = \theta \in \mathbb{J}$, i.e. it is a noisy version of the parameter, and takes the form an Empirical mean model as in §01102.04, that is $\widehat{\theta} = \theta + n^{-1/2}\dot{\epsilon}$ with error process $\dot{\epsilon} = n^{1/2}(\widehat{\mathbb{P}}_n(\psi) - \mathbb{P}_\theta(\psi)) \in \mathcal{M}(\mathcal{X}^{\otimes n} \otimes \mathcal{J})$ satisfying Assumption §01101.04. \square

§01103.05 **Direct sequence model (dSM).** Consider $\mathbb{J} = \ell_2 = \mathbb{L}_2(\mathcal{U}_k)$ as in §01101.14. Let $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}}$ be a sequence of real-valued random variables with mean zero and let $n \in \mathbb{N}$ be a sample size. The observable noisy version $\widehat{\theta} = \theta + n^{-1/2}\dot{\epsilon} \sim \mathbb{P}_\theta^n$ with ℓ_2 -mean $\theta \in \ell_2$ takes the form of a Sequence model as in §01102.05, that is

$$\widehat{\theta}_j = \theta_j + n^{-1/2}\dot{\epsilon}_j, \quad j \in \mathbb{N}. \quad (01.06)$$

If $\dot{\epsilon}$ admits a covariance function (possibly depending on θ), say $\text{cov}_{\dot{\epsilon}}^\theta \in \mathcal{M}(2^{\mathbb{N}}) = \mathbb{R}^{\mathbb{N}^2}$, then we eventually write $\widehat{\theta} \sim P_{(\theta, n^{-1}\text{cov}_{\dot{\epsilon}}^\theta)}$ for short. If in addition $\dot{\epsilon}$ admits a covariance operator $\Gamma_\theta \in \mathbb{L}(\ell_2)$ (an infinite matrix) then we write $\widehat{\theta} \sim P_{(\theta, n^{-1}\Gamma)}$. \square

§01103.06 **Gaussian direct sequence model (GdSM).** Let $\dot{B} := (\dot{B}_j)_{j \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ be a Gaussian white noise process. The observable noisy version $\widehat{\theta} = \theta + n^{-1/2}\dot{B}$ with ℓ_2 -mean $\theta \in \ell_2$ takes the form of a Gaussian sequence model as in §01102.06, that is

$$\widehat{\theta}_j = \theta_j + n^{-1/2}\dot{B}_j, \quad j \in \mathbb{N} \quad \text{with} \quad (\dot{B}_j)_{j \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}}. \quad (01.07)$$

We denote by N_θ^n the distribution of the stochastic process $\widehat{\theta}$. \square

§01|04 Diagonal statistical inverse problem

§01104.01 **Notation.** Consider the measure space $(\mathcal{J}, \mathcal{J}, \nu)$ and the Hilbert space $\mathbb{J} = \mathbb{L}_2(\nu)$ as in **Notation** §01101.01. For $w_\bullet \in \mathbb{R}^{\mathcal{J}}$ define the multiplication map $M_w : \mathbb{R}^{\mathcal{J}} \rightarrow \mathbb{R}^{\mathcal{J}}$ with $a_\bullet \mapsto M_w a_\bullet := w_\bullet a_\bullet := (w_j a_j)_{j \in \mathcal{J}}$. If $w_\bullet \in \mathcal{M}(\mathcal{J})$, i.e. w_\bullet is \mathcal{J} - \mathcal{B} -measurable, then we have $M_w : \mathcal{M}(\mathcal{J}) \rightarrow \mathcal{M}(\mathcal{J})$ too. If in addition $w_\bullet \in \mathbb{L}_\infty(\nu)$ then we have also $M_w \in \mathbb{L}(\mathbb{J})$ identifying again equivalence classes and representatives. We set

$$\mathbb{L}^{\mathbb{L}(\mathbb{J})} := \{M_w : w_\bullet \in \mathbb{L}_\infty(\nu)\} \subseteq \mathbb{L}(\mathbb{J})$$

noting that $\|M_w\|_{\mathbb{L}(\mathbb{J})} = \sup \{\|w_\bullet a_\bullet\|_{\mathbb{J}} : \|a_\bullet\|_{\mathbb{J}} \leq 1\} \leq \|w_\bullet\|_{\mathbb{L}_\infty(\nu)}$ for each $M_w \in \mathbb{L}^{\mathbb{L}(\mathbb{J})}$. Finally, given surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ we define

$$\mathbb{L}^{U,V}(\mathbb{L}^{\mathbb{L}(\mathbb{J})}) := V^*(\mathbb{L}^{\mathbb{L}(\mathbb{J})})U := \{V^*M_w U \in \mathbb{L}(\mathbb{H}, \mathbb{G}) : M_w \in \mathbb{L}^{\mathbb{L}(\mathbb{J})}\}.$$

As a consequence, for each $T \in \mathbb{L}^{U,V}(\mathbb{L}^{\mathbb{L}(\mathbb{J})})$ we have $VTU^* = M_w \in \mathbb{L}^{\mathbb{L}(\mathbb{J})}$ for some $w_\bullet \in \mathbb{L}_\infty(\nu)$. \square

§01104.02 **Notation.** For $A \in \mathcal{J}$ we denote by $\mathbb{1}^A = (\mathbb{1}_j^A)_{j \in \mathcal{J}}$ the indicator function where for each $j \in \mathcal{J}$, $\mathbb{1}_j^A = 1$ if $j \in A$ and $\mathbb{1}_j^A = 0$ otherwise. Obviously, $\mathbb{1}^A$ is \mathcal{J} - \mathcal{B} -measurable, i.e. $\mathbb{1}^A \in \mathcal{M}(\mathcal{J})$, and it belongs to $\mathbb{L}_\infty(\nu)$, and to $\mathbb{L}_2(\nu)$ whenever $\nu(A) \in \mathbb{R}_{\geq 0}$. Since $\{j\} \in \mathcal{J}$ we have $\mathbb{1}^{\{j\}} \in \mathcal{J}$ and $\mathbb{1}^{\{j\}} \in \mathbb{L}_\infty(\nu)$. In particular, it follows $\mathbb{1}_\cdot = \mathbb{1}^\mathcal{J} \in \mathbb{L}_\infty(\nu)$ and $M_\mathbb{1} \in \mathbb{L}(\mathbb{J})$. For each $w_\cdot \in \mathbb{L}_\infty(\nu)$ set

$$\mathbb{J}w_\cdot := \{\{a, w_\cdot\}_\nu : a \in \mathcal{L}_2(\nu)\} = \{a, w_\cdot : a \in \mathbb{J} = \mathbb{L}_2(\nu)\}$$

and hence in particular $\mathbb{J}\mathbb{1}^A = \{a, \mathbb{1}^A : a \in \mathbb{J}\}$. Given $0_\cdot = (0)_{j \in \mathcal{J}}$ for $w_\cdot \in \mathcal{M}(\mathcal{J})$ we write further

$$\mathcal{N}_w := \{w_\cdot = 0_\cdot\} := \{j \in \mathcal{J} : w_j = 0\} \in \mathcal{J},$$

and denote by $\text{dom}(M_w) = \{a \in \mathbb{J} : a, w_\cdot \in \mathbb{J}\}$, $\text{ran}(M_w) = \{a, w_\cdot : a \in \text{dom}(M_w) \subseteq \mathbb{J}\}$ and $\text{ker}(M_w) = \{a \in \mathbb{J} : \{a, w_\cdot\}_\nu = 0_\cdot\}$, respectively, the domain, range and nullspace of $M_w : \mathbb{J} \supseteq \text{dom}(M_w) \rightarrow \mathbb{J}$. We write $w_\cdot \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$, if $w_\cdot \in \mathcal{M}(\mathcal{J})$ and $\nu(\mathcal{N}_w) = 0$. Similarly, for $w_\cdot \in \mathcal{M}(\mathcal{J})$ with $\nu(\{w_\cdot \leq 0_\cdot\}) = 0$ we write $w_\cdot \in \mathcal{M}_{> 0, \nu}(\mathcal{J})$. \square

§01104.03 **Notation.** Consider the special case $(\mathcal{J}, \mathcal{J}, \nu) = (\mathbb{N}, 2^\mathbb{N}, \nu_\mathbb{N})$ where $(\mathbb{1}^{\{j\}})_{j \in \mathbb{N}}$ forms an orthonormal basis in ℓ_2 . For each infinite matrix $A_{\cdot, \cdot} \in \mathbb{L}(\ell_2) \subseteq \mathbb{R}^{\mathbb{N}^2} = \mathcal{M}(2^\mathbb{N})$ with

$$A_{\cdot, \cdot} = (A_{j, j_0} := \langle A_{\cdot, \cdot} \mathbb{1}^{\{j_0\}}, \mathbb{1}^{\{j\}} \rangle_{\ell_2})_{j, j_0 \in \mathbb{N}}$$

and for each $j, j_0 \in \mathbb{N}$ and $a \in \ell_2$ we have

$$A_{j, \cdot} := (A_{j, j_0})_{j_0 \in \mathbb{N}} = A_{\cdot, \cdot}^* \mathbb{1}^{\{j\}} \in \ell_2, \quad A_{\cdot, j_0} := (A_{j, j_0})_{j \in \mathbb{N}} = A_{\cdot, \cdot} \mathbb{1}^{\{j_0\}} \in \ell_2,$$

$$\text{and } \langle A_{j, \cdot}, a \rangle_{\ell_2} = \nu_\mathbb{N}(A_{j, \cdot} a) = \sum_{j_0 \in \mathbb{N}} A_{j, j_0} a_{j_0} = \langle A_{\cdot, \cdot} a, \mathbb{1}^{\{j\}} \rangle_{\ell_2} \in \mathbb{R}.$$

If $A_{\cdot, \cdot} \in \mathbb{L}(\ell_2)$ equals a multiplication operator $A_{\cdot, \cdot} = M_\mathfrak{s} \in \mathbb{L}(\ell_2)$ for some $\mathfrak{s}_\cdot \in \ell_\infty$ (where $\ell_\infty := \mathbb{L}_\infty(\nu_\mathbb{N})$ is the set of all bounded real-valued sequences with respect to the counting measure $\nu_\mathbb{N}$ over \mathbb{N}) then we call $A_{\cdot, \cdot} \in \mathbb{L}(\ell_2)$ *diagonal*. Note that $A_{\cdot, \cdot} \in \mathbb{L}(\ell_2)$ is diagonal if and only if $A_{j, j_0} = 0$ for all $j \in \mathbb{N}$ and $j_0 \in \mathbb{N}_j = \mathbb{N} \setminus \{j\}$. For each $T \in \mathbb{L}^{\text{U.V}}(\mathbb{L}(\ell_2))$ with $T_{\cdot, \cdot} = VTU^* = M_\mathfrak{s} \in \mathbb{L}(\ell_2)$, the sequence $\mathfrak{s}_\cdot \in \ell_\infty$ is called singular values of T and $(\mathfrak{s}_\cdot, U, V)$ singular value decomposition of T . In other words, each $T \in \mathbb{L}^{\text{U.V}}(\mathbb{L}(\mathbb{J}))$ is diagonal wrt. to U and V . \square

§01104.04 **Assumption.** For $\mathbb{J} = \mathbb{L}_2(\nu)$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$, fixed and presumed to be *known in advance*, $T \in \mathbb{L}^{\text{U.V}}(\mathbb{L}(\mathbb{J})) \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$ and hence $A = VT = M_\mathfrak{s}U$ or in equal $\mathfrak{s}_\cdot \in \mathbb{L}_\infty(\nu)$ is also presumed to be *known* where $g_\cdot = VT\theta = M_\mathfrak{s}U\theta = M_\mathfrak{s}\theta_\cdot = \mathfrak{s}_\cdot\theta_\cdot \in \mathbb{J}$ or in equal $g_\cdot \in \mathbb{J}\mathfrak{s}_\cdot$. \square

§01104.05 **Notation.** Under Assumption §01104.04 given $\mathfrak{s}_\cdot \in \mathbb{L}_\infty(\nu)$ and $g_\cdot \in \mathbb{J}\mathfrak{s}_\cdot$ we consider the reconstruction of $\theta_\cdot = U\theta \in \mathbb{J}$ (or in equal $\theta_\cdot = U^*\theta \in \mathbb{H}$) from a noisy version of $g_\cdot = Vg_\cdot = A\theta_\cdot = \mathfrak{s}_\cdot\theta_\cdot \in \mathbb{J}$. Keep in mind, that we identify the equivalence class and its representative g_\cdot . \square

§01104.06 **Diagonal statistical inverse problem.** Consider as in Definition §01102.03 a stochastic process $\dot{\epsilon}_\cdot = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ satisfying Assumption §01101.04 with mean zero and a sample size $n \in \mathbb{N}$. Under Assumption §01104.04, where $\mathfrak{s}_\cdot \in \mathbb{L}_\infty(\nu)$ is *known in advance*, the observable noisy image has \mathbb{J} -mean $g_\cdot = \mathfrak{s}_\cdot\theta_\cdot$ and takes the form $\widehat{g}_\cdot = g_\cdot + n^{-1/2}\dot{\epsilon}_\cdot = \mathfrak{s}_\cdot\theta_\cdot + n^{-1/2}\dot{\epsilon}_\cdot$. We denote by $\mathbb{P}_{q, \mathfrak{s}_\cdot}^n$ the distribution of \widehat{g}_\cdot . If $\dot{\epsilon}_\cdot$ admits (possibly depending on $g_\cdot = \mathfrak{s}_\cdot\theta_\cdot$) a covariance function, say $\text{cov}_{\mathfrak{s}_\cdot}^{\theta_\cdot} \in \mathcal{M}(\mathcal{J}^2)$, or a covariance operator, say $\Gamma_{q, \mathfrak{s}_\cdot} \in \mathbb{L}(\mathbb{J})$, then we eventually write $\dot{\epsilon}_\cdot \sim \mathbb{P}_{(q, \text{cov}_{\mathfrak{s}_\cdot}^{\theta_\cdot})}$ and $\widehat{g}_\cdot \sim \mathbb{P}_{(g_\cdot, n^{-1} \text{cov}_{\mathfrak{s}_\cdot}^{\theta_\cdot})}$ or $\dot{\epsilon}_\cdot \sim \mathbb{P}_{(q, \Gamma_{q, \mathfrak{s}_\cdot})}$ and $\widehat{g}_\cdot \sim \mathbb{P}_{(g_\cdot, n^{-1} \Gamma_{q, \mathfrak{s}_\cdot})}$ for short. The reconstruction of $\theta_\cdot \in \mathbb{J}$ (in equal $\theta_\cdot = U^*\theta \in \mathbb{H}$) from a noisy version $\widehat{g}_\cdot \sim \mathbb{P}_{q, \mathfrak{s}_\cdot}^n$ of the image $g_\cdot = \mathfrak{s}_\cdot\theta_\cdot \in \mathbb{J}$ is called a *diagonal statistical inverse problem*. \square

§01104.07 **Diagonal inverse empirical mean model (dieMM)**. Consider the reconstruction of $\theta \in \mathbb{J}$ (in equal $\theta = U^* \theta \in \mathbb{H}$) in an Empirical mean model as in §01102.04. Under Assumption §01104.04, where $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ is *known in advance*, the observable noisy image has \mathbb{J} -mean $Vg = g = \mathfrak{s}\theta \in \mathbb{J}\mathfrak{s} \subseteq \mathbb{J}$ and takes the form of an Empirical mean model as in §01102.04, that is $\hat{g} = \mathfrak{s}\theta + n^{-1/2}\dot{\epsilon}$ with error process

$$\dot{\epsilon} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_g)(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{P}_g(\psi)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J})$$

satisfying Assumption §01101.04. □

§01104.08 **Diagonal inverse sequence model (diSM)**. Consider $\mathbb{J} = \ell_2 = \mathbb{L}_2(\nu_{\mathbb{N}})$ as in §01101.14. Let $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}}$ be a sequence of real-valued random variables with mean zero and let $n \in \mathbb{N}$ be a sample size. Under Assumption §01104.04, where $\mathfrak{s} \in \mathbb{L}_\infty$ is *known in advance*, the observable noisy image has ℓ_2 -mean $g = \mathfrak{s}\theta \in \ell_2$ and takes the form of a Sequence model as in §01102.05, that is $\hat{g} = g + n^{-1/2}\dot{\epsilon} = \mathfrak{s}\theta + n^{-1/2}\dot{\epsilon}$ or in equal

$$\hat{g}_j = g_j + n^{-1/2}\dot{\epsilon}_j = \mathfrak{s}_j\theta_j + n^{-1/2}\dot{\epsilon}_j, \quad j \in \mathbb{N}. \quad (01.08)$$

We denote by $\mathbb{P}_{\theta|\mathfrak{s}}^n$ the distribution of \hat{g} . □

§01104.09 **Gaussian diagonal inverse sequence model (GdiSM)**. Let $\dot{B} := (\dot{B}_j)_{j \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ be a Gaussian white noise process. The observable noisy version $\hat{g} = g + n^{-1/2}\dot{B}$ with ℓ_2 -mean $g = \mathfrak{s}\theta$ takes the form of a Gaussian sequence model as in §01102.06, that is

$$\hat{g}_j = \mathfrak{s}_j\theta_j + n^{-1/2}\dot{B}_j, \quad j \in \mathbb{N} \quad \text{with} \quad (\dot{B}_j)_{j \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}}. \quad (01.09)$$

We denote by $N_{\theta|\mathfrak{s}}^n$ the distribution of the stochastic process \hat{g} . □

§01|04|01 Examples of diagonal inverse empirical mean models

§01104.10 **Diagonal inverse regression with uniform design**. Consider the measure space $([0, 1], \mathcal{B}_{[0,1]}, \lambda_{[0,1]})$ and the Hilbert space $\mathbb{L}_2(\lambda_{[0,1]})$ as in Model §01102.09. Let $T \in \mathbb{L}^{U,V}(\mathbb{L}^U(\ell_2)) \subseteq \mathbb{L}(\mathbb{H}, \mathbb{L}_2([0, 1]))$ be *known in advance*, i.e. $T_{\bullet,\bullet} = VTU^* = M_{\mathfrak{s}} \in \mathbb{L}^U(\ell_2)$ and in other words T has a *known* singular value decomposition (\mathfrak{s}, U, V) with sequence of singular values $\mathfrak{s} \in \ell_\infty$. Let (X, Y) be a $[0, 1] \times \mathbb{R}$ -valued random vector. As in Model §01102.09 we assume in what follows that the regressor X is uniformly distributed on the interval $[0, 1]$, i.e. $X \sim U_{[0,1]} = \lambda_{[0,1]} = \mathbb{P}^X$ and that given $T\theta = g \in \mathbb{L}_2([0, 1])$ for some $\theta \in \mathbb{H}$ the joint distribution of (X, Y) is given by $U_{\theta|T} := U_{[0,1]} \odot \mathbb{P}_{T\theta}^{Y|X}$ without fully specifying the regular conditional distribution $\mathbb{P}_{T\theta}^{Y|X}$ which however satisfies $\mathbb{P}_{T\theta}^{Y|X}(\text{id}_{\mathbb{R}}) = \mathbb{P}_{T\theta}(Y|X) = T\theta = g \in \mathbb{L}_2([0, 1])$. Keep in mind that we tactically identify X and Y with the coordinate map $\Pi_{[0,1]}$ and $\Pi_{\mathbb{R}}$, respectively, and thus (X, Y) with the identity $\text{id}_{[0,1] \times \mathbb{R}}$. Consequently, if $Y \in \mathbb{L}_2(U_{\theta|T})$ and $h \in \mathbb{L}_2(\mathbb{P}^X) = \mathbb{L}_2(\lambda_{[0,1]})$, hence $h(X) \in \mathbb{L}_2(U_{\theta|T})$, then we obtain $Yh(X) \in \mathbb{L}_1(U_{\theta|T})$ and

$$U_{\theta|T}(Yh(X)) = \mathbb{P}^X(\mathbb{P}_{T\theta}^{Y|X}(Y)h) = \mathbb{P}^X((T\theta)h) = \lambda_{[0,1]}((T\theta)h) = \langle T\theta, h \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} \in \mathbb{R}$$

identifying again equivalence classes and their representatives. We consider the statistical product experiment $(([0, 1] \times \mathbb{R})^n, (\mathcal{B}_{[0,1]} \otimes \mathcal{B}_{\mathbb{R}})^{\otimes n}, U_{\Theta \times \{T\}}^{\otimes n} := (U_{\theta|T}^{\otimes n})_{\theta \in \Theta})$ of size $n \in \mathbb{N}$ and for $\theta \in \Theta$ we denote by $((X_i, Y_i))_{i \in \llbracket n \rrbracket} \sim U_{\theta|T}^{\otimes n}$ an iid. sample of $(X, Y) \sim U_{\theta|T} = U_{[0,1]} \odot \mathbb{P}_{T\theta}^{Y|X}$. Keep in mind that $V \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2)$ and $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ are generalised Fourier series transform as in **Notation** §01102.07 which are fixed and *known in advanced*. Evidently, for each $\theta \in \Theta \subseteq \mathbb{H}$ the generalised Fourier coefficients $\theta = (\theta_j)_{j \in \mathbb{N}} = U\theta$ and $g = (g_j)_{j \in \mathbb{N}} = Vg = M_{\mathfrak{s}}\theta = \mathfrak{s}\theta$ satisfy

$$g_j = \mathfrak{s}_j\theta_j = \langle M_{\mathfrak{s}}\theta, \mathbf{1}^{\{j\}} \rangle_{\ell_2} = \langle T\theta, V^* \mathbf{1}^{\{j\}} \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \langle T\theta, v_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = U_{\theta|T}(Y v_j(X))$$

for each $j \in \mathbb{N}$. The stochastic process $\psi = (\psi_j(X, Y) := Y v_j(X))_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}})$ fulfils Assumption §01101.04 and $g = \mathfrak{s}\theta = U_{\theta\Gamma}(\psi)$. Similar to an Empirical mean model §01102.04 we define $\widehat{g} = (\widehat{g}_j := \widehat{\mathbb{P}}_n(\psi_j))_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}})$. By construction $g = \mathfrak{s}\theta = U_{\theta\Gamma}(\psi) \in \mathcal{M}(2^{\mathbb{N}})$ is the ℓ_2 -mean of \widehat{g} . For each $j \in \mathbb{N}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\widehat{\mathbb{P}}_n(\psi_j) - U_{\theta\Gamma}(\psi_j)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(U_{\theta\Gamma}^{\otimes n})$ with $U_{\theta\Gamma}^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $\psi \in (\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}}$ the stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}} = n^{1/2}(\widehat{\mathbb{P}}_n - U_{\theta\Gamma})(\psi) = n^{1/2}(\widehat{\mathbb{P}}_n(\psi) - U_{\theta\Gamma}(\psi)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}})$$

satisfies Assumption §01101.04, and by construction $\widehat{g} = g + n^{-1/2}\dot{\epsilon} = \mathfrak{s}\theta + n^{-1/2}\dot{\epsilon}$ is a noisy version of $g = \mathfrak{s}\theta$. \square

§01104.11 **Notation (Circular additive convolution)**. Let $\mathfrak{q}, \mathfrak{p}$ be two Lebesgue densities on $([0, 1], \mathcal{B}_{[0,1]})$, then their *circular additive convolution* is given by

$$g(y) := (\mathfrak{q} \otimes \mathfrak{p})(y) := \int_{[0,1]} \mathfrak{q}(y - x - \lfloor y - x \rfloor) \mathfrak{p}(x) \lambda_{[0,1]}(dx) \quad \forall y \in [0, 1].$$

We note that g is again a density on $([0, 1], \mathcal{B}_{[0,1]})$. Consider the *complex* Hilbert spaces $\mathbb{L}_2(\lambda_{[0,1]})$ and $\mathbb{J} := \ell_2(\mathbb{Z})$, the exponentials $e_j := (e_j)_{j \in \mathbb{Z}}$ given by $e_j(x) := \exp(-i2\pi jx)$ for $x \in [0, 1]$ and $j \in \mathbb{Z}$, and the *Fourier-series transform* $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2(\mathbb{Z}))$ with

$$g \mapsto Fg := g. = ((Fg)_j := g_j := \langle g, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})})_{j \in \mathbb{Z}}$$

(see **Notations** §01102.10 and §01102.12). Let $\varphi \in \mathbb{L}_1(\lambda_{[0,1]})$ and let $\lfloor \cdot \rfloor$ be the floor function, then the *circular additive convolution operator* $\otimes_{\varphi} : \mathbb{L}_2(\lambda_{[0,1]}) \rightarrow \mathbb{L}_2(\lambda_{[0,1]})$ with $h \mapsto \otimes_{\varphi} h$ defined by

$$(\otimes_{\varphi} h)(t) := (\varphi \otimes h)(t) := \int_{[0,1]} \varphi(t - s - \lfloor t - s \rfloor) h(s) \lambda_{[0,1]}(ds) \quad \forall t \in [0, 1]$$

satisfies $\|\otimes_{\varphi}\|_{\mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]})}) \leq \|\varphi\|_{\mathbb{L}_1(\lambda_{[0,1]})} = \lambda_{[0,1]}(|\varphi|)$. Since $\varphi \in \mathbb{L}_1(\lambda_{[0,1]})$ and for each $j \in \mathbb{Z}$, $e_j \in \mathbb{L}_{\infty}(\lambda_{[0,1]})$ we have $\varphi \bar{e}_j \in \mathbb{L}_1(\lambda_{[0,1]})$ too. More precisely, for each $j \in \mathbb{Z}$ we have

$$|\lambda_{[0,1]}(\varphi \bar{e}_j)| \leq \|\varphi \bar{e}_j\|_{\mathbb{L}_1(\lambda_{[0,1]})} \leq \|\varphi\|_{\mathbb{L}_1(\lambda_{[0,1]})} \|\bar{e}_j\|_{\mathbb{L}_{\infty}(\lambda_{[0,1]})} = \|\varphi\|_{\mathbb{L}_1(\lambda_{[0,1]})}$$

and hence $\varphi_j := \lambda_{[0,1]}(\varphi \bar{e}_j) = (\lambda_{[0,1]}(\varphi \bar{e}_j))_{j \in \mathbb{Z}} \in \ell_{\infty}(\mathbb{Z})$ with a slight abuse of notation satisfies $\|\varphi_j\|_{\ell_{\infty}(\mathbb{Z})} = \|\lambda_{[0,1]}(\varphi \bar{e}_j)\|_{\ell_{\infty}(\mathbb{Z})} \leq \|\varphi\|_{\mathbb{L}_1(\lambda_{[0,1]})}$. Obviously, if $\varphi \in \mathbb{L}_2(\lambda_{[0,1]})$ (implying $\varphi \in \mathbb{L}_1(\lambda_{[0,1]})$) then $\varphi_j = \lambda_{[0,1]}(\varphi \bar{e}_j) = (\lambda_{[0,1]}(\varphi \bar{e}_j))_{j \in \mathbb{Z}} = \langle \varphi, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = F\varphi \in \ell_2(\mathbb{Z})$. However, for each $\varphi \in \mathbb{L}_1(\lambda_{[0,1]})$ and $h \in \mathbb{L}_2(\lambda_{[0,1]})$ the *circular convolution theorem* states

$$(\otimes_{\varphi} h)_j = \langle \otimes_{\varphi} h, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \lambda_{[0,1]}(\varphi \bar{e}_j) \langle h, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \lambda_{[0,1]}(\varphi \bar{e}_j) (Fh)_j = \varphi_j h_j \quad \forall j \in \mathbb{Z},$$

or $(\otimes_{\varphi} h)_j = F(\otimes_{\varphi} h) = \lambda_{[0,1]}(\varphi \bar{e}_j) (Fh)_j = \varphi_j h_j$ in short. Consequently, (φ_j, F, F) is a singular value decomposition of \otimes_{φ} with $\varphi_j \in \ell_{\infty}(\mathbb{Z})$, and thus $\otimes_{\varphi} \in \mathbb{L}^{\mathbb{F}, \mathbb{F}}(\mathbb{L}^{\mathbb{M}}(\ell_2(\mathbb{Z}))) = F^*(\mathbb{L}^{\mathbb{M}}(\ell_2(\mathbb{Z})))F$. \square

§01104.12 **Circular density deconvolution**. Consider the *complex* Hilbert spaces $\mathbb{L}_2(\lambda_{[0,1]})$ and $\mathbb{J} := \ell_2(\mathbb{Z})$. Let \mathbb{D}_2 be a set of square-integrable Lebesgue densities on $([0, 1], \mathcal{B}_{[0,1]})$, and hence $\mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]}) \subseteq \mathbb{L}_1(\lambda_{[0,1]})$ (by the usual embedding of real-valued functions) as in **Notation** §01102.10. We denote for each density $\mathfrak{p} \in \mathbb{L}_1(\lambda_{[0,1]})$ by $\mathbb{P}_{\mathfrak{p}} := \mathfrak{p} \lambda_{[0,1]} \in \mathcal{W}(\mathcal{B}_{[0,1]})$ the associated probability measure. Given a Lebesgue density $\mathfrak{q} \in \mathbb{L}_1(\lambda_{[0,1]})$ presumed to be fixed and *known in advance* for each Lebesgue density $\mathfrak{p} \in \mathbb{D}_2$ we consider the Lebesgue density $g = \mathfrak{q} \otimes \mathfrak{p} \in \mathbb{L}_2(\lambda_{[0,1]})$ (see **Notation** §01104.11) and denote by $\mathbb{P}_{\mathfrak{p}|\mathfrak{q}} := (\mathfrak{q} \otimes \mathfrak{p}) \lambda_{[0,1]} = g \lambda_{[0,1]} \in \mathcal{W}(\mathcal{B}_{[0,1]})$ the associated probability measure. We consider the statistical product experiment $([0, 1]^n, \mathcal{B}_{[0,1]}^{\otimes n}, \mathbb{P}_{\mathbb{D}_2 \times \{\mathfrak{q}\}}^{\otimes n} := (\mathbb{P}_{\mathfrak{p}|\mathfrak{q}}^{\otimes n})_{\mathfrak{p} \in \mathbb{D}_2})$. Let $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2(\mathbb{Z}))$ be

the Fourier-series transform (see **Notation** §01102.12). Evidently, for $g \in \mathbb{L}_2(\lambda_{[0,1]}) \subseteq \mathbb{L}_1(\lambda_{[0,1]})$ its Fourier-series $g = (g_j)_{j \in \mathbb{Z}} = \mathbb{F}g$ satisfies $g_j = \lambda_{[0,1]}(g \bar{e}_j) = \mathbb{P}_{\mathbb{P}|q}(\bar{e}_j)$ for each $j \in \mathbb{Z}$. Moreover, considering the Fourier-series $\mathbb{p} = (\mathbb{p}_j)_{j \in \mathbb{Z}} = \mathbb{F}\mathbb{p}$ of $\mathbb{p} \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]})$ by the *circular convolution theorem* we have $g = \mathbb{F}(\mathbb{q} \otimes \mathbb{p}) = \mathbb{q} \cdot \mathbb{p}$ with $\mathbb{q} = \lambda_{[0,1]}(\mathbb{q} \bar{e}_j) \in \ell_\infty(\mathbb{Z})$ and $\mathbb{p} = \mathbb{F}\mathbb{p} \in \ell_2(\mathbb{Z})$ (see **Notation** §01104.11). Moreover, the stochastic process $\bar{e} = (\bar{e}_j)_{j \in \mathbb{Z}}$ on $([0, 1], \mathcal{B}_{[0,1]})$ is $(\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}})$ - \mathcal{B} -measurable, i.e. $\bar{e} \in \mathcal{M}(\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}})$ for short. We define $\hat{g} = (\hat{g}_j := \hat{\mathbb{P}}_n(\bar{e}_j))_{j \in \mathbb{Z}} = \hat{\mathbb{P}}_n(\bar{e}) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes n} \otimes 2^{\mathbb{Z}})$ similar to an Empirical mean model §01102.04 where for each $j \in \mathbb{Z}$

$$y = (y_i)_{i \in \llbracket n \rrbracket} \mapsto \hat{g}_j(y) = (\hat{\mathbb{P}}_n(\bar{e}_j))(y) = n^{-1} \sum_{i \in \llbracket n \rrbracket} \bar{e}_j(y_i) = n^{-1} \sum_{i \in \llbracket n \rrbracket} \exp(i2\pi j y_i).$$

By construction $g = \mathbb{q} \cdot \mathbb{p} = \mathbb{P}_{\mathbb{P}|q}(\bar{e}) \in \ell_2(\mathbb{Z})$ is the mean of \hat{g} . For each $j \in \mathbb{Z}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\hat{\mathbb{P}}_n(\bar{e}_j) - \mathbb{P}_{\mathbb{P}|q}(\bar{e}_j)) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(\mathbb{P}_{\mathbb{P}|q}^{\otimes n})$ with $\mathbb{P}_{\mathbb{P}|q}^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $\bar{e} = (\bar{e}_j)_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}})$ the complex valued stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{Z}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{\mathbb{P}|q})(\bar{e}) = n^{1/2}(\hat{\mathbb{P}}_n(\bar{e}) - \mathbb{P}_{\mathbb{P}|q}(\bar{e})) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes n} \otimes 2^{\mathbb{Z}})$$

satisfies Assumption §01102.11. Since $\hat{g}_j = g_j + n^{-1/2}\dot{\epsilon}_j = \mathbb{q}_j \mathbb{p}_j + n^{-1/2}\dot{\epsilon}_j$ for each $j \in \mathbb{R}$ by construction $\hat{g} = g + n^{-1/2}\dot{\epsilon} = \mathbb{q} \cdot \mathbb{p} + n^{-1/2}\dot{\epsilon}$ is a noisy version of $g = \mathbb{q} \cdot \mathbb{p}$.

§01104.13 **Circular regression deconvolution with uniform design.** Consider the measure space $([0, 1], \mathcal{B}_{[0,1]}, \lambda_{[0,1]})$ and the *complex* Hilbert spaces $\mathbb{L}_2(\lambda_{[0,1]})$ and $\ell_2(\mathbb{Z})$ as in **Notation** §01102.10. Let the circular convolution operator $\otimes_\varphi \in \mathbb{L}^{\mathbb{F}, \mathbb{F}}(\mathbb{L}(\ell_2(\mathbb{Z}))) \subseteq \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}))$ with $\varphi \in \mathbb{L}_1(\lambda_{[0,1]})$ be *known in advance* (see **Notation** §01104.11), i.e. $\mathbb{F} \otimes_\varphi \mathbb{F}^* = \mathbb{M}_\varphi \in \mathbb{L}(\ell_2(\mathbb{Z}))$ and in other words \otimes_φ has a *known* singular value decomposition $(\varphi, \mathbb{F}, \mathbb{F})$ with sequence of singular values $\varphi \in \ell_\infty(\mathbb{Z})$. Let (X, Y) be a $[0, 1) \times \mathbb{R}$ -valued random vector. As in Model §01102.14 we assume in what follows that the regressor X is uniformly distributed on the interval $[0, 1)$, i.e. $X \sim \mathbb{U}_{[0,1]} = \lambda_{[0,1]} = \mathbb{P}^X$ and that given $\otimes_\varphi f =: g \in \mathbb{L}_2(\lambda_{[0,1]})$ for some $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]}) = \mathbb{H}$ the joint distribution of (X, Y) is given by $\mathbb{U}_{f|\varphi} := \mathbb{U}_{[0,1]} \odot \mathbb{P}_{\otimes_\varphi f}^{Y|X}$ and without fully specifying the regular conditional distribution $\mathbb{P}_{\otimes_\varphi f}^{Y|X}$ which however satisfies $\mathbb{P}_{\otimes_\varphi f}^{Y|X}(\text{id}_{\mathbb{R}}) = \mathbb{P}_{\otimes_\varphi f}(Y|X) = \otimes_\varphi f = g \in \mathbb{L}_2(\lambda_{[0,1]})$. Keep in mind that we tactically identify X and Y with the coordinate map $\Pi_{[0,1]}$ and $\Pi_{\mathbb{R}}$, respectively, and thus (X, Y) with the identity $\text{id}_{[0,1] \times \mathbb{R}}$. Consequently, if $Y \in \mathbb{L}_2(\mathbb{U}_{f|\varphi})$ and $h \in \mathbb{L}_2(\mathbb{P}^X) = \mathbb{L}_2(\lambda_{[0,1]})$, hence $\bar{h}(X) \in \mathbb{L}_2(\mathbb{U}_{f|\varphi})$, then we obtain $Y \bar{h}(X) \in \mathbb{L}_1(\mathbb{U}_{f|\varphi})$ and

$$\mathbb{U}_{f|\varphi}(Y \bar{h}(X)) = \mathbb{P}^X(\mathbb{P}_{\otimes_\varphi f}^{Y|X}(Y) \bar{h}) = \mathbb{P}^X((\otimes_\varphi f) \bar{h}) = \lambda_{[0,1]}((\otimes_\varphi f) \bar{h}) = \langle \otimes_\varphi f, \bar{h} \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} \in \mathbb{C}$$

identifying again equivalence classes and their representatives. We consider the statistical product experiment $(([0, 1) \times \mathbb{R})^n, (\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n}, \mathbb{U}_{\mathbb{F} \times \{\varphi\}}^{\otimes n} := (\mathbb{U}_{f|\varphi}^{\otimes n})_{f \in \mathbb{F}_2})$ of size $n \in \mathbb{N}$ and for $f \in \mathbb{F}_2$ we denote by $((X_i, Y_i))_{i \in \llbracket n \rrbracket} \sim \mathbb{U}_{f|\varphi}^{\otimes n}$ an iid. sample of $(X, Y) \sim \mathbb{U}_{f|\varphi} = \mathbb{U}_{[0,1]} \odot \mathbb{P}_{\otimes_\varphi f}^{Y|X}$. Keep in mind that $\mathbb{F} \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2(\mathbb{Z}))$ is the Fourier series transform as in **Notation** §01102.12 which is fixed and evidently *known in advanced*. For each $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]}) = \mathbb{H}$ the Fourier coefficients $f = (f_j)_{j \in \mathbb{Z}} = \mathbb{F}f$ and $g = (g_j)_{j \in \mathbb{Z}} = \mathbb{F}g = \mathbb{F}(\otimes_\varphi f) = \mathbb{M}_\varphi f = \varphi \cdot f$ (**Notation** §01104.11) satisfy

$$g_j = \varphi_j f_j = \langle \mathbb{M}_\varphi f, \mathbf{1}^{\{j\}} \rangle_{\ell_2(\mathbb{Z})} = \langle \otimes_\varphi f, \mathbb{F}^* \mathbf{1}^{\{j\}} \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \langle \otimes_\varphi f, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \mathbb{U}_{f|\varphi}(Y \bar{e}_j(X))$$

for each $j \in \mathbb{Z}$. The stochastic process $\psi = (\psi_j(X, Y) := Y \bar{e}_j(X))_{j \in \mathbb{Z}} \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{Z}})$ fulfils Assumption §01101.04 and $g = \varphi \cdot f = \mathbb{U}_{f|\varphi}(\psi)$. Similar to an Empirical mean model §01102.04 we define $\hat{g} = (\hat{g}_j := \hat{\mathbb{P}}_n(\psi_j))_{j \in \mathbb{Z}} = \hat{\mathbb{P}}_n(\psi) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{Z}})$. By construction $g = \varphi \cdot f = \mathbb{U}_{f|\varphi}(\psi) \in \ell_2(\mathbb{Z})$ is the mean of \hat{g} . For each $j \in \mathbb{Z}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\hat{\mathbb{P}}_n(\psi_j) - \mathbb{U}_{f|\varphi}(\psi_j)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(\mathbb{U}_{f|\varphi}^{\otimes n})$ with $\mathbb{U}_{f|\varphi}^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $\psi \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{Z}})$ the stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{Z}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{U}_{f|\varphi})(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{U}_{f|\varphi}(\psi)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{Z}})$$

satisfies Assumption §01101.04, and by construction $\widehat{g} = g + n^{-1/2}\dot{\epsilon} = \varphi.f + n^{-1/2}\dot{\epsilon}$ is a noisy version of $g = \varphi.f$. \square

§01104.14 **Notation (Additive convolution on \mathbb{R}).** Let $\mathfrak{q}, \mathfrak{p}$ be two Lebesgue densities on $(\mathbb{R}, \mathcal{B})$, then their *additive convolution* is given by

$$g(y) := (\mathfrak{q} * \mathfrak{p})(y) := \int_{\mathbb{R}} \mathfrak{q}(y-x)\mathfrak{p}(x)\lambda(dx) = \int_{\mathbb{R}} \mathfrak{p}(y-x)\mathfrak{q}(x)\lambda(dx) \quad \text{for } \lambda\text{-a.e. } y \in \mathbb{R}.$$

We note that g is again a density on $(\mathbb{R}, \mathcal{B})$ (keep in mind that we identify representatives and equivalence classes). Consider the *complex* Hilbert space $\mathbb{L}_2 = \mathbb{L}_2(\lambda)$, the exponentials $\mathbf{e}_j := (\mathbf{e}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B})$ given by $\mathbf{e}_j(x) := \exp(-i2\pi jx)$ for $x, j \in \mathbb{R}$, and the *Fourier-Plancherel transform* $F \in \mathbb{L}(\mathbb{L}_2)$ satisfying

$$Fh = \mathbf{h}_\bullet = ((Fh)_j := h_j := \lambda(h\bar{\mathbf{e}}_j))_{j \in \mathbb{R}} = \lambda(h\bar{\mathbf{e}}_\bullet), \quad \forall h \in \mathbb{L}_1 \cap \mathbb{L}_2$$

(see **Notations** §01102.10 and §01102.15). Consider $\mathfrak{p} \in \mathbb{L}_p$ and $\mathfrak{q} \in \mathbb{L}_q$ with conjugate exponents ($1/p + 1/q = 1$) then the integral $\int_{\mathbb{R}} \mathfrak{p}(y-x)\mathfrak{q}(x)\lambda(dx)$ exists for all $y \in \mathbb{R}$ and hence $(\mathfrak{q} * \mathfrak{p})(y)$ is for all $y \in \mathbb{R}$ defined. In the case $\mathfrak{q} \in \mathbb{L}_1$ and $\mathfrak{p} \in \mathbb{L}_p$ with $p \in \overline{\mathbb{R}}_{\geq 1}$ the integral $(\mathfrak{q} * \mathfrak{p})(y)$ exists for λ -a.e. $y \in \mathbb{R}$ only. However, the λ -a.e.-defined function $\mathfrak{q} * \mathfrak{p}$ belongs to \mathbb{L}_p and satisfies $\|\mathfrak{q} * \mathfrak{p}\|_{\mathbb{L}_p} \leq \|\mathfrak{q}\|_{\mathbb{L}_1} \|\mathfrak{p}\|_{\mathbb{L}_p}$. Werner [2011] p.337 for $p = 2$ and general case $p \in \overline{\mathbb{R}}_{\geq 1}$ lecture notes P. Maréchal (Analyse pour les problèmes inverses d'imagerie). For $\varphi \in \mathbb{L}_1$ the *additive convolution operator* $*_\varphi : \mathbb{L}_2 \rightarrow \mathbb{L}_2$ with $h \mapsto *_\varphi h$ defined by

$$(*_\varphi h)(t) := (\varphi * h)(t) := \int_{\mathbb{R}} \varphi(t-s)h(s)\lambda(ds) \quad \text{for } \lambda\text{-a.e. } y \in \mathbb{R}$$

satisfies $\|*_\varphi\|_{\mathbb{L}(\mathbb{L}_2)} \leq \|\varphi\|_{\mathbb{L}_1} = \lambda(|\varphi|)$. Since $\varphi \in \mathbb{L}_1$ and for each $j \in \mathbb{R}$, $\mathbf{e}_j \in \mathbb{L}_\infty$ we have $\varphi\bar{\mathbf{e}}_j \in \mathbb{L}_1$ too. More precisely, for each $j \in \mathbb{R}$ we have

$$|\lambda(\varphi\bar{\mathbf{e}}_j)| \leq \|\varphi\bar{\mathbf{e}}_j\|_{\mathbb{L}_1} \leq \|\varphi\|_{\mathbb{L}_1} \|\bar{\mathbf{e}}_j\|_{\mathbb{L}_\infty} = \|\varphi\|_{\mathbb{L}_1}$$

and hence $\varphi_\bullet := \lambda(\varphi\bar{\mathbf{e}}_\bullet) = (\varphi_j := \lambda(\varphi\bar{\mathbf{e}}_j))_{j \in \mathbb{R}} \in \mathbb{L}_\infty$ with a slight abuse of notation satisfies $\|\varphi_\bullet\|_{\mathbb{L}_\infty} = \|\lambda(\varphi\bar{\mathbf{e}}_\bullet)\|_{\mathbb{L}_\infty} \leq \|\varphi\|_{\mathbb{L}_1}$. Obviously, if $\varphi \in \mathbb{L}_1 \cap \mathbb{L}_2$ then $\varphi_\bullet = \lambda(\varphi\bar{\mathbf{e}}_\bullet) = F\varphi \in \mathbb{L}_2$. However, for each $\varphi \in \mathbb{L}_1$ and $h \in \mathbb{L}_1 \cap \mathbb{L}_2$ the *convolution theorem* states

$$(*_\varphi h)_j = \lambda((*_\varphi h)\bar{\mathbf{e}}_j) = \lambda(\varphi\bar{\mathbf{e}}_j)\lambda(h\bar{\mathbf{e}}_j) = \varphi_j(Fh)_j = \varphi_j h_j \quad \text{for } \lambda\text{-a.e. } j \in \mathbb{R}.$$

or $(*_\varphi h)_\bullet = F(*_\varphi h) = \lambda(\varphi\bar{\mathbf{e}}_\bullet)(Fh) = \varphi_\bullet h$. λ -a.s. in short. Consequently, (φ_\bullet, F, F) is a singular value decomposition of $*_\varphi$ with $\varphi_\bullet \in \mathbb{L}_\infty$, and thus $*_\varphi \in \mathbb{L}^{\mathbb{F}}(\mathbb{L}(\mathbb{L}_2)) = \mathbb{F}^*(\mathbb{L}(\mathbb{L}_2))\mathbb{F}$. \square

§01104.15 **Density additive deconvolution on \mathbb{R} .** Consider the *complex* Hilbert space $\mathbb{L}_2 = \mathbb{L}_2(\lambda)$. Let \mathbb{D}_2 be a set of square-integrable Lebesgue densities on $(\mathbb{R}, \mathcal{B})$, and hence $\mathbb{D}_2 \subseteq \mathbb{L}_2 \cap \mathbb{L}_1$ (by the usual embedding of real-valued functions) as in **Notation** §01102.10. We denote for each density $\mathfrak{p} \in \mathbb{L}_1$ by $\mathbb{P}_\mathfrak{p} := \mathfrak{p}\lambda \in \mathcal{W}(\mathcal{B})$ the associated probability measure. Given a Lebesgue density $\mathfrak{q} \in \mathbb{L}_1(\lambda)$ presumed to be fixed and *known in advance* for each Lebesgue density $\mathfrak{p} \in \mathbb{D}_2$ we consider the Lebesgue density $g = *_\mathfrak{q}\mathfrak{p} = \mathfrak{q} * \mathfrak{p} \in \mathbb{L}_2 \cap \mathbb{L}_1$ (see **Notation** §01104.14) and denote by $\mathbb{P}_{\mathfrak{p}|\mathfrak{q}} := (\mathfrak{q} * \mathfrak{p})\lambda = g\lambda \in \mathcal{W}(\mathcal{B})$ the associated probability measure. We consider the statistical product experiment $(\mathbb{R}^n, \mathcal{B}^{\otimes n}, \mathbb{P}_{\mathbb{D}_2 \times \{\mathfrak{q}\}}^{\otimes n} := (\mathbb{P}_{\mathfrak{p}|\mathfrak{q}}^{\otimes n})_{\mathfrak{p} \in \mathbb{D}_2})$. Let $F \in \mathbb{L}(\mathbb{L}_2)$ be the Fourier-Plancherel transform (see **Notation** §01102.15). Evidently, for $g \in \mathbb{L}_2 \cap \mathbb{L}_1$ its Fourier-Plancherel transform $\mathbf{g}_\bullet = (g_j)_{j \in \mathbb{R}} = Fg$ satisfies $g_j = \lambda(g\bar{\mathbf{e}}_j) = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(\bar{\mathbf{e}}_j)$ for all $j \in \mathbb{R}$. Moreover, considering the Fourier-Plancherel transform $\mathfrak{p}_\bullet = (\mathfrak{p}_j)_{j \in \mathbb{R}} = F\mathfrak{p}$ of $\mathfrak{p} \in \mathbb{D}_2 \subseteq \mathbb{L}_2 \cap \mathbb{L}_1$ by the *additive convolution theorem* we have $\mathbf{g}_\bullet = F(*_\mathfrak{q}\mathfrak{p}) = \lambda(\mathfrak{q}\bar{\mathbf{e}}_\bullet)(F\mathfrak{p}) = \mathfrak{q}_\bullet \mathfrak{p}_\bullet$ λ -a.s. with $\mathfrak{q}_\bullet = \lambda(\mathfrak{q}\bar{\mathbf{e}}_\bullet) \in \mathbb{L}_\infty$

and $\mathbb{p} = F\mathbb{p} \in \mathbb{L}_2$ (see [Notation §01104.14](#)). Moreover, the stochastic process $\bar{\epsilon} = (\bar{\epsilon}_j)_{j \in \mathbb{R}}$ on $(\mathbb{R}, \mathcal{B})$ is \mathcal{B}^2 - \mathcal{B} -measurable, i.e. $\bar{\epsilon} \in \mathcal{M}(\mathcal{B}^2)$ for short. We define

$$\hat{g} = (\hat{g}_j := \hat{\mathbb{P}}_n(\bar{\epsilon}_j))_{j \in \mathbb{R}} = \hat{\mathbb{P}}_n(\bar{\epsilon}) \in \mathcal{M}(\mathcal{B}^{\otimes n} \otimes \mathcal{B})$$

similar to an Empirical mean model [§01102.04](#) where for each $j \in \mathbb{R}$

$$y = (y_i)_{i \in \llbracket n \rrbracket} \mapsto \hat{g}_j(y) = (\hat{\mathbb{P}}_n(\bar{\epsilon}_j))(y) = n^{-1} \sum_{i \in \llbracket n \rrbracket} \bar{\epsilon}_j(y_i) = n^{-1} \sum_{i \in \llbracket n \rrbracket} \exp(i2\pi j y_i).$$

By construction $g = \mathbb{q} \boxtimes \mathbb{p} = \mathbb{P}_{\mathbb{p}|\mathbb{q}}(\bar{\epsilon}) \in \mathbb{L}_2$ is the mean of \hat{g} . For each $j \in \mathbb{R}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\hat{\mathbb{P}}_n(\bar{\epsilon}_j) - \mathbb{P}_{\mathbb{p}|\mathbb{q}}(\bar{\epsilon}_j)) \in \mathcal{M}(\mathcal{B}^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(\mathbb{P}_{\mathbb{p}|\mathbb{q}}^{\otimes n})$ with $\mathbb{P}_{\mathbb{p}|\mathbb{q}}^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $\bar{\epsilon} = (\bar{\epsilon}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}^2)$ the complex valued stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{R}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{\mathbb{p}|\mathbb{q}})(\bar{\epsilon}) = n^{1/2}(\hat{\mathbb{P}}_n(\bar{\epsilon}) - \mathbb{P}_{\mathbb{p}|\mathbb{q}}(\bar{\epsilon})) \in \mathcal{M}(\mathcal{B}^{\otimes n} \otimes \mathcal{B})$$

satisfies Assumption [§01102.11](#), and by construction $\hat{g} = g + n^{-1/2}\dot{\epsilon} = \mathbb{q} \boxtimes \mathbb{p} + n^{-1/2}\dot{\epsilon}$ is a noisy version of $g = \mathbb{q} \boxtimes \mathbb{p}$.

[§01104.16](#) **Notation (Multiplicative convolution on $\mathbb{R}_{>0}$)**. Let \mathbb{q}, \mathbb{p} be two Lebesgue densities on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ satisfying , then their *multiplicative convolution* is given by

$$\begin{aligned} g(y) &:= (\mathbb{q} \boxtimes \mathbb{p})(y) := \int_{\mathbb{R}_{>0}} \mathbb{q}(y/x)\mathbb{p}(x)x^{-1}\lambda_{>0}(dx) \\ &= \int_{\mathbb{R}_{>0}} \mathbb{p}(y/x)\mathbb{q}(x)x^{-1}\lambda_{>0}(dx) \quad \text{for } \lambda_{>0}\text{-a.e. } y \in \mathbb{R}_{>0}. \end{aligned}$$

We note that g is again a density on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ (keep in mind that we identify representatives and equivalence classes). For $c \in \mathbb{R}$ fixed and *known in advance* consider the *complex* Hilbert spaces $\mathbb{L}_2(x^c) = \mathbb{L}_2(\mathbb{R}_{>0}, \mathcal{B}_{>0}, x^c\lambda_{>0})$ and $\mathbb{L}_2 = \mathbb{L}_2(\mathbb{R}, \mathcal{B}, \lambda)$, the kernel $x^c\bar{x}_j = (x^c\bar{x}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ given by $(x^c\bar{x}_j)(x) = x^c x^{j2\pi}$ for $x \in \mathbb{R}_{>0}, j \in \mathbb{R}$ and the *Mellin transform* $M_c \in \mathbb{L}(\mathbb{L}_2(x^{2c-1}), \mathbb{L}_2)$ (see [Notation §01102.17](#)) satisfying

$$M_c h = h_* = ((M_c h)_j = x^{c-1}\lambda_{>0}(\bar{x}_j h) = x^{2c-1}\lambda_{>0}(x^{-c}\bar{x}_j h))_{j \in \mathbb{R}}, \quad \forall h \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$$

(see [Notations §01102.10 and §01102.17](#)). Consider $\mathbb{p}, \mathbb{q} \in \mathbb{L}_1(x^{c-1}) = \mathbb{L}_1(\mathbb{R}_{>0}, \mathcal{B}_{>0}, x^{c-1}\lambda_{>0})$ then the integral $\int_{\mathbb{R}_{>0}} \mathbb{p}(y/x)\mathbb{q}(x)x^{-1}\lambda_{>0}(dx)$ exists for $\lambda_{>0}$ -a.e. $y \in \mathbb{R}_{>0}$ and hence $(\mathbb{q} \boxtimes \mathbb{p})(y)$ is for $\lambda_{>0}$ -a.e. $y \in \mathbb{R}$ defined and the $\lambda_{>0}$ -a.e.-defined function $\mathbb{q} \boxtimes \mathbb{p}$ belongs to $\mathbb{L}_1(x^{c-1})$ and satisfies $\|\mathbb{q} \boxtimes \mathbb{p}\|_{\mathbb{L}_1(x^{c-1})} \leq \|\mathbb{q}\|_{\mathbb{L}_1(x^{c-1})}\|\mathbb{p}\|_{\mathbb{L}_1(x^{c-1})}$. In case $\mathbb{p} \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ and $\mathbb{q} \in \mathbb{L}_1(x^{c-1})$ the integral $(\mathbb{q} \boxtimes \mathbb{p})(y)$ is for $\lambda_{>0}$ -a.e. $y \in \mathbb{R}$ defined and the $\lambda_{>0}$ -a.e.-defined function $\mathbb{q} \boxtimes \mathbb{p}$ belongs to $\mathbb{L}_2(x^{2c-1})$ and satisfies $\|\mathbb{q} \boxtimes \mathbb{p}\|_{\mathbb{L}_2(x^{2c-1})} \leq \|\mathbb{q}\|_{\mathbb{L}_1(x^{c-1})}\|\mathbb{p}\|_{\mathbb{L}_2(x^{2c-1})}$. (Phd thesis of S. Brenner Miguel [2023]). For $\varphi \in \mathbb{L}_1(x^{c-1})$ the *multiplicative convolution operator* $\boxtimes_\varphi : \mathbb{L}_2(x^{2c-1}) \rightarrow \mathbb{L}_2(x^{2c-1})$ with $h \mapsto \boxtimes_\varphi h$ defined by

$$(\boxtimes_\varphi h)(t) := (\varphi \boxtimes h)(t) := \int_{\mathbb{R}_{>0}} \varphi(t/s)h(s)s^{-1}\lambda_{>0}(ds) \quad \text{for } \lambda_{>0}\text{-a.e. } y \in \mathbb{R}_{>0}$$

satisfies $\|\boxtimes_\varphi\|_{\mathbb{L}(\mathbb{L}_2(x^{2c-1}))} \leq \|\varphi\|_{\mathbb{L}_1(x^{c-1})} = x^{c-1}\lambda_{>0}(|\varphi|)$. Since $\varphi \in \mathbb{L}_1(x^{c-1})$ and for each $j \in \mathbb{R}, x_j \in \mathbb{L}_\infty(\lambda_{>0})$ we have $\bar{x}_j\varphi \in \mathbb{L}_1(x^{c-1})$ too. More precisely, for each $j \in \mathbb{R}$ we have

$$|x^{c-1}\lambda_{>0}(\bar{x}_j\varphi)| \leq \|\bar{x}_j\varphi\|_{\mathbb{L}_1(x^{c-1})} \leq \|\varphi\|_{\mathbb{L}_1(x^{c-1})}\|\bar{x}_j\|_{\mathbb{L}_\infty(\lambda_{>0})} = \|\varphi\|_{\mathbb{L}_1(x^{c-1})}$$

and hence $\varphi_{\cdot} := x^{c-1}\lambda_{>0}(\overline{x}_{\cdot}\varphi) = (\varphi_j := x^{c-1}\lambda_{>0}(\overline{x}_j\varphi))_{j \in \mathbb{R}} \in \mathbb{L}_{\infty}$ with a slight abuse of notation satisfies $\|\varphi_{\cdot}\|_{\mathbb{L}_{\infty}} = \|x^{c-1}\lambda_{>0}(\overline{x}_{\cdot}\varphi)\|_{\mathbb{L}_{\infty}} \leq \|\varphi\|_{\mathbb{L}_1(x^{c-1})} = x^{c-1}\lambda_{>0}(|\varphi|)$. Obviously, if $\varphi \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ then $\varphi_{\cdot} = x^{c-1}\lambda_{>0}(\overline{x}_{\cdot}\varphi) = (x^{2c-1}\lambda_{>0}(x^{-c}\overline{x}_j\varphi))_{j \in \mathbb{R}} = M_c\varphi \in \mathbb{L}_2$. However, for each $\varphi \in \mathbb{L}_1(x^{c-1})$ and $h \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ the *convolution theorem* states

$$\begin{aligned} (\boxtimes_{\varphi} h)_j &= x^{c-1}\lambda_{>0}(\overline{x}_j(\boxtimes_{\varphi} h)) = x^{c-1}\lambda_{>0}(\overline{x}_j\varphi)x^{c-1}\lambda_{>0}(\overline{x}_j h) \\ &= x^{c-1}\lambda_{>0}(\overline{x}_j\varphi)(M_c h)_j = \varphi_j h_j \quad \text{for } \lambda\text{-a.e. } j \in \mathbb{R}. \end{aligned}$$

or $(\boxtimes_{\varphi} h)_{\cdot} = M_c(\boxtimes_{\varphi} h) = x^{c-1}\lambda_{>0}(\overline{x}_{\cdot}\varphi)(M_c h) = \varphi_{\cdot} h_{\cdot}$ λ -a.s. in short. Consequently, $(\varphi_{\cdot}, M_c, M_c)$ is a singular value decomposition of \boxtimes_{φ} with $\varphi_{\cdot} \in \mathbb{L}_{\infty}$, and thus $\boxtimes_{\varphi} \in \mathbb{L}^{M_c, M_c}(\mathbb{L}^{\infty}(\mathbb{L}_2)) = M_c^*(\mathbb{L}^{\infty}(\mathbb{L}_2))M_c$. \square

§01104.17 **Density multiplicative deconvolution on $\mathbb{R}_{>0}$.** Consider the *complex* Hilbert spaces $\mathbb{L}_2(x^{2c-1}) = \mathbb{L}_2(\mathbb{R}_{>0}, \mathcal{B}_{>0}, x^{2c-1}\lambda_{>0})$ and $\mathbb{L}_2 = \mathbb{L}_2(\lambda)$. Let $\mathbb{D}_2 \subseteq \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ be a set of Lebesgue-densities on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ (by the usual embedding of real-valued functions) as in **Notation** §01102.10. We denote for each Lebesgue density \mathfrak{p} on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ by $\mathbb{P}_{\mathfrak{p}} := \mathfrak{p}\lambda_{>0} \in \mathcal{W}(\mathcal{B}_{>0})$ the associated probability measure. Given a Lebesgue density $\mathfrak{q} \in \mathbb{L}_1(x^{c-1})$ presumed to be fixed and *known in advance* for each Lebesgue density $\mathfrak{p} \in \mathbb{D}_2$ we consider the Lebesgue density $g = \boxtimes_{\mathfrak{q}}\mathfrak{p} = \mathfrak{q}\boxtimes\mathfrak{p} \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ (see **Notation** §01104.16) and denote by $\mathbb{P}_{\mathfrak{p}|\mathfrak{q}} := (\mathfrak{q}\boxtimes\mathfrak{p})\lambda_{>0} = g\lambda_{>0} \in \mathcal{W}(\mathcal{B}_{>0})$ the associated probability measure. We consider the statistical product experiment $(\mathbb{R}_{>0}, \mathcal{B}_{>0}^{\otimes n}, \mathbb{P}_{\mathbb{D}_2 \times \{\mathfrak{q}\}}^{\otimes n} := (\mathbb{P}_{\mathfrak{p}|\mathfrak{q}}^{\otimes n})_{\mathfrak{p} \in \mathbb{D}_2})$. Let $M_c \in \mathbb{L}(\mathbb{L}_2(x^{2c-1}), \mathbb{L}_2)$ be the Mellin transform (see **Notation** §01102.17). Evidently, for $g \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ its Mellin transform $g_{\cdot} = (g_j)_{j \in \mathbb{R}} = M_c g$ satisfies $g_j = x^{c-1}\lambda_{>0}(\overline{x}_j g) = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(x^{c-1}\overline{x}_j)$ for all $j \in \mathbb{R}$. Moreover, considering the Mellin transform $\mathfrak{p}_{\cdot} = (\mathfrak{p}_j)_{j \in \mathbb{R}} = M_c \mathfrak{p}$ of $\mathfrak{p} \in \mathbb{D}_2 \subseteq \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ by the *multiplicative convolution theorem* we have $g_{\cdot} = M_c(\boxtimes_{\mathfrak{q}}\mathfrak{p}) = x^{c-1}\lambda_{>0}(\overline{x}_{\cdot}\mathfrak{q})(M_c \mathfrak{p}) = \mathfrak{q}_{\cdot}\mathfrak{p}_{\cdot}$ λ -a.s. with $\mathfrak{q}_{\cdot} = x^{c-1}\lambda_{>0}(\overline{x}_{\cdot}\mathfrak{q}) \in \mathbb{L}_{\infty}$ and $\mathfrak{p}_{\cdot} = M_c \mathfrak{p} \in \mathbb{L}_2$ (see **Notation** §01104.16). Moreover, the complex-valued stochastic process $x^{c-1}\overline{x}_{\cdot} = (x^{c-1}\overline{x}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ is $\mathcal{B}_{>0} \otimes \mathcal{B}$ -measurable, i.e. $x^{c-1}\overline{x}_{\cdot} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ for short. We define

$$\widehat{g}_{\cdot} = (\widehat{g}_j := \widehat{\mathbb{P}}_n(x^{c-1}\overline{x}_j))_{j \in \mathbb{R}} = \widehat{\mathbb{P}}_n(x^{c-1}\overline{x}_{\cdot}) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes n} \otimes \mathcal{B})$$

similar to an Empirical mean model §01102.04 where for each $j \in \mathbb{R}$

$$y = (y_i)_{i \in [n]} \mapsto \widehat{g}_j(y) = (\widehat{\mathbb{P}}_n(x^{c-1}\overline{x}_j))(y) = n^{-1} \sum_{i \in [n]} (x^{c-1}\overline{x}_j)(y_i) = n^{-1} \sum_{i \in [n]} y_i^{c-1+i2\pi j}.$$

By construction $g_{\cdot} = \mathfrak{q}_{\cdot}\mathfrak{p}_{\cdot} = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(x^{c-1}\overline{x}_{\cdot}) \in \mathbb{L}_2$ is the mean of \widehat{g}_{\cdot} . For each $j \in \mathbb{R}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\widehat{\mathbb{P}}_n(x^{c-1}\overline{x}_j) - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(x^{c-1}\overline{x}_j)) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(\mathbb{P}_{\mathfrak{p}|\mathfrak{q}}^{\otimes n})$ with $\mathbb{P}_{\mathfrak{p}|\mathfrak{q}}^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $x^{c-1}\overline{x}_{\cdot} = (x^{c-1}\overline{x}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ the complex valued stochastic process

$$\dot{\epsilon}_{\cdot} = (\dot{\epsilon}_j)_{j \in \mathbb{R}} = n^{1/2}(\widehat{\mathbb{P}}_n - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}})(x^{c-1}\overline{x}_{\cdot}) = n^{1/2}(\widehat{\mathbb{P}}_n(x^{c-1}\overline{x}_{\cdot}) - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(x^{c-1}\overline{x}_{\cdot})) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes n} \otimes \mathcal{B})$$

satisfies Assumption §01102.11, and by construction $\widehat{g}_{\cdot} = g_{\cdot} + n^{-1/2}\dot{\epsilon}_{\cdot} = \mathfrak{q}_{\cdot}\mathfrak{p}_{\cdot} + n^{-1/2}\dot{\epsilon}_{\cdot}$ is a noisy version of $g_{\cdot} = \mathfrak{q}_{\cdot}\mathfrak{p}_{\cdot}$.

§01|05 Non-diagonal statistical inverse problem

§01105.01 **Notation.** Consider the measure space $(\mathcal{J}, \mathcal{J}, \nu)$ and the Hilbert space $\mathbb{J} = \mathbb{L}_2(\nu)$ as in **Notation** §01101.01. For $T_{j_{\cdot}} \in \mathcal{M}(\mathcal{J}^2)$ denote for each $j, j_{\cdot} \in \mathcal{J}$ by $T_{j_{\cdot}} : \mathcal{J} \rightarrow \mathbb{R}$ and $T_{j_{\cdot}} : \mathcal{J} \rightarrow \mathbb{R}$ the map $j \mapsto T_{j_{\cdot}}(j)$ and $j_{\cdot} \mapsto T_{j_{\cdot}}(j_{\cdot})$, respectively. Then we have $T_{j_{\cdot}}, T_{j_{\cdot}} \in \mathcal{M}(\mathcal{J})$ for each $j, j_{\cdot} \in \mathcal{J}$. If in addition $T_{j_{\cdot}} \in \mathbb{J}$ for ν -a.e. $j \in \mathcal{J}$ then for each $a_{\cdot} \in \mathbb{J}$ it follows $(T_{j_{\cdot}}, a_{\cdot})_j := \langle T_{j_{\cdot}}, a_{\cdot} \rangle_{\mathbb{J}} = \nu(T_{j_{\cdot}}, a_{\cdot}) \in \mathbb{R}$ for ν -a.e. $j \in \mathcal{J}$ and thus $(T_{j_{\cdot}}, a_{\cdot})_{\cdot} : j \mapsto (T_{j_{\cdot}}, a_{\cdot})_j$ is ν -a.e. defined

and $(T_{\cdot, \cdot} a_{\cdot}) \in \mathcal{M}(\mathcal{J})$. If for each $a_{\cdot} \in \mathbb{J}$ in addition $\|(T_{\cdot, \cdot} a_{\cdot})\|_{\mathbb{J}}^2 = \nu((T_{\cdot, \cdot} a_{\cdot})^2) \in \mathbb{R}_{\geq 0}$ and hence $(T_{\cdot, \cdot} a_{\cdot}) \in \mathbb{J}$, then setting $a_{\cdot} \mapsto T a_{\cdot} := (T_{\cdot, \cdot} a_{\cdot})$ defines an *integral operator* $T : \mathbb{J} \rightarrow \mathbb{J}$ which we identify here and subsequently with its kernel $T_{\cdot, \cdot} \in \mathcal{M}(\mathcal{J}^2)$. Evidently, the operator $T_{\cdot, \cdot} : \mathbb{J} \rightarrow \mathbb{J}$ is bounded, i.e. $T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J})$, if $\|T_{\cdot, \cdot}\|_{\mathbb{L}(\mathbb{J})} = \sup \{\nu((T_{\cdot, \cdot} a_{\cdot})^2) : a_{\cdot} \in \mathbb{J}, \|a_{\cdot}\|_{\mathbb{J}} \leq 1\} \in \mathbb{R}_{\geq 0}$. We set $\mathbb{L}(\mathbb{J}) := \{T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J}) : \text{with kernel } T_{\cdot, \cdot} \in \mathcal{M}(\mathcal{J}^2)\}$. Finally, given surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ we define $\mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J})) := V^*(\mathbb{L}(\mathbb{J}))U := \{V^*T_{\cdot, \cdot}U \in \mathbb{L}(\mathbb{H}, \mathbb{G}) : T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J})\}$. As a consequence, for each $T \in \mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J}))$ we have $VTU^* = T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J})$ for some kernel $T_{\cdot, \cdot} \in \mathcal{M}(\mathcal{J}^2)$. In the special case $(\mathcal{J}, \mathcal{J}, \nu) = (\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$, where $\mathbb{R}^{\mathbb{N}^2} = \mathcal{M}(2^{\mathbb{N}^2})$ is the set of all infinite real-valued matrices, we have $\mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$ (compare [Notation §01104.03](#)). \square

§01105.02 **Assumption.** For $\mathbb{J} = \mathbb{L}_2(\nu)$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$, fixed and presumed to be known in advance, $T \in \mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J}))$ and hence $T_{\cdot, \cdot} = VTU^* \in \mathbb{L}(\mathbb{J})$ with kernel $T_{\cdot, \cdot} \in \mathcal{M}(\mathcal{J}^2)$ is also known where $g_{\cdot} = T_{\cdot, \cdot} \theta_{\cdot} \in \mathbb{J}$ or inequal $g_{\cdot} \in \text{ran}(T_{\cdot, \cdot}) = \{T_{\cdot, \cdot} a_{\cdot} : a_{\cdot} \in \mathbb{J}\}$. \square

§01105.03 **Notation.** Under Assumption §01105.02 given $T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J})$ and $g_{\cdot} \in \text{ran}(T_{\cdot, \cdot})$ we consider the reconstruction of $\theta_{\cdot} = U\theta \in \mathbb{J}$ (or in equal $\theta = U^*\theta \in \mathbb{H}$) from a noisy version of the image $g_{\cdot} = VTU^*\theta = T_{\cdot, \cdot} \theta_{\cdot} \in \mathbb{J}$. Keep in mind, that we identify the equivalence class and its representative g_{\cdot} . \square

§01105.04 **Non-diagonal statistical inverse problem.** Consider as in [Definition §01102.03](#) a stochastic process $\dot{\epsilon}_{\cdot} = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ satisfying Assumption §01101.04 with mean zero and a sample size $n \in \mathbb{N}$. Under Assumption §01105.02, where $T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J})$ with kernel $T_{\cdot, \cdot} \in \mathcal{M}(\mathcal{J}^2)$ is known in advance, the observable noisy image has \mathbb{J} -mean $g_{\cdot} = T_{\cdot, \cdot} \theta_{\cdot}$ and takes the form $\hat{g}_{\cdot} = g_{\cdot} + n^{-1/2} \dot{\epsilon}_{\cdot} = T_{\cdot, \cdot} \theta_{\cdot} + n^{-1/2} \dot{\epsilon}_{\cdot}$ or in equal

$$\hat{g}_j = g_j + n^{-1/2} \dot{\epsilon}_j = \langle T_{j, \cdot}, \theta_{\cdot} \rangle_{\mathbb{J}} + n^{-1/2} \dot{\epsilon}_j, \quad \nu\text{-a.e. } j \in \mathcal{J}. \quad (01.10)$$

We denote by $\mathbb{P}_{q, T_{\cdot, \cdot}}^n$ the distribution of \hat{g}_{\cdot} . If $\dot{\epsilon}_{\cdot}$ admits (possibly depending on $g = T\theta$) a covariance function, say $\text{cov}_{\dot{\epsilon}_{\cdot}}^{\mathbb{T}\theta} \in \mathcal{M}(\mathcal{J}^2)$, or a covariance operator, say $\Gamma_{\mathbb{T}\theta} \in \mathbb{L}(\mathbb{J})$, then we eventually write $\dot{\epsilon}_{\cdot} \sim P_{(q, \text{cov}_{\dot{\epsilon}_{\cdot}}^{\mathbb{T}\theta})}$ and $\hat{g}_{\cdot} \sim P_{(g, n^{-1} \text{cov}_{\dot{\epsilon}_{\cdot}}^{\mathbb{T}\theta})}$ or $\dot{\epsilon}_{\cdot} \sim P_{(q, \Gamma_{\mathbb{T}\theta})}$ and $\hat{g}_{\cdot} \sim P_{(g, n^{-1} \Gamma_{\mathbb{T}\theta})}$ for short. The reconstruction of $\theta_{\cdot} \in \mathbb{J}$ (in equal $\theta = U^*\theta \in \mathbb{H}$) from a noisy version $\hat{g}_{\cdot} \sim \mathbb{P}_g^n$ of the image $g_{\cdot} = T_{\cdot, \cdot} \theta_{\cdot} \in \mathbb{J}$ is called a *non-diagonal statistical inverse problem*. \square

§01105.05 **Non-diagonal inverse empirical mean model (nieMM).** Consider the reconstruction of $\theta_{\cdot} \in \mathbb{J}$ (in equal $\theta = U^*\theta \in \mathbb{H}$) in an Empirical mean model as in §01102.04. Under Assumption §01105.02, where $T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J})$ with kernel $T_{\cdot, \cdot} \in \mathcal{M}(\mathcal{J}^2)$ is known in advance, the observable noisy image has \mathbb{J} -mean $Vg = g_{\cdot} = T_{\cdot, \cdot} \theta_{\cdot} \in \mathbb{J}$ and takes the form of an Empirical mean model as in §01102.04, that is $\hat{g}_{\cdot} = T_{\cdot, \cdot} \theta_{\cdot} + n^{-1/2} \dot{\epsilon}_{\cdot}$ or in equal (01.10) with error process

$$\dot{\epsilon}_{\cdot} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_g)(\psi_{\cdot}) = n^{1/2}(\hat{\mathbb{P}}_n(\psi_{\cdot}) - \mathbb{P}_g(\psi_{\cdot})) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J})$$

satisfying Assumption §01101.04. \square

§01105.06 **Comment.** In the special case $\mathbb{J} = \ell_2$ (compare [Notation §01104.03](#)) a **Diagonal statistical inverse problem** §01104.06 with multiplication operator, i.e. $T_{\cdot, \cdot} = M_{\mathbb{1}_{\cdot}}$, is indeed the diagonal case of the statistical inverse problem given in Assumption §01105.02. Moreover, introducing $\mathbb{1}_{\cdot} := (1)_{j \in \mathbb{N}}$ the multiplication operator $\text{id}_{\cdot, \cdot} := M_{\mathbb{1}_{\cdot}} \in \mathbb{L}(\ell_2)$ with diagonal kernel $\text{id}_{\cdot, \cdot} \in \mathcal{M}(2^{\mathbb{N}^2})$ equals the identity on ℓ_2 , i.e. $\text{id}_{\ell_2} = \text{id}_{\cdot, \cdot}$. As a consequence for $\mathbb{J} = \ell_2$ a statistical direct problem as in [Definition §01103.03](#) is also a statistical inverse problem with known identity operator, i.e. $T_{\cdot, \cdot} = \text{id}_{\cdot, \cdot}$. \square

§01105.07 **Non-diagonal inverse sequence model (niSM)**. Consider $\mathbb{J} = \ell_2 = \mathbb{L}_2(\nu_{\mathbb{N}})$ as in §01101.14. Let $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}}$ be a sequence of real-valued random variables with mean zero and let $n \in \mathbb{N}$ be a sample size. Under Assumption §01105.02, where $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}(\ell_2)$ with kernel $\mathbb{T}_{\cdot, \cdot} \in \mathcal{M}(2^{\mathbb{N}})$ is known in advance, the observable noisy image has ℓ_2 -mean $g = \mathbb{T}_{\cdot, \cdot} \theta$ and takes the form of a Sequence model as in §01102.05, that is $\hat{g} = g + n^{-1/2} \dot{\epsilon} = \mathbb{T}_{\cdot, \cdot} \theta + n^{-1/2} \dot{\epsilon}$ or in equal

$$\hat{g}_j = g_j + n^{-1/2} \dot{\epsilon}_j = \langle \mathbb{T}_{j, \cdot}, \theta \rangle_{\ell_2} + n^{-1/2} \dot{\epsilon}_j, \quad j \in \mathbb{N}. \quad (01.11)$$

We denote by $\mathbb{P}_{\theta|\mathbb{T}}^n$ the distribution of \hat{g} . \square

§01105.08 **Gaussian non-diagonal inverse sequence model (GniSM)**. Let $\dot{B} := (\dot{B}_j)_{j \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}$ be a Gaussian white noise process. The observable noisy version $\hat{g} = g + n^{-1/2} \dot{B}$ with ℓ_2 -mean $g = \mathbb{T}_{\cdot, \cdot} \theta$ takes the form of a Gaussian sequence model as in §01102.06, that is

$$\hat{g}_j = \langle \mathbb{T}_{j, \cdot}, \theta \rangle_{\ell_2} + n^{-1/2} \dot{B}_j, \quad j \in \mathbb{N} \quad \text{with} \quad (\dot{B}_j)_{j \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}. \quad (01.12)$$

We denote by $\mathbb{N}_{\theta|\mathbb{T}}^n$ the distribution of the stochastic process \hat{g} . \square

§01|05|01 Examples of non-diagonal inverse empirical mean models

§01105.09 **Non-diagonal inverse regression with uniform design**. Consider the measure space $([0, 1], \mathcal{B}_{[0,1]}, \lambda_{[0,1]})$ and the *real* Hilbert space $\mathbb{L}_2(\lambda_{[0,1]})$ as in Model §01102.09. Let $\mathbb{T} \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}))$ and hence $\mathbb{T}_{\cdot, \cdot} = \mathbb{V}\mathbb{T}\mathbb{U}^* \in \mathbb{L}(\ell_2)$ with kernel $\mathbb{T}_{\cdot, \cdot} \in \mathcal{M}(2^{\mathbb{N}})$ be *known in advance*. Let (X, Y) be a $[0, 1] \times \mathbb{R}$ -valued random vector. As in Model §01102.09 we assume in what follows that the regressor X is uniformly distributed on the interval $[0, 1]$, i.e. $X \sim \mathbb{U}_{[0,1]} = \lambda_{[0,1]} = \mathbb{P}^X$ and that given $\mathbb{T}\theta = g \in \mathbb{L}_2([0, 1])$ for some $\theta \in \mathbb{H}$ the joint distribution of (X, Y) is given by $\mathbb{U}_{\theta|\mathbb{T}} := \mathbb{U}_{[0,1]} \odot \mathbb{P}_{\mathbb{T}\theta}^{Y|X}$ without fully specifying the regular conditional distribution $\mathbb{P}_{\mathbb{T}\theta}^{Y|X}$ which however satisfies $\mathbb{P}_{\mathbb{T}\theta}^{Y|X}(\text{id}_{\mathbb{R}}) = \mathbb{P}_{\mathbb{T}\theta}(Y|X) = \mathbb{T}\theta = g \in \mathbb{L}_2([0, 1])$. Keep in mind that we tactically identify X and Y with the coordinate map $\Pi_{[0,1]}$ and $\Pi_{\mathbb{R}}$, respectively, and thus (X, Y) with the identity $\text{id}_{[0,1] \times \mathbb{R}}$. Consequently, if $Y \in \mathbb{L}_2(\mathbb{U}_{\theta|\mathbb{T}})$ and $h \in \mathbb{L}_2(\mathbb{P}^X) = \mathbb{L}_2(\lambda_{[0,1]})$, hence $h(X) \in \mathbb{L}_2(\mathbb{U}_{\theta|\mathbb{T}})$, then we obtain $Yh(X) \in \mathbb{L}_1(\mathbb{U}_{\theta|\mathbb{T}})$ and

$$\mathbb{U}_{\theta|\mathbb{T}}(Yh(X)) = \mathbb{P}^X(\mathbb{P}_{\mathbb{T}\theta}^{Y|X}(Y)h) = \mathbb{P}^X((\mathbb{T}\theta)h) = \lambda_{[0,1]}((\mathbb{T}\theta)h) = \langle \mathbb{T}\theta, h \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} \in \mathbb{R}$$

identifying again equivalence classes and their representatives. We consider the statistical product experiment $(([0, 1] \times \mathbb{R})^n, (\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n}, \mathbb{U}_{\Theta \times \{\mathbb{T}\}}^{\otimes n} := (\mathbb{U}_{\theta|\mathbb{T}}^{\otimes n})_{\theta \in \Theta})$ of size $n \in \mathbb{N}$ and for $\theta \in \Theta$ we denote by $((X_i, Y_i))_{i \in [n]} \sim \mathbb{U}_{\theta|\mathbb{T}}^{\otimes n}$ an iid. sample of $(X, Y) \sim \mathbb{U}_{\theta|\mathbb{T}} = \mathbb{U}_{[0,1]} \odot \mathbb{P}_{\mathbb{T}\theta}^{Y|X}$. Keep in mind that $\mathbb{V} \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2)$ and $\mathbb{U} \in \mathbb{L}(\mathbb{H}, \ell_2)$ are generalised Fourier series transform as in **Notation** §01102.07 which are fixed and *known in advance*. Evidently, for each $\theta \in \Theta \subseteq \mathbb{H}$ the generalised Fourier coefficients $\theta = (\theta_j)_{j \in \mathbb{N}} = \mathbb{U}\theta$ and $g = (g_j)_{j \in \mathbb{N}} = \mathbb{V}g = \mathbb{T}_{\cdot, \cdot} \theta$ satisfy

$$g_j = \langle \mathbb{T}_{j, \cdot}, \theta \rangle_{\ell_2} = \langle \mathbb{T}_{\cdot, \cdot} \theta, \mathbb{1}^{[j]} \rangle_{\ell_2} = \langle \mathbb{T}\theta, \mathbb{V}^* \mathbb{1}^{[j]} \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \lambda_{[0,1]}((\mathbb{T}\theta)\mathbb{V}_j) = \mathbb{U}_{\theta|\mathbb{T}}(Y\mathbb{V}_j(X))$$

for each $j \in \mathbb{N}$. The stochastic process $\psi = (\psi_j(X, Y) := Y\mathbb{V}_j(X))_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}})$ fulfils Assumption §01101.04 and $g = \mathbb{T}_{\cdot, \cdot} \theta = \mathbb{U}_{\theta|\mathbb{T}}(\psi)$. Similar to an Empirical mean model §01102.04 we define $\hat{g} = (\hat{g}_j := \hat{\mathbb{P}}_n(\psi_j))_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}})$. By construction $g = \mathbb{T}_{\cdot, \cdot} \theta = \mathbb{U}_{\theta|\mathbb{T}}(\psi) \in \mathcal{M}(2^{\mathbb{N}})$ is the ℓ_2 -mean of \hat{g} . For each $j \in \mathbb{N}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\hat{\mathbb{P}}_n(\psi_j) - \mathbb{U}_{\theta|\mathbb{T}}(\psi_j)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(\mathbb{U}_{\theta|\mathbb{T}}^{\otimes n})$ with $\mathbb{U}_{\theta|\mathbb{T}}^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $\psi \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}})$ the stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{U}_{\theta|\mathbb{T}})(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{U}_{\theta|\mathbb{T}}(\psi)) \in \mathcal{M}((\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}})$$

satisfies Assumption §01101.04. Since $\hat{g}_j = g_j + n^{-1/2} \dot{\epsilon}_j = \langle \mathbb{T}_{j, \cdot}, \theta \rangle_{\ell_2} + n^{-1/2} \dot{\epsilon}_j$ for each $j \in \mathbb{N}$ by construction $\hat{g} = g + n^{-1/2} \dot{\epsilon} = \mathbb{T}_{\cdot, \cdot} \theta + n^{-1/2} \dot{\epsilon}$ is a noisy version of $g = \mathbb{T}_{\cdot, \cdot} \theta$. \square

§01105.10 **Regression with known design.** Consider the measure space $(\mathcal{D}, \mathcal{B}_\mathcal{D}, \lambda_\mathcal{D})$ where $\lambda_\mathcal{D}$ denotes the restriction of the Lebesgue measure to the Borel- σ -algebra $\mathcal{B}_\mathcal{D}$ over $\mathcal{D} \in \mathcal{B}$, and the *real* Hilbert space $\mathbb{L}_2(\lambda_\mathcal{D}) := \mathbb{L}_2(\mathcal{D}, \mathcal{B}_\mathcal{D}, \lambda_\mathcal{D})$ of square Lebesgue-integrable real-valued functions. Let (X, Y) be a $\mathcal{D} \times \mathbb{R}$ -valued random vector. We assume in what follows that the marginal distribution $\mathbb{P}^X \in \mathcal{W}(\mathcal{B}_\mathcal{D})$ of the regressor X admits a Lebesgue density $\varphi \in \mathbb{L}_1(\lambda_\mathcal{D})$ presumed to be fixed and *known in advance*, that is $\mathbb{P}^X = \varphi \lambda_\mathcal{D}$. For a real random variable $\xi \sim \mathbb{P}^\xi \in \mathcal{W}(\mathcal{B})$ and $a \in \mathbb{R}$ we denote by $\mathbb{P}_a^\xi \in \mathcal{W}(\mathcal{B})$ the distribution of $a + \xi$. We assume that for each $B \in \mathcal{B}$ the map $\mathbb{P}^\xi(B) : a \mapsto \mathbb{P}_a^\xi(B)$ is \mathcal{B} - $\mathcal{B}_{[0,1]}$ -measurable. Then $\mathbb{P}^\xi : \mathbb{R} \times \mathcal{B} \rightarrow [0, 1]$ with $(a, B) \mapsto \mathbb{P}_a^\xi(B)$ is a Markov kernel from $(\mathbb{R}, \mathcal{B})$ to $(\mathbb{R}, \mathcal{B})$. In this situation, for any $f \in \mathcal{M}(\mathcal{B}_\mathcal{D})$ the map $\mathbb{P}_{f(X)}^\xi : \mathcal{D} \times \mathcal{B} \rightarrow [0, 1]$ with $(x, B) \mapsto \mathbb{P}_{f(x)}^\xi(B)$ is a Markov kernel from $(\mathcal{D}, \mathcal{B}_\mathcal{D})$ to $(\mathbb{R}, \mathcal{B})$. If ξ and X are *independent* and $Y = f(X) + \xi$ for some $f \in \mathcal{M}(\mathcal{B}_\mathcal{D})$, which is assumed throughout this model, then $\mathbb{P}_{f(X)}^\xi$ is a regular version of the conditional distribution of Y given X , in symbols $\mathbb{P}_f^{Y|X} = \mathbb{P}_{f(X)}^\xi$. In other words there exists a \mathbb{P}^X -null set $\mathcal{N} \in \mathcal{B}_\mathcal{D}$ such that $\mathbb{P}_f^{Y|X=x}(B) = \mathbb{P}_{f(x)}^\xi(B)$ for all $B \in \mathcal{B}$ and $x \in \mathcal{N}^c$ (Witting [1985], Satz 129, p.130). In summary the joint distribution of (X, Y) is given by $\mathbb{P}_{f|\varphi}^{X,Y} := \varphi \lambda_\mathcal{D} \odot \mathbb{P}_{f(X)}^\xi$ without fully specifying the error distribution $\mathbb{P}^\xi \in \mathcal{W}(\mathcal{B})$ and thus the regular conditional distribution $\mathbb{P}_f^{Y|X} = \mathbb{P}_{f(X)}^\xi$. (Since $\lambda_\mathcal{D}$ dominates $\varphi \lambda_\mathcal{D}$ each representative of $\{f\}_{\lambda_\mathcal{D}}$ induces the same joint distribution $\mathbb{P}_{\{f\}_{\lambda_\mathcal{D}}|\varphi}^{X,Y} = \mathbb{P}_{f|\varphi}^{X,Y} \in \mathcal{W}(\mathcal{B}_\mathcal{D} \otimes \mathcal{B})$.) We tactically identify X and Y with the coordinate map $\Pi_\mathcal{D}$ and $\Pi_\mathbb{R}$, respectively, and thus (X, Y) with the identity $\text{id}_{\mathcal{D} \times \mathbb{R}}$ such that $\mathbb{P}_{f|\varphi} = \mathbb{P}_{f|\varphi}^{X,Y} \in \mathcal{W}(\mathcal{B}_\mathcal{D} \otimes \mathcal{B})$. Let in addition $\varphi \in \mathbb{L}_\infty(\lambda_\mathcal{D})$, then $M_\varphi \in \mathbb{L}(\mathbb{L}_2(\lambda_\mathcal{D}))$ with $h \mapsto M_\varphi h := \varphi h$. Note that then for each $h \in \mathbb{L}_2(\lambda_\mathcal{D})$ we have $M_\varphi h \in \mathbb{L}_2(\lambda_\mathcal{D})$ and hence for each representative $h \in \mathcal{L}_2(\varphi \lambda_\mathcal{D})$. (Since $\lambda_\mathcal{D}$ dominates $\varphi \lambda_\mathcal{D}$ for each $h \in \overline{\mathcal{M}}(\mathcal{B}_\mathcal{D})$ we have $\{h\}_{\lambda_\mathcal{D}} \subseteq \{h\}_{\varphi \lambda_\mathcal{D}}$. If $\lambda_\mathcal{D}$ and $\varphi \lambda_\mathcal{D}$ dominate mutually each other, i.e. they share the same null sets, then $\{h\}_{\varphi \lambda_\mathcal{D}} = \{h\}_{\lambda_\mathcal{D}}$ and hence $\mathbb{L}_2(\lambda_\mathcal{D}) \subseteq \mathbb{L}_2(\varphi \lambda_\mathcal{D})$.) If in addition $\mathbb{P}^\xi \in \mathbb{P}_{\{0\} \times \mathbb{R}_\mathcal{B}} \subseteq \mathcal{W}_2(\mathcal{B})$, i.e. ξ has mean zero and a finite second moment, and $f \in \mathbb{L}_2(\lambda_\mathcal{D})$, then for each representative $f \in \mathcal{L}_2(\varphi \lambda_\mathcal{D})$, $f(X) \in \mathcal{L}_2(\mathbb{P}_{f|\varphi})$ and $Y \in \mathcal{L}_2(\mathbb{P}_{f|\varphi})$ too. In particular it follows $\mathbb{P}_f^{Y|X}(\text{id}_\mathbb{R}) = \mathbb{P}_f(Y|X) = \{f\}_{\varphi \lambda_\mathcal{D}} \in \mathbb{L}_2(\mathbb{P}^X) = \mathbb{L}_2(\varphi \lambda_\mathcal{D})$. Consequently, for each $h \in \mathbb{L}_2(\lambda_\mathcal{D})$, hence $h(X) \in \mathcal{L}_2(\mathbb{P}_{f|\varphi})$ we obtain $Yh(X) \in \mathcal{L}_1(\mathbb{P}_{f|\varphi})$ and

$$\mathbb{P}_{f|\varphi}(Yh(X)) = \mathbb{P}^X(\mathbb{P}_f^{Y|X}(Y)h) = \varphi \lambda_\mathcal{D}(fh) = \lambda_\mathcal{D}(\varphi fh) = \langle M_\varphi f, h \rangle_{\mathbb{L}_2(\lambda_\mathcal{D})} \in \mathbb{R}$$

identifying again equivalence classes and their representatives. We note that $M_\varphi \in \mathbb{L}(\mathbb{L}_2(\lambda_\mathcal{D}))$ with density $\varphi \in \mathbb{L}_\infty(\lambda_\mathcal{D})$ is positive semi-definite, i.e. $M_\varphi \in \mathbb{L}^{\geq}(\mathbb{L}_2(\lambda_\mathcal{D}))$ and if in addition $\varphi \in \mathcal{M}_{>0, \lambda_\mathcal{D}}(\mathcal{B}_\mathcal{D})$ (i.e. $\varphi \in \mathcal{M}_{\geq 0}(\mathcal{B}_\mathcal{D})$ and $\lambda_\mathcal{D}(\mathcal{N}_\varphi) = 0$) then it is strictly positive definite, i.e. $M_\varphi \in \mathbb{L}^{\gg}(\mathbb{L}_2(\lambda_\mathcal{D}))$. Keep in mind that $U \in \mathbb{L}(\mathbb{L}_2(\lambda_\mathcal{D}), \ell_2)$ is generalised Fourier series transform as in **Notation** §01102.07 which is fixed and *known in advance*. Evidently, we have $M_{\cdot, \cdot}^\varphi := UM_\varphi U^* \in \mathbb{L}^{\geq}(\ell_2) \subseteq \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$ and for each $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_\mathcal{D})$ and $g := M_\varphi f \in \mathbb{L}_2(\lambda_\mathcal{D})$ the generalised Fourier coefficients $\underline{f} = (f_j)_{j \in \mathbb{N}} = Uf$ and $\underline{g} = (g_j)_{j \in \mathbb{N}} = Ug = M_{\cdot, \cdot}^\varphi \underline{f}$ for each $j \in \mathbb{N}$ satisfy

$$g_j = \langle M_{j, \cdot}^\varphi, \underline{f} \rangle_{\ell_2} = \langle M_{\cdot, j}^\varphi, \mathbf{1}_{\cdot}^{[j]} \rangle_{\ell_2} = \langle M_\varphi f, U^* \mathbf{1}_{\cdot}^{[j]} \rangle_{\mathbb{L}_2(\lambda_\mathcal{D})} = \langle M_\varphi f, u_j \rangle_{\mathbb{L}_2(\lambda_\mathcal{D})} = \mathbb{P}_{f|\varphi}(Y u_j(X)) \in \mathbb{R}$$

The stochastic process $\psi = (\psi_j(X, Y) := Y u_j(X))_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_\mathcal{D} \otimes \mathcal{B}) \otimes 2^\mathbb{N})$ fulfils Assumption §01101.04 and $\underline{g} = M_{\cdot, \cdot}^\varphi \underline{f} = \mathbb{P}_{f|\varphi}(\psi)$. Similar to an Empirical mean model §01102.04 we consider the statistical product experiment $((\mathcal{D} \times \mathbb{R})^n, (\mathcal{B}_\mathcal{D} \otimes \mathcal{B})^{\otimes n}, \mathbb{P}_{\bar{x} \times \{\varphi\}}^{\otimes n} := (\mathbb{P}_{f|\varphi}^{\otimes n})_{f \in \mathbb{F}_2})$ of size $n \in \mathbb{N}$ and for $f \in \mathbb{F}_2$ we denote by $((X_i, Y_i))_{i \in [n]} \sim \mathbb{P}_{f|\varphi}^{\otimes n}$ an iid. sample of $(X, Y) \sim \mathbb{P}_{f|\varphi} = \varphi \lambda_\mathcal{D} \odot \mathbb{P}_f^{Y|X}$. We define $\hat{\underline{g}} = (\hat{g}_j := \hat{\mathbb{P}}_n(\psi_j))_{j \in \mathbb{N}} = \hat{\mathbb{P}}_n(\psi) \in \mathcal{M}((\mathcal{B}_\mathcal{D} \otimes \mathcal{B})^{\otimes n} \otimes 2^\mathbb{N})$. By construction $\underline{g} = M_{\cdot, \cdot}^\varphi \underline{f} = \mathbb{P}_{f|\varphi}(\psi) \in \mathcal{M}(2^\mathbb{N})$ is the ℓ_2 -mean of $\hat{\underline{g}}$. For each $j \in \mathbb{N}$ the statistic $\dot{\epsilon}_j := n^{1/2}(\hat{\mathbb{P}}_n(\psi_j) - \mathbb{P}_{f|\varphi}(\psi_j)) \in \mathcal{M}((\mathcal{B}_\mathcal{D} \otimes \mathcal{B})^{\otimes n})$ is centred, i.e. $\dot{\epsilon}_j \in \mathbb{L}_1(\mathbb{P}_{f|\varphi}^{\otimes n})$ with $\mathbb{P}_{f|\varphi}^{\otimes n}(\dot{\epsilon}_j) = 0$, and exploiting $\psi \in \mathcal{M}((\mathcal{B}_\mathcal{D} \otimes \mathcal{B}) \otimes 2^\mathbb{N})$ the stochastic process

$$\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{f|\varphi})(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{P}_{f|\varphi}(\psi)) \in \mathcal{M}((\mathcal{B}_\mathcal{D} \otimes \mathcal{B})^{\otimes n} \otimes 2^\mathbb{N})$$

satisfies Assumption §01101.04. Since $\hat{g}_j = g_j + n^{-1/2} \dot{\epsilon}_j = \langle M_{j, \cdot}^\varphi, \underline{f} \rangle_{\ell_2} + n^{-1/2} \dot{\epsilon}_j$ for each $j \in \mathbb{N}$ by construction $\hat{\underline{g}} = \underline{g} + n^{-1/2} \dot{\epsilon} = M_{\cdot, \cdot}^\varphi \underline{f} + n^{-1/2} \dot{\epsilon}$ is a noisy version of $\underline{g} = M_{\cdot, \cdot}^\varphi \underline{f}$. \square

§02 Noisy image and noisy operator

§02100.01 **Assumption.** The Hilbert space $\mathbb{J} = \mathbb{L}_2(\mathcal{J}, \mathcal{J}, \nu)$ with σ -finite measure $\nu \in \mathcal{M}_\sigma(\mathcal{J})$, σ -algebra \mathcal{J} over \mathcal{J} containing all elementary events $\{j\}$, $j \in \mathcal{J}$, and the surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$, i.e. $UU^* = \text{id}_{\mathbb{J}} = VV^*$, are fixed and presumed to be known in advance. \square

§02|01 Noisy non-diagonal operator

§02101.01 **Notation.** Under Assumption §02100.01 we consider the reconstruction of $\theta = U\theta \in \mathbb{J}$ (or in equal $\theta = U^*\theta \in \mathbb{H}$) from noisy versions of $Vg = g = T_{\cdot,\cdot}\theta \in \mathbb{J}$ and $T_{\cdot,\cdot} = VTU^* \in \mathbb{L}_{\cdot}(\mathbb{J})$. \square

§02101.02 **Assumption.** The real-valued stochastic process $Y_{\cdot,\cdot} = (Y_{j|j_0})_{j,j_0 \in \mathcal{J}}$ on a common measurable space (Ω, \mathcal{A}) as a function $\Omega \times \mathcal{J}^2 \rightarrow \mathbb{R}$ with $(\omega, j, j_0) \mapsto Y_{j|j_0}(\omega)$ is $\mathcal{A} \otimes \mathcal{J}^2$ - \mathcal{B} -measurable, $Y_{\cdot,\cdot} \in \mathcal{M}(\mathcal{A} \otimes \mathcal{J}^2)$ for short. \square

§02101.03 **Noisy non-diagonal operator.** Let $\dot{\eta}_{\cdot,\cdot} = (\dot{\eta}_{j|j_0})_{j,j_0 \in \mathcal{J}}$ be a stochastic process satisfying Assumption §02101.02 with mean zero and let $k \in \mathbb{N}$ be a sample size. The stochastic process $\hat{T}_{\cdot,\cdot} = T_{\cdot,\cdot} + k^{-1/2}\dot{\eta}_{\cdot,\cdot}$ with mean kernel $T_{\cdot,\cdot} \in \mathcal{M}(\mathcal{J}^2)$ is called a *noisy version* of the non-diagonal operator $T_{\cdot,\cdot} = VTU^* \in \mathbb{L}_{\cdot}(\mathbb{J})$, or *noisy non-diagonal operator* for short. We denote by \mathbb{P}_T^k the distribution of $\hat{T}_{\cdot,\cdot}$. If $\dot{\eta}_{\cdot,\cdot}$ admits a covariance function (possibly depending on T), say $\text{cov} \in \mathcal{M}(\mathcal{J}^4)$, then we eventually write $\dot{\eta}_{\cdot,\cdot} \sim P_{(T_{\cdot,\cdot}, \text{cov})}$ and $\hat{T}_{\cdot,\cdot} \sim P_{(T_{\cdot,\cdot}, k^{-1} \text{cov})}$ for short. \square

§02101.04 **Empirical mean model.** For each $T \in \mathbb{T} \subseteq \mathbb{L}_{\cdot}^V(\mathbb{L}_{\cdot}(\mathbb{J}))$ let $\mathbb{P}_T \in \mathcal{W}(\mathcal{Z})$ be a probability measure on a measurable space $(\mathcal{Z}, \mathcal{Z})$. Similar to an Empirical mean function §01101.10 consider a stochastic process $\psi_{\cdot,\cdot} = (\psi_{j|j_0})_{j,j_0 \in \mathcal{J}} \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J}^2)$ which in addition for all $T \in \mathbb{T}$ with $T_{\cdot,\cdot} = VTU^* \in \mathbb{L}_{\cdot}(\mathbb{J})$ and kernel $T_{\cdot,\cdot} \in \mathcal{M}(\mathcal{J}^2)$ satisfies $\psi_{j|j_0} \in \mathcal{L}_1(\mathbb{P}_T) := \mathcal{L}_1(\mathcal{Z}, \mathcal{Z}, \mathbb{P}_T)$ for each $j, j_0 \in \mathcal{J}$ and $\mathbb{P}_T(\psi_{\cdot,\cdot}) = (T_{j|j_0} = \mathbb{P}_T(\psi_{j|j_0}))_{j,j_0 \in \mathcal{J}} = T_{\cdot,\cdot}$. Considering a statistical product experiment $(\mathcal{Z}^k, \mathcal{Z}^{\otimes k}, \mathbb{P}_T^{\otimes k} = (\mathbb{P}_T^{\otimes k})_{T \in \mathbb{T}})$ similar to an Empirical mean function §01101.10 we define $\hat{T}_{\cdot,\cdot} = (\hat{T}_{j|j_0} := \hat{\mathbb{P}}_k(\psi_{j|j_0}))_{j,j_0 \in \mathcal{J}} = \hat{\mathbb{P}}_k(\psi_{\cdot,\cdot}) \in \mathcal{M}(\mathcal{Z}^{\otimes k} \otimes \mathcal{J}^2)$. For $T \in \mathbb{T}$ assuming a $\mathbb{P}_T^{\otimes k}$ -sample the mean kernel of $\hat{T}_{\cdot,\cdot}$ is by construction $\mathbb{P}_T(\psi_{\cdot,\cdot}) = T_{\cdot,\cdot} = VTU^* \in \mathbb{L}_{\cdot}(\mathbb{J})$. Moreover for each $j, j_0 \in \mathcal{J}$ the statistic $\dot{\eta}_{j|j_0} := k^{1/2}(\hat{\mathbb{P}}_k(\psi_{j|j_0}) - \mathbb{P}_T(\psi_{j|j_0})) \in \mathcal{M}(\mathcal{Z}^{\otimes k})$ is centred, i.e. $\dot{\eta}_{j|j_0} \in \mathbb{L}_1(\mathbb{P}_T^{\otimes k}) = \mathbb{L}_1(\mathcal{Z}^k, \mathcal{Z}^{\otimes k}, \mathbb{P}_T^{\otimes k})$ with $\mathbb{P}_T^{\otimes k}(\dot{\eta}_{j|j_0}) = 0$, and exploiting $\psi_{\cdot,\cdot} \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J}^2)$ the stochastic process

$$\dot{\eta}_{\cdot,\cdot} = (\dot{\eta}_{j|j_0})_{j,j_0 \in \mathcal{J}} = k^{1/2}(\hat{\mathbb{P}}_k - \mathbb{P}_T)(\psi_{\cdot,\cdot}) = k^{1/2}(\hat{\mathbb{P}}_k(\psi_{\cdot,\cdot}) - \mathbb{P}_T(\psi_{\cdot,\cdot})) \in \mathcal{M}(\mathcal{Z}^{\otimes k} \otimes \mathcal{J}^2)$$

satisfies Assumption §01101.04. Since $\hat{T}_{j|j_0} = T_{j|j_0} + k^{-1/2}\dot{\eta}_{j|j_0}$ for each $j, j_0 \in \mathcal{J}$ the stochastic process $\hat{T}_{\cdot,\cdot} = T_{\cdot,\cdot} + k^{-1/2}\dot{\eta}_{\cdot,\cdot}$ is a noisy version of the operator $T_{\cdot,\cdot} = VTU^* \in \mathbb{L}_{\cdot}(\mathbb{J})$. \square

§02101.05 **Bivariate sequence model.** Consider the measure space $(\mathcal{J}, \mathcal{J}, \nu) = (\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ as in §01101.14. Let $\dot{\eta}_{\cdot,\cdot} = (\dot{\eta}_{j|j_0})_{j,j_0 \in \mathbb{N}}$ be a real-valued stochastic process satisfying Assumption §01101.04 with mean zero and let $k \in \mathbb{N}$ be a sample size. The observable noisy version $\hat{T}_{\cdot,\cdot} = T_{\cdot,\cdot} + k^{-1/2}\dot{\eta}_{\cdot,\cdot} \sim \mathbb{P}_T^k$ with mean kernel $T_{\cdot,\cdot} \in \mathcal{M}(2^{\mathbb{N}}) = \mathbb{R}^{\mathbb{N}^2}$ takes the form of a *bivariate sequence model*

$$\hat{T}_{j|j_0} = T_{j|j_0} + k^{-1/2}\dot{\eta}_{j|j_0}, \quad j, j_0 \in \mathbb{N}. \quad (02.01)$$

If $\dot{\eta}_{\cdot,\cdot}$ admits a covariance function (possibly depending on $T_{\cdot,\cdot}$), say $\text{cov} \in \mathcal{M}(2^{\mathbb{N}^4})$, then we eventually write $\hat{T}_{\cdot,\cdot} \sim P_{(T_{\cdot,\cdot}, k^{-1} \text{cov})}$ for short. \square

§02I01.06 **Gaussian bivariate sequence model.** Let $\dot{W}_{\bullet,\bullet} := (\dot{W}_{j|j_0})_{j,j_0 \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}^2}$ be a Gaussian white noise process. The observable noisy version $\hat{T}_{\bullet,\bullet} = T_{\bullet,\bullet} + k^{-1/2} \dot{W}_{\bullet,\bullet}$ with mean kernel $T_{\bullet,\bullet} \in \mathcal{M}(2^{\mathbb{N}^2})$ takes the form of a *Gaussian bivariate sequence model*

$$\hat{T}_{j|j_0} = T_{j|j_0} + k^{-1/2} \dot{W}_{j|j_0}, \quad j, j_0 \in \mathbb{N} \quad \text{with} \quad (\dot{W}_{j|j_0})_{j,j_0 \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}^2} \quad (02.02)$$

and we denote by N_T^k the distribution of the stochastic process $\hat{T}_{\bullet,\bullet}$. \square

§02|01|01 Examples of empirical mean models

§02I01.07 **Conditional expectation operator.** Consider the Borel-measurable spaces $(\mathcal{X}, \mathcal{B}_x)$ and $(\mathcal{Z}, \mathcal{B}_z)$ for $\mathcal{X}, \mathcal{Z} \in \mathcal{B}$. Let (Z, X) be a $\mathcal{Z} \times \mathcal{X}$ -valued random vector. We denote by $\mathbb{P}^Z \in \mathcal{W}(\mathcal{B}_z)$ and $\mathbb{P}^X \in \mathcal{W}(\mathcal{B}_x)$ the marginal distribution of Z and X , respectively, by $\mathbb{P}^{X|Z}$ a regular conditional distribution of X given Z , and by $\mathbb{P}^{Z,X} = \mathbb{P}^Z \odot \mathbb{P}^{X|Z} \in \mathcal{W}(\mathcal{B}_z \otimes \mathcal{B}_x)$ the joint distribution of (Z, X) . We tactically identify Z and X with the coordinate map Π_z and Π_x , respectively, and thus (Z, X) with the identity $\text{id}_{\mathcal{Z} \times \mathcal{X}}$ such that $\mathbb{P} = \mathbb{P}^{Z,X} \in \mathcal{W}(\mathcal{B}_z \otimes \mathcal{B}_x)$. Introduce further the Hilbert spaces $\mathbb{L}_2(\mathbb{P}^X) := \mathbb{L}_2(\mathcal{X}, \mathcal{B}_x, \mathbb{P}^X) =: \mathbb{H}$, $\mathbb{L}_2(\mathbb{P}^Z) := \mathbb{L}_2(\mathcal{Z}, \mathcal{B}_z, \mathbb{P}^Z) =: \mathbb{G}$ and $\mathbb{L}_2(\mathbb{P}^{Z,X}) := \mathbb{L}_2(\mathcal{Z} \times \mathcal{X}, \mathcal{B}_z \otimes \mathcal{B}_x, \mathbb{P}^{Z,X})$. For each $h \in \mathbb{H} = \mathbb{L}_2(\mathbb{P}^X)$, and hence $h(X) \in \mathbb{L}_2(\mathbb{P}^{Z,X})$ we have $\mathbb{P}^{X|Z} h := \mathbb{P}^{X|Z}(h) = \mathbb{P}(h(X)|Z) \in \mathbb{L}_2(\mathbb{P}^Z) = \mathbb{G}$. We call $\mathbb{P}^{X|Z} : \mathbb{H} \rightarrow \mathbb{G}$ with $h \mapsto \mathbb{P}^{X|Z} h$ *conditional expectation operator*. Since by exploiting Jensens inequality for each $h \in \mathbb{H} = \mathbb{L}_2(\mathbb{P}^X)$ we have

$$\|\mathbb{P}^{X|Z} h\|_{\mathbb{G}}^2 = \mathbb{P}^Z(|\mathbb{P}^{X|Z}(h)|^2) = \mathbb{P}^Z(|\mathbb{P}(h(X)|Z)|^2) \leq \mathbb{P}^Z(\mathbb{P}(h^2(X)|Z)) = \mathbb{P}^X(h^2) = \|h\|_{\mathbb{H}}^2$$

it follows $\mathbb{P}^{X|Z} \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ with $\|\mathbb{P}^{X|Z}\|_{\mathbb{L}(\mathbb{H}, \mathbb{G})} \leq 1$. Its adjoint $(\mathbb{P}^{X|Z})^* \in \mathbb{L}(\mathbb{G}, \mathbb{H})$ satisfies $(\mathbb{P}^{X|Z})^* = \mathbb{P}^{Z|X}$. Moreover, for each $h \in \mathbb{H} = \mathbb{L}_2(\mathbb{P}^X)$ and $g \in \mathbb{G} = \mathbb{L}_2(\mathbb{P}^Z)$, hence $h(X), g(Z) \in \mathbb{L}_2(\mathbb{P}^{Z,X})$, we have

$$\langle \mathbb{P}^{X|Z} h, g \rangle_{\mathbb{G}} = \mathbb{P}^{Z,X}(g(Z)\mathbb{P}(h(X)|Z)) = \mathbb{P}^{Z,X}(g(Z)h(X)) = \langle h, \mathbb{P}^{Z|X} g \rangle_{\mathbb{H}}.$$

Evidently, the conditional expectation operator $\mathbb{P}^{X|Z}$ determines fully (and vice versa) the regular conditional distribution $\mathbb{P}^{X|Z}$ of X given Z . However, in general the marginal distributions \mathbb{P}^X and \mathbb{P}^Z , and hence the Hilbert spaces $\mathbb{H} = \mathbb{L}_2(\mathbb{P}^X)$ and $\mathbb{G} = \mathbb{L}_2(\mathbb{P}^Z)$ are not known in advance. We assume in what follows that $\mathcal{X} = \mathcal{Z} = [0, 1]$ and that X and Z is uniformly distributed on the interval $[0, 1]$, i.e. $X \sim U_{[0,1]} = \lambda_{[0,1]} = \mathbb{P}^X$ and $Z \sim U_{[0,1]} = \lambda_{[0,1]} = \mathbb{P}^Z$. We denote by $U_{\mathbb{P}^{X|Z}} := U_{[0,1]} \odot \mathbb{P}^{X|Z}$ the joint distribution of (Z, X) which is now fully specified once the conditional expectation operator $\mathbb{P}^{X|Z} \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$ is known. We consider the statistical product experiment $([0, 1]^{2k}, \mathcal{B}_{[0,1]}^{\otimes 2k}, U_{\mathbb{T}}^{\otimes k} := (U_{\mathbb{P}^{X|Z}}^{\otimes k})_{\mathbb{P}^{X|Z} \in \mathbb{T}})$ of size $k \in \mathbb{N}$ and for $\mathbb{P}^{X|Z} \in \mathbb{T}$ we denote by $((Z_i, X_i))_{i \in \llbracket k \rrbracket} \sim U_{\mathbb{P}^{X|Z}}^{\otimes k}$ an iid. sample of $(Z, X) \sim U_{\mathbb{P}^{X|Z}} = U_{[0,1]} \odot \mathbb{P}^{X|Z}$. Let $U, V \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2)$ be generalised Fourier series transforms as in **Notation** §01I02.07 which are fixed and *known* in advanced. Then $\mathbb{P}_{\bullet,\bullet}^{X|Z} := \mathbb{V}\mathbb{P}^{X|Z} U^* \in \mathbb{L}(\ell_2)$ is an operator with kernel (infinite matrix) $\mathbb{P}_{j|j_0}^{X|Z} \in \mathcal{M}(2^{\mathbb{N}^2})$ satisfying $\mathbb{P}_{j|j_0}^{X|Z} = \langle \mathbb{P}^{X|Z} u_j, v_{j_0} \rangle_{\mathbb{G}} = U_{\mathbb{P}^{X|Z}}(u_j(X)v_{j_0}(Z))_{j,j_0 \in \mathbb{N}}$. Therefore the stochastic process $\psi_{\bullet,\bullet} = (\psi_{j|j_0}(Z, X) := u_j(X)v_{j_0}(Z))_{j,j_0 \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes 2} \otimes 2^{\mathbb{N}^2})$ fulfils Assumption §02I01.02 and $\mathbb{P}_{\bullet,\bullet}^{X|Z} = U_{\mathbb{P}^{X|Z}}(\psi_{\bullet,\bullet})$. Similar to an Empirical mean model §02I01.04 we define $\hat{\mathbb{P}}_{\bullet,\bullet}^{X|Z} = (\hat{\mathbb{P}}_{j|j_0}^{X|Z} := \hat{\mathbb{P}}_n(\psi_{j|j_0}))_{j,j_0 \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes 2k} \otimes 2^{\mathbb{N}^2})$. By construction $\mathbb{P}_{\bullet,\bullet}^{X|Z} = U_{\mathbb{P}^{X|Z}}(\psi_{\bullet,\bullet}) \in \mathcal{M}(2^{\mathbb{N}^2})$ is the mean kernel of $\hat{\mathbb{P}}_{\bullet,\bullet}^{X|Z}$. For each $j, j_0 \in \mathbb{N}$ the statistic $\hat{\eta}_{j|j_0} := k^{1/2}(\hat{\mathbb{P}}_k(\psi_{j|j_0}) - U_{\mathbb{P}^{X|Z}}(\psi_{j|j_0})) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes 2k})$ is centred, i.e. $\hat{\eta}_{j|j_0} \in \mathbb{L}_1(U_{\mathbb{P}^{X|Z}}^{\otimes k})$ with $U_{\mathbb{P}^{X|Z}}^{\otimes k}(\hat{\eta}_{j|j_0}) = 0$, and exploiting $\psi_{\bullet,\bullet} \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes 2} \otimes 2^{\mathbb{N}^2})$ the stochastic process

$$\hat{\eta}_{\bullet,\bullet} = (\hat{\eta}_{j|j_0})_{j,j_0 \in \mathbb{N}} = k^{1/2}(\hat{\mathbb{P}}_k - U_{\mathbb{P}^{X|Z}})(\psi_{\bullet,\bullet}) = k^{1/2}(\hat{\mathbb{P}}_k(\psi_{\bullet,\bullet}) - U_{\mathbb{P}^{X|Z}}(\psi_{\bullet,\bullet})) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes 2k} \otimes 2^{\mathbb{N}^2})$$

satisfies Assumption §02|01.02. Since $\widehat{\mathbb{P}}_{j|j_0}^{X|Z} = \mathbb{P}_{j|j_0}^{X|Z} + k^{-1/2}\dot{\eta}_{j|j_0}$ for each $j, j_0 \in \mathbb{N}$ by construction $\widehat{\mathbb{P}}_{\cdot|\cdot}^{X|Z} = \mathbb{P}_{\cdot|\cdot}^{X|Z} + k^{-1/2}\dot{\eta}_{\cdot|\cdot}$ is a noisy version of $\mathbb{P}_{\cdot|\cdot}^{X|Z}$. \square

§02|01.08 **Covariance operator.** Let $(\mathbb{H}, \langle \cdot, \cdot \rangle_{\mathbb{H}})$ be a separable Hilbert space equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{H}}$ and X be an \mathbb{H} -valued *random function*. We tactically identify X with the identity $\text{id}_{\mathbb{H}}$ on \mathbb{H} such that X is defined on the measure space $(\mathbb{H}, \mathcal{B}_{\mathbb{H}}, \mathbb{P})$ and $X \sim \mathbb{P} = \mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{\mathbb{H}})$. Here and subsequently, we assume that $\|X\|_{\mathbb{H}}^2 \in \mathcal{L}_2(\mathbb{P})$ and $\mathbb{P}(\langle x, X \rangle_{\mathbb{H}}) = 0$ for all $x \in \mathbb{H}$. In this situation X admits a *covariance operator* $\Gamma^X \in \mathbb{L}(\mathbb{H})$ (see **Remark** §01|01.07). Let us denote by $\mathbb{P}_{\Gamma^X} \in \mathcal{W}(\mathcal{B}_{\mathbb{H}})$ the distribution of X which is not fully specified given $\Gamma^X \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{H})$. We consider the statistical product experiment $(\mathbb{H}^k, \mathcal{B}_{\mathbb{H}}^{\otimes k}, \mathbb{P}_{\Gamma^X}^{\otimes k} = (\mathbb{P}_{\Gamma^X}^{\otimes k})_{\Gamma^X \in \mathbb{T}})$. Let $(u_j)_{j \in \mathbb{N}}$ be an *orthonormal system* in \mathbb{H} and denote by $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ its associated generalised Fourier series transform (see **Notation** §01|02.07). Then $\Gamma_{\cdot|\cdot}^X = U\Gamma^X U^* \in \mathbb{L}(\ell_2)$ is a positive semi-definite operator with kernel (infinite matrix) $\Gamma_{j|j_0}^X \in \mathcal{M}(2^{\mathbb{N}})$ which satisfies $\Gamma_{j|j_0}^X = (\Gamma_{j|j_0}^X = \mathbb{P}_{\Gamma^X}(\langle u_j, X \rangle_{\mathbb{H}} \langle X, u_{j_0} \rangle_{\mathbb{H}}))_{j, j_0 \in \mathbb{N}}$. The process $\psi_{\cdot|\cdot} = (\psi_{j|j_0}(X) := \langle u_j, X \rangle_{\mathbb{H}} \langle X, u_{j_0} \rangle_{\mathbb{H}})_{j, j_0 \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_{\mathbb{H}} \otimes 2^{\mathbb{N}})$ fulfils Assumption §02|01.02 and $\Gamma_{\cdot|\cdot}^X = \mathbb{P}_{\Gamma^X}(\psi_{\cdot|\cdot})$. Similar to an Empirical mean model §02|01.04 we define $\widehat{\Gamma}_{\cdot|\cdot}^X = (\widehat{\Gamma}_{j|j_0}^X := \widehat{\mathbb{P}}_k(\psi_{j|j_0}))_{j, j_0 \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_{\mathbb{H}}^{\otimes k} \otimes 2^{\mathbb{N}})$. By construction $\Gamma_{\cdot|\cdot}^X = \mathbb{P}_{\Gamma^X}(\psi_{\cdot|\cdot}) \in \mathcal{M}(2^{\mathbb{N}})$ is the mean kernel of $\widehat{\Gamma}_{\cdot|\cdot}^X$. For each $j, j_0 \in \mathbb{N}$ the statistic $\dot{\eta}_{j|j_0} := n^{1/2}(\widehat{\mathbb{P}}_k(\psi_{j|j_0}) - \mathbb{P}_{\Gamma^X}(\psi_{j|j_0})) \in \mathcal{M}(\mathcal{B}_{\mathbb{H}}^{\otimes k})$ is centred, i.e. $\dot{\eta}_{j|j_0} \in \mathbb{L}_1(\mathbb{P}_{\Gamma^X}^{\otimes k})$ with $\mathbb{P}_{\Gamma^X}^{\otimes k}(\dot{\eta}_{j|j_0}) = 0$, and exploiting $\psi_{\cdot|\cdot} \in \mathcal{M}(\mathcal{B}_{\mathbb{H}} \otimes 2^{\mathbb{N}})$ the stochastic process

$$\dot{\eta}_{\cdot|\cdot} = (\dot{\eta}_{j|j_0})_{j, j_0 \in \mathbb{N}} = k^{1/2}(\widehat{\mathbb{P}}_k - \mathbb{P}_{\Gamma^X})(\psi_{\cdot|\cdot}) = k^{1/2}(\widehat{\mathbb{P}}_k(\psi_{\cdot|\cdot}) - \mathbb{P}_{\Gamma^X}(\psi_{\cdot|\cdot})) \in \mathcal{M}(\mathcal{B}_{\mathbb{H}}^{\otimes k} \otimes 2^{\mathbb{N}})$$

satisfies Assumption §02|01.02. Since $\widehat{\Gamma}_{j|j_0}^X = \Gamma_{j|j_0}^X + k^{-1/2}\dot{\eta}_{j|j_0}$ for each $j, j_0 \in \mathbb{N}$ by construction $\widehat{\Gamma}_{\cdot|\cdot}^X = \Gamma_{\cdot|\cdot}^X + k^{-1/2}\dot{\eta}_{\cdot|\cdot}$ is a noisy version of $\Gamma_{\cdot|\cdot}^X$. \square

§02|01.09 **Cross-covariance operator.** Let $(\mathbb{H}, \langle \cdot, \cdot \rangle_{\mathbb{H}})$ and $(\mathbb{G}, \langle \cdot, \cdot \rangle_{\mathbb{G}})$ be separable Hilbert space equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{H}}$ and $\mathcal{B}_{\mathbb{G}}$, respectively. Consider an \mathbb{H} -valued *random function* X and an \mathbb{G} -valued *random function* Z . Then (Z, X) is an $(\mathbb{G} \times \mathbb{H}, \mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}})$ -valued *random function*. We denote by $\mathbb{P}^Z \in \mathcal{W}(\mathcal{B}_{\mathbb{G}})$ and $\mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{\mathbb{H}})$ the marginal distribution of Z and X , respectively, and by $\mathbb{P}^{Z, X} \in \mathcal{W}(\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}})$ the joint distribution of (Z, X) . We tactically identify Z and X with the coordinate map $\Pi_{\mathbb{G}}$ and $\Pi_{\mathbb{H}}$, respectively, and thus (Z, X) with the identity $\text{id}_{\mathbb{G} \times \mathbb{H}}$ such that $\mathbb{P} = \mathbb{P}^{Z, X} \in \mathcal{W}(\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}})$. Here and subsequently, we assume that $\|Z\|_{\mathbb{G}}^2 \in \mathcal{L}_2(\mathbb{P})$, $\|X\|_{\mathbb{H}}^2 \in \mathcal{L}_2(\mathbb{P})$, $\mathbb{P}(\langle z, Z \rangle_{\mathbb{G}}) = 0$ and $\mathbb{P}(\langle x, X \rangle_{\mathbb{H}}) = 0$ for all $z \in \mathbb{G}$ and $x \in \mathbb{H}$. In this situation Z and X admits a *covariance operator* $\Gamma^Z \in \mathbb{L}(\mathbb{G})$ and $\Gamma^X \in \mathbb{L}(\mathbb{H})$, respectively (see **Remark** §01|01.07), and (Z, X) admits a *cross-covariance operator* $\Gamma^{Z, X} \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ satisfying

$$\langle \Gamma^{Z, X} x, z \rangle_{\mathbb{G}} = \mathbb{P}^{Z, X}(\langle z, Z \rangle_{\mathbb{G}} \langle X, x \rangle_{\mathbb{H}}) \quad \forall x \in \mathbb{H}, z \in \mathbb{G}.$$

where $\|\Gamma^{Z, X}\|_{\mathbb{L}(\mathbb{H}, \mathbb{G})} \leq \|\Gamma^Z\|_{\mathbb{L}(\mathbb{G})}^{1/2} \|\Gamma^X\|_{\mathbb{L}(\mathbb{H})}^{1/2}$ (Baker [1973] p.275). Let us denote by $\mathbb{P}_{\Gamma^{Z, X}} \in \mathcal{W}(\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}})$ the distribution of (Z, X) which is not fully specified given $\Gamma^{Z, X} \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$. We consider the statistical product experiment $((\mathbb{G} \times \mathbb{H})^k, (\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}})^{\otimes k}, \mathbb{P}_{\Gamma^{Z, X}}^{\otimes k} = (\mathbb{P}_{\Gamma^{Z, X}}^{\otimes k})_{\Gamma^{Z, X} \in \mathbb{T}})$. Let $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ and $V \in \mathbb{L}(\mathbb{G}, \ell_2)$ be generalised Fourier series transforms as in **Notation** §01|02.07 which are fixed and known in advanced. Then $\Gamma_{\cdot|\cdot}^{Z, X} := V\Gamma^{Z, X}U^* \in \mathbb{L}(\ell_2)$ is an operator with kernel (infinite matrix) $\Gamma_{j|j_0}^{Z, X} \in \mathcal{M}(2^{\mathbb{N}})$ satisfying $\Gamma_{j|j_0}^{Z, X} = (\Gamma_{j|j_0}^{Z, X} = \langle \Gamma^{Z, X} u_j, v_j \rangle_{\mathbb{G}} = \mathbb{P}_{\Gamma^{Z, X}}(\langle v_j, Z \rangle_{\mathbb{G}} \langle X, u_j \rangle_{\mathbb{H}}))_{j, j_0 \in \mathbb{N}}$. Therefore the stochastic process $\psi_{\cdot|\cdot} = (\psi_{j|j_0}(Z, X) := \langle v_j, Z \rangle_{\mathbb{G}} \langle X, u_j \rangle_{\mathbb{H}})_{j, j_0 \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}}) \otimes 2^{\mathbb{N}})$ fulfils Assumption §02|01.02 and $\Gamma_{\cdot|\cdot}^{Z, X} = \mathbb{P}_{\Gamma^{Z, X}}(\psi_{\cdot|\cdot})$. Similar to an Empirical mean model §02|01.04 we define $\widehat{\Gamma}_{\cdot|\cdot}^{Z, X} = (\widehat{\Gamma}_{j|j_0}^{Z, X} := \widehat{\mathbb{P}}_n(\psi_{j|j_0}))_{j, j_0 \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}})^{\otimes k} \otimes 2^{\mathbb{N}})$. By construction $\Gamma_{\cdot|\cdot}^{Z, X} = \mathbb{P}_{\Gamma^{Z, X}}(\psi_{\cdot|\cdot}) \in$

$\mathcal{M}(2^{\mathbb{N}})$ is the mean kernel of $\widehat{\Gamma}_{\cdot, \cdot}^{Z^X}$. For each $j, j_o \in \mathbb{N}$ the statistic $\dot{\eta}_{j|j_o} := k^{1/2}(\widehat{\mathbb{P}}_k(\psi_{j|j_o}) - \mathbb{P}_{\Gamma^{zx}}(\psi_{j|j_o})) \in \mathcal{M}((\mathcal{B}_e \otimes \mathcal{B}_h)^{\otimes k})$ is centred, i.e. $\dot{\eta}_{j|j_o} \in \mathbb{L}_1(\mathbb{P}_{\Gamma^{zx}} \otimes k)$ with $\mathbb{P}_{\Gamma^{zx}}^{\otimes k}(\dot{\eta}_{j|j_o}) = 0$, and the stochastic process

$$\dot{\eta}_{\cdot, \cdot} = (\dot{\eta}_{j|j_o})_{j, j_o \in \mathbb{N}} = k^{1/2}(\widehat{\mathbb{P}}_k - \mathbb{P}_{\Gamma^{zx}})(\psi_{\cdot, \cdot}) = k^{1/2}(\widehat{\mathbb{P}}_k(\psi_{\cdot, \cdot}) - \mathbb{P}_{\Gamma^{zx}}(\psi_{\cdot, \cdot})) \in \mathcal{M}((\mathcal{B}_e \otimes \mathcal{B}_h)^{\otimes k} \otimes 2^{\mathbb{N}})$$

satisfies Assumption §02|01.02 exploiting $\psi_{\cdot, \cdot} \in \mathcal{M}((\mathcal{B}_e \otimes \mathcal{B}_h)^{\otimes k} \otimes 2^{\mathbb{N}})$. Since $\widehat{\Gamma}_{j|j_o}^{Z^X} = \Gamma_{j|j_o}^{Z^X} + k^{-1/2}\dot{\eta}_{j|j_o}$ for each $j, j_o \in \mathbb{N}$ by construction $\widehat{\Gamma}_{\cdot, \cdot}^{Z^X} = \Gamma_{\cdot, \cdot}^{Z^X} + k^{-1/2}\dot{\eta}_{\cdot, \cdot}$ is a noisy version of $\Gamma_{\cdot, \cdot}^{Z^X}$. \square

§02|01.10 **Design operator.** Consider the measure space $(\mathcal{D}, \mathcal{B}_D, \lambda_D)$ where λ_D denotes the restriction of the Lebesgue measure to the Borel- σ -algebra \mathcal{B}_D over $\mathcal{D} \in \mathcal{B}$, and the *real* Hilbert space $\mathbb{L}_2(\lambda_D) := \mathbb{L}_2(\mathcal{D}, \mathcal{B}_D, \lambda_D)$ of square Lebesgue-integrable real-valued functions. Let $\mathbb{P}_\varphi \in \mathcal{W}(\mathcal{B}_D)$ admit a Lebesgue density $\varphi \in \mathbb{L}_1(\lambda_D)$, that is $\mathbb{P}_\varphi = \varphi \lambda_D$ (compare Regression with known design §01|05.10). Let in addition $\varphi \in \mathbb{L}_\infty(\lambda_D)$, then $M_\varphi \in \mathbb{L}(\mathbb{L}_2(\lambda_D))$ with $h \mapsto M_\varphi h := \varphi h$. Note that then for each $h \in \mathbb{L}_2(\lambda_D)$ we have $M_\varphi h \in \mathbb{L}_2(\lambda_D)$. Consequently, for each $g, h \in \mathbb{L}_2(\lambda_D)$, hence $g, h \in \mathcal{L}_2(\mathbb{P}_\varphi)$ we obtain $gh \in \mathcal{L}_1(\mathbb{P}_\varphi)$ and

$$\mathbb{P}_\varphi(gh) = \varphi \lambda_D(gh) = \lambda_D(\varphi gh) = \langle M_\varphi g, h \rangle_{\mathbb{L}_2(\lambda_D)} \in \mathbb{R}$$

identifying again equivalence classes and their representatives. We note that $M_\varphi \in \mathbb{L}(\mathbb{L}_2(\lambda_D))$ with density $\varphi \in \mathbb{L}_\infty(\lambda_D)$ is positive semi-definite, i.e. $M_\varphi \in \mathbb{L}^{\geq}(\mathbb{L}_2(\lambda_D))$ and if in addition $\varphi \in \mathcal{M}_{>0, \lambda_D}(\mathcal{B}_D)$ (i.e. $\varphi \in \mathcal{M}_{\geq 0}(\mathcal{B}_D)$ and $\lambda_D(\mathcal{N}_\varphi) = 0$) then it is strictly positive definite, i.e. $M_\varphi \in \mathbb{L}^{\gg}(\mathbb{L}_2(\lambda_D))$. Keep in mind that $U \in \mathbb{L}(\mathbb{L}_2(\lambda_D), \ell_2)$ is generalised Fourier series transform as in Notation §01|02.07 which is fixed and *known in advance*. Evidently, we have $M_{\cdot, \cdot}^\varphi := UM_\varphi U^* \in \mathbb{L}^{\geq}(\ell_2) \subseteq \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$ satisfying $M_{\cdot, \cdot}^\varphi = (M_{j|j_o}^\varphi = \langle M_{\cdot, \cdot}^\varphi u_j, u_{j_o} \rangle_{\mathbb{G}} = \mathbb{P}_\varphi(u_j u_{j_o}))_{j, j_o \in \mathbb{N}}$. Therefore the stochastic process $\psi_{\cdot, \cdot} = (\psi_{j|j_o} := u_j u_{j_o})_{j, j_o \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_D \otimes 2^{\mathbb{N}})$ fulfils Assumption §02|01.02 and $M_{\cdot, \cdot}^\varphi = \mathbb{P}_\varphi(\psi_{\cdot, \cdot})$. Similar to an Empirical mean model §02|01.04 we define $\widehat{M}_{\cdot, \cdot} = (\widehat{M}_{j|j_o} := \widehat{\mathbb{P}}_k(\psi_{j|j_o}))_{j, j_o \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_D^{\otimes k} \otimes 2^{\mathbb{N}})$. By construction $M_{\cdot, \cdot}^\varphi = \mathbb{P}_\varphi(\psi_{\cdot, \cdot}) \in \mathcal{M}(2^{\mathbb{N}})$ is the mean kernel of $\widehat{M}_{\cdot, \cdot}$. For each $j, j_o \in \mathbb{N}$ the statistic $\dot{\eta}_{j|j_o} := k^{1/2}(\widehat{\mathbb{P}}_k(\psi_{j|j_o}) - \mathbb{P}_\varphi(\psi_{j|j_o})) \in \mathcal{M}(\mathcal{B}_D^{\otimes k})$ is centred, i.e. $\dot{\eta}_{j|j_o} \in \mathbb{L}_1(\mathbb{P}_\varphi^{\otimes k})$ with $\mathbb{P}_\varphi^{\otimes k}(\dot{\eta}_{j|j_o}) = 0$, and exploiting $\psi_{\cdot, \cdot} \in \mathcal{M}(\mathcal{B}_D \otimes 2^{\mathbb{N}})$ the stochastic process

$$\dot{\eta}_{\cdot, \cdot} = (\dot{\eta}_{j|j_o})_{j, j_o \in \mathbb{N}} = k^{1/2}(\widehat{\mathbb{P}}_k - \mathbb{P}_\varphi)(\psi_{\cdot, \cdot}) = k^{1/2}(\widehat{\mathbb{P}}_k(\psi_{\cdot, \cdot}) - \mathbb{P}_\varphi(\psi_{\cdot, \cdot})) \in \mathcal{M}(\mathcal{B}_D^{\otimes k} \otimes 2^{\mathbb{N}})$$

satisfies Assumption §02|01.02. Since $\widehat{M}_{j|j_o} = M_{j|j_o}^\varphi + k^{-1/2}\dot{\eta}_{j|j_o}$ for each $j, j_o \in \mathbb{N}$ by construction $\widehat{M}_{\cdot, \cdot} = M_{\cdot, \cdot}^\varphi + k^{-1/2}\dot{\eta}_{\cdot, \cdot}$ is a noisy version of $M_{\cdot, \cdot}^\varphi$. \square

§02|02 Non-diagonal statistical inverse problem with noisy operator

§02|02.01 **Assumption.** For $\mathbb{J} = \mathbb{L}_2(\nu)$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$, fixed and presumed to be known in advance, the operator $T \in \mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J}))$ and hence $T_{\cdot, \cdot} = VTU^* \in \mathbb{L}(\mathbb{J})$ with kernel $T_{\cdot, \cdot} \in \mathcal{J}^2$ is *not known* in advance where $g = T_{\cdot, \cdot}\theta \in \mathbb{J}$ or inequal $g \in \text{ran}(T_{\cdot, \cdot}) = \{T_{\cdot, \cdot}a : a \in \mathbb{J}\}$. \square

§02|02.02 **Notation.** Under Assumption §02|02.01 given $T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J})$ and $g \in \text{ran}(T_{\cdot, \cdot})$ we consider the reconstruction of $\theta = U\theta \in \mathbb{J}$ (or in equal $\theta = U^*\theta \in \mathbb{H}$) from a noisy version of the image $g = VTU^*\theta = T_{\cdot, \cdot}\theta \in \mathbb{J}$ and a noisy version of the operator $T_{\cdot, \cdot} \in \mathbb{L}(\mathbb{J})$. Keep in mind, that we identify the equivalence class and its representative g . \square

§02|02.03 **Non-diagonal statistical inverse problem with noisy operator.** As in Definition §01|02.03 consider a stochastic process $\dot{\epsilon}_{\cdot} = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ satisfying Assumption §01|01.04 with mean zero and a sample size $n \in \mathbb{N}$, and in addition as in Definition §02|01.03 a stochastic process $\dot{\eta}_{\cdot, \cdot} =$

$(\dot{\eta}_{j\iota})_{j,j_\circ \in \mathcal{J}}$ satisfying Assumption §02101.02 with mean zero and a sample size $k \in \mathbb{N}$. Under Assumption §02102.01 where $\mathbb{T}_{j\iota} \in \mathbb{L}(\mathbb{J})$ with kernel $\mathbb{T}_{j\iota} \in \mathcal{M}(\mathcal{J}^2)$ is *not known* anymore, the observable noisy image (Definition §01102.03) has \mathbb{J} -mean $\underline{g} = \mathbb{T}_{j\iota} \theta$ and the observable noisy non-diagonal operator (Definition §02101.03) has mean kernel $\mathbb{T}_{j\iota} \in \mathcal{M}(\mathcal{J}^2)$, and take the form $\widehat{\underline{g}} = \underline{g} + n^{-1/2} \dot{\underline{\epsilon}}$ and $\widehat{\mathbb{T}}_{j\iota} = \mathbb{T}_{j\iota} + k^{-1/2} \dot{\eta}_{j\iota}$, respectively, or in equal

$$\widehat{\underline{g}}_j = \langle \mathbb{T}_{j\iota}, \theta \rangle_{\mathbb{J}} + n^{-1/2} \dot{\epsilon}_j \quad \text{and} \quad \widehat{\mathbb{T}}_{j\iota} = \mathbb{T}_{j\iota} + k^{-1/2} \dot{\eta}_{j\iota}, \quad j, j_\circ \in \mathcal{J}. \quad (02.03)$$

We denote by $\mathbb{P}_{\theta|\mathbb{T}}^{n,k}$ the joint distribution of $(\widehat{\underline{g}}, \widehat{\mathbb{T}}_{j\iota})$. The reconstruction of $\theta \in \mathbb{J}$ (or in equal $\theta = \mathbb{U}^* \theta \in \mathbb{H}$) from a noisy version $(\widehat{\underline{g}}, \widehat{\mathbb{T}}_{j\iota}) \sim \mathbb{P}_{\theta|\mathbb{T}}^{n,k}$ of the image $\underline{g} = \mathbb{V} \mathbb{T} \mathbb{U}^* \theta = \mathbb{T}_{j\iota} \theta \in \mathbb{J}$ and the operator $\mathbb{T} \in \mathbb{L}^{\mathbb{U},\mathbb{V}}(\mathbb{L}(\mathbb{J}))$ is called a *non-diagonal statistical inverse problem with noisy operator*. \square

§02102.04 **Non-diagonal inverse empirical mean model (nieMM) with noisy operator.** Consider the reconstruction of $\theta \in \mathbb{J}$ (in equal $\theta = \mathbb{U}^* \theta \in \mathbb{H}$) in an Empirical mean model as in §01102.04. Under Assumption §02102.01, where $\mathbb{T}_{j\iota} \in \mathbb{L}(\mathbb{J})$ with kernel $\mathbb{T}_{j\iota} \in \mathcal{M}(\mathcal{J}^2)$ is *not known* in advance, the observable noisy image has \mathbb{J} -mean $\mathbb{V} \underline{g} = \underline{g} = \mathbb{T}_{j\iota} \theta \in \mathbb{J}$ and the observable noisy non-diagonal operator (Definition §02101.03) has mean kernel $\mathbb{T}_{j\iota} \in \mathcal{J}^2$, and take, respectively, the form of an Empirical mean model as in §01102.04 and Empirical mean model as in §02101.04. More precisely, for each $\theta \in \Theta \subseteq \mathbb{H}$ and $\mathbb{T} \in \mathbb{T} \subseteq \mathbb{L}^{\mathbb{U},\mathbb{V}}(\mathbb{L}(\mathbb{J}))$ let $\mathbb{P}_{\theta|\mathbb{T}} \in \mathcal{W}(\mathcal{Z})$ be a probability measure on a measurable space $(\mathcal{Z}, \mathcal{Z})$. Similar to §01102.04 and §02101.04 consider stochastic processes $\psi_j^{\ominus|\mathbb{T}} \in \mathcal{Z} \otimes \mathcal{J}$ and $\psi_{j\iota}^{\mathbb{T}} \in \mathcal{Z} \otimes \mathcal{J}^2$ which in addition for all $\theta \in \Theta$ and $\mathbb{T} \in \mathbb{T}$ satisfy $\psi_j^{\ominus|\mathbb{T}}, \psi_{j\iota}^{\mathbb{T}} \in \mathcal{L}_1(\mathbb{P}_{\theta|\mathbb{T}})$ for each $j, j_\circ \in \mathcal{J}$ and $\mathbb{E}_{\theta|\mathbb{T}}(\psi_j^{\ominus|\mathbb{T}}) = \underline{g} = \mathbb{T}_{j\iota} \theta$ and $\mathbb{E}_{\theta|\mathbb{T}}(\psi_{j\iota}^{\mathbb{T}}) = \mathbb{T}_{j\iota}$. The observable noisy versions take the form $\widehat{\underline{g}} = \mathbb{T}_{j\iota} \theta + n^{-1/2} \dot{\underline{\epsilon}}$ and $\widehat{\mathbb{T}}_{j\iota} = \mathbb{T}_{j\iota} + k^{-1/2} \dot{\eta}_{j\iota}$, or in equal (02.03) with error processes $\dot{\underline{\epsilon}} = n^{1/2}(\widehat{\mathbb{P}}_n(\psi_j^{\ominus|\mathbb{T}}) - \mathbb{E}_{\theta|\mathbb{T}}(\psi_j^{\ominus|\mathbb{T}})) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J})$ and $\dot{\eta}_{j\iota} = k^{1/2}(\widehat{\mathbb{P}}_k(\psi_{j\iota}^{\mathbb{T}}) - \mathbb{E}_{\theta|\mathbb{T}}(\psi_{j\iota}^{\mathbb{T}})) \in \mathcal{M}(\mathcal{Z}^{\otimes k} \otimes \mathcal{J}^2)$ satisfying Assumption §01101.04 and Assumption §02101.02. \square

§02102.05 **Non-diagonal inverse sequence model (niSM) with noisy operator.** Consider $\mathbb{J} = \ell_2 = \mathbb{L}_2(\nu_{\mathbb{N}})$ as in §01101.14. Let $\dot{\underline{\epsilon}} = (\dot{\epsilon}_j)_{j \in \mathbb{N}}$ and $\dot{\eta}_{j\iota} = (\dot{\eta}_{j\iota})_{j,j_\circ \in \mathbb{N}}$ be real-valued stochastic processes satisfying Assumption §01101.04 and Assumption §02101.02 with mean zero and let $n, k \in \mathbb{N}$ be sample sizes. Under Assumption §02102.01, where $\mathbb{T}_{j\iota} \in \mathbb{L}(\ell_2)$ with kernel $\mathbb{T}_{j\iota} \in \mathcal{M}(2^{\mathbb{N}^2})$ is not known in advance, the observable noisy image has ℓ_2 -mean $\underline{g} = \mathbb{T}_{j\iota} \theta$ and the observable noisy operator has mean kernel $\mathbb{T}_{j\iota} \in \mathcal{M}(2^{\mathbb{N}^2})$, and take the form of a Sequence model as in §01102.05 and Bivariate sequence model as in §02101.05, that is $\widehat{\underline{g}} = \mathbb{T}_{j\iota} \theta + n^{-1/2} \dot{\underline{\epsilon}}$ and $\widehat{\mathbb{T}}_{j\iota} = \mathbb{T}_{j\iota} + k^{-1/2} \dot{\eta}_{j\iota}$, or in equal

$$\widehat{\underline{g}}_j = \langle \mathbb{T}_{j\iota}, \theta \rangle_{\ell_2} + n^{-1/2} \dot{\epsilon}_j \quad \text{and} \quad \widehat{\mathbb{T}}_{j\iota} = \mathbb{T}_{j\iota} + k^{-1/2} \dot{\eta}_{j\iota}, \quad j, j_\circ \in \mathbb{N}. \quad (02.04)$$

We denote by $\mathbb{P}_{\theta|\mathbb{T}}^{n,k}$ the joint distribution of $(\widehat{\underline{g}}, \widehat{\mathbb{T}}_{j\iota})$. \square

§02102.06 **Gaussian non-diagonal inverse sequence model (GniSM) with noisy operator.** Consider $\mathbb{J} = \ell_2$ as in §01101.14. Let $\dot{\mathbb{B}} := (\dot{\mathbb{B}}_j)_{j \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{\mathbb{W}}_{j\iota} := (\dot{\mathbb{W}}_{j\iota})_{j,j_\circ \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}^2}$ be Gaussian white noise process. The observable noisy versions $\widehat{\underline{g}} = \underline{g} + n^{-1/2} \dot{\mathbb{B}}$ with ℓ_2 -mean $\underline{g} = \mathbb{T}_{j\iota} \theta$ and $\widehat{\mathbb{T}}_{j\iota} = \mathbb{T}_{j\iota} + k^{-1/2} \dot{\mathbb{W}}_{j\iota}$ with mean kernel $\mathbb{T}_{j\iota} \in \mathcal{M}(2^{\mathbb{N}^2})$ take the form of a Gaussian sequence model as in §01102.06 and Gaussian bivariate sequence model as in §02101.06, that is

$$\widehat{\underline{g}}_j = \langle \mathbb{T}_{j\iota}, \theta \rangle_{\ell_2} + n^{-1/2} \dot{\mathbb{B}}_j \quad \text{and} \quad \widehat{\mathbb{T}}_{j\iota} = \mathbb{T}_{j\iota} + k^{-1/2} \dot{\mathbb{W}}_{j\iota}, \quad j, j_\circ \in \mathbb{N}$$

with $(\dot{\mathbb{B}}_j)_{j \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}$ and $(\dot{\mathbb{W}}_{j\iota})_{j,j_\circ \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}^2}$. (02.05)

We denote by $\mathbb{N}_{\theta|\mathbb{T}}^{n,k}$ the joint distribution of the stochastic process $(\widehat{\underline{g}}, \widehat{\mathbb{T}}_{j\iota})$. \square

§02|02|01 Examples of non-diagonal inverse empirical mean models with noisy operator

§02|02.07 **Instrumental regression.** Consider for $\mathcal{Z}, \mathcal{X} \in \mathcal{B}$ the Borel-measurable spaces $(\mathcal{Z}, \mathcal{B}_z), (\mathcal{X}, \mathcal{B}_x)$ and $(\mathbb{R}, \mathcal{B})$. Let (Z, X, Y) be a $\mathcal{Z} \times \mathcal{X} \times \mathbb{R}$ -valued random vector with joint distribution $\mathbb{P}^{Z, X, Y} \in \mathcal{W}(\mathcal{B}_z \otimes \mathcal{B}_x \otimes \mathcal{B})$. We denote by $\mathbb{P}^Z \in \mathcal{W}(\mathcal{B}_z)$ the marginal distribution of Z , by $\mathbb{P}^{X|Z}$ and $\mathbb{P}^{Y|Z}$ a regular conditional distribution of X given Z and Y given Z , respectively, and by $\mathbb{P}^{Z, X} = \mathbb{P}^Z \odot \mathbb{P}^{X|Z} \in \mathcal{W}(\mathcal{B}_z \otimes \mathcal{B}_x)$ and $\mathbb{P}^{Z, Y} = \mathbb{P}^Z \odot \mathbb{P}^{Y|Z} \in \mathcal{W}(\mathcal{B}_z \otimes \mathcal{B})$ the marginal distributions of (Z, X) and (Z, Y) . We tactically identify Z, X and Y with the coordinate map Π_z, Π_x and Π_r , respectively, and thus (Z, X, Y) with the identity $\text{id}_{\mathcal{Z} \times \mathcal{X} \times \mathbb{R}}$ such that $\mathbb{P} = \mathbb{P}^{Z, X, Y} \in \mathcal{W}(\mathcal{B}_z \otimes \mathcal{B}_x \otimes \mathcal{B})$. If in addition $Y \in \mathbb{L}_1(\mathbb{P}) = \mathbb{L}_1(\mathcal{Z} \times \mathcal{X} \times \mathbb{R}, \mathcal{B}_z \otimes \mathcal{B}_x \otimes \mathcal{B}, \mathbb{P})$ then $\mathbb{P}^{Y|Z}(\text{id}_{\mathcal{B}_y}) = \mathbb{P}(Y|Z) =: g \in \mathbb{L}_1(\mathbb{P}^Z)$ is unique up to \mathbb{P}^Z -a.s. equality (compare Regression with uniform design §01|02.09). Introduce further the Hilbert spaces $\mathbb{L}_2(\mathbb{P}^X) := \mathbb{L}_2(\mathcal{X}, \mathcal{B}_x, \mathbb{P}^X)$, $\mathbb{L}_2(\mathbb{P}^Z) := \mathbb{L}_2(\mathcal{Z}, \mathcal{B}_z, \mathbb{P}^Z)$ and as in §02|01.07 the *conditional expectation operator* $\mathbb{P}^{X|Z} \in \mathbb{L}(\mathbb{L}_2(\mathbb{P}^X), \mathbb{L}_2(\mathbb{P}^Z))$ with $h \mapsto \mathbb{P}^{X|Z} h := \mathbb{P}^{X|Z}(h) = \mathbb{P}(h(X)|Z)$. In what follows we assume that $Y \in \mathbb{L}_2(\mathbb{P})$ and hence $g \in \mathbb{L}_2(\mathbb{P}^Z)$, and that in addition $g \in \text{ran}(\mathbb{P}^{X|Z}) \subseteq \mathbb{L}_2(\mathbb{P}^Z)$. In this situation there exists $f \in \mathbb{L}_2(\mathbb{P}^X)$ such that for any $h \in \mathbb{L}_2(\mathbb{P}^Z)$

$$\langle g, h \rangle_{\mathbb{L}_2(\mathbb{P}^Z)} = \mathbb{P}^Z(\mathbb{P}(Y|Z)h(Z)) = \mathbb{P}^Z(\mathbb{P}(f(X)|Z)h(Z)) = \langle \mathbb{P}^{X|Z} f, h \rangle_{\mathbb{L}_2(\mathbb{P}^Z)}$$

or in equal \mathbb{P} -a.s. we have $Y = f(X) + \xi$ with $\mathbb{P}(\xi|Z) = 0$. We note that for all $h \in \mathbb{L}_2(\mathbb{P}^Z)$ we have $\langle g, h \rangle_{\mathbb{L}_2(\mathbb{P}^Z)} = \mathbb{P}(Yh(Z))$. We assume moreover that $\mathcal{X} = \mathcal{Z} = [0, 1]$ and that X and Z is uniformly distributed on the interval $[0, 1]$, i.e. $X \sim U_{[0,1]} = \lambda_{[0,1]} = \mathbb{P}^X$ and $Z \sim U_{[0,1]} = \lambda_{[0,1]} = \mathbb{P}^Z$. Consequently, we set $\mathbb{H} := \mathbb{L}_2(\mathbb{P}^X) = \mathbb{L}_2(\lambda_{[0,1]})$ and $\mathbb{G} := \mathbb{L}_2(\mathbb{P}^Z) = \mathbb{L}_2(\lambda_{[0,1]})$. We denote by $U_{\mathbb{P}^{X|Z}} := U_{[0,1]} \odot \mathbb{P}^{X|Z}$ the joint distribution of (Z, X) which is now fully specified once the conditional expectation operator $\mathbb{P}^{X|Z} \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}))$ is given (see Model §02|01.07). Moreover, for $\mathbb{P}^{X|Z} \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}))$ and $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]}) = \mathbb{H}$, and hence $g := \mathbb{P}^{X|Z} f \in \mathbb{L}_2(\lambda_{[0,1]}) = \mathbb{G}$, we denote by $U_{\mathbb{P}^{Y|Z}} := U_{[0,1]} \odot \mathbb{P}_g^{Y|Z}$ the joint distribution of (Z, Y) without fully specifying the regular conditional distribution $\mathbb{P}_g^{Y|Z}$ which however satisfies $\mathbb{P}_g^{Y|Z}(\text{id}_{\mathcal{B}_y}) = \mathbb{P}_g(Y|Z) = g = \mathbb{P}^{X|Z} f$ (see Model §01|05.09). Let $U, V \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2)$ be generalised Fourier series transforms as in Notation §01|02.07 which are fixed and *known* in advance. Following Model §02|01.07 $\mathbb{P}_\bullet^{X|Z} := V\mathbb{P}^{X|Z}U^* \in \mathbb{L}(\ell_2)$ is an operator with kernel (infinite matrix) $\mathbb{P}_\bullet^{X|Z} \in \mathcal{M}(2^{\mathbb{N}^2})$ satisfying $\mathbb{P}_\bullet^{X|Z} = (\mathbb{P}_{j|j_0}^{X|Z} = \langle \mathbb{P}^{X|Z} u_j, v_{j_0} \rangle_{\mathbb{G}} = U_{\mathbb{P}^{X|Z}}(u_j(X)v_{j_0}(Z)))_{j, j_0 \in \mathbb{N}}$. Therefore the stochastic process $\psi_\bullet = (\psi_{j|j_0}(Z, X) := u_j(X)v_{j_0}(Z))_{j, j_0 \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_{[0,1]}^2 \otimes 2^{\mathbb{N}^2})$ fulfils Assumption §02|01.02 and $\mathbb{P}_\bullet^{X|Z} = U_{\mathbb{P}^{X|Z}}(\psi_\bullet)$. Moreover, similar to Model §01|05.09 for each $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]}) = \mathbb{H}$ the generalised Fourier coefficients $g = (g_j)_{j \in \mathbb{N}} = Vg = V\mathbb{P}^{X|Z}U^*Uf = \mathbb{P}_\bullet^{X|Z} f_\bullet$, satisfy $g = U_{\mathbb{P}^{X|Z}}(Yv_\bullet(Z))$. The stochastic process $\psi = (\psi_j := Yv_j(X))_{j \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_{[0,1]} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}}$ fulfils Assumption §01|01.04 and $g = \mathbb{P}_\bullet^{X|Z} f_\bullet = U_{\mathbb{P}^{X|Z}}(\psi_\bullet^{\mathbb{E}_1^{\mathbb{T}}})$. The observable noisy versions take the form $\hat{g} = \mathbb{P}_\bullet^{X|Z} f_\bullet + n^{-1/2} \dot{\epsilon}_\bullet$ and $\hat{\mathbb{P}}_\bullet^{X|Z} = \mathbb{P}_\bullet^{X|Z} + k^{-1/2} \dot{\eta}_\bullet$, or in equal (02.03) with error processes

$$\begin{aligned} \dot{\epsilon}_\bullet &= n^{1/2}(\hat{\mathbb{P}}_n - U_{\mathbb{P}^{X|Z}})(\psi_\bullet) = n^{1/2}(\hat{\mathbb{P}}_n(\psi_\bullet) - U_{\mathbb{P}^{X|Z}}(\psi_\bullet)) \in \mathcal{M}(\mathcal{B}_{[0,1]} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}} \quad \text{and} \\ \dot{\eta}_\bullet &= k^{1/2}(\hat{\mathbb{P}}_k - U_{\mathbb{P}^{X|Z}})(\psi_\bullet) = k^{1/2}(\hat{\mathbb{P}}_k(\psi_\bullet) - U_{\mathbb{P}^{X|Z}}(\psi_\bullet)) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes 2k} \otimes 2^{\mathbb{N}^2}) \end{aligned}$$

satisfying Assumption §01|01.04 and Assumption §02|01.02. \square

§02|02.08 **Functional linear regression.** Let $(\mathbb{H}, \langle \cdot, \cdot \rangle_{\mathbb{H}})$ be a separable Hilbert space equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{H}}$ and let (X, Y) be an $\mathbb{H} \times \mathbb{R}$ -valued random vector with joint distribution $\mathbb{P}^{X, Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{H}} \otimes \mathcal{B})$. We denote by $\mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{\mathbb{H}})$ the marginal distribution of X . We tactically identify X and Y with the coordinate map $\Pi_{\mathbb{H}}$ and $\Pi_{\mathbb{R}}$, respectively, and thus (X, Y) with the

identity $\text{id}_{\mathbb{H} \times \mathbb{R}}$ such that $\mathbb{P} = \mathbb{P}^{X,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{H}} \otimes \mathcal{B})$. Here and subsequently, we assume that $Y, \|X\|_{\mathbb{H}}^2 \in \mathbb{L}_2(\mathbb{P}) = \mathbb{L}_2(\mathbb{H} \times \mathbb{R}, \mathcal{B}_{\mathbb{H}} \otimes \mathcal{B}, \mathbb{P})$ and $\mathbb{P}(\langle x, X \rangle_{\mathbb{H}}) = 0$ for all $x \in \mathbb{H}$. In this situation X admits a *covariance operator* $\Gamma^X \in \mathbb{L}(\mathbb{H})$ (see [Remark §01101.07](#)) and there is $g \in \mathbb{H}$ satisfying $\langle g, x \rangle_{\mathbb{H}} = \mathbb{P}(Y \langle X, x \rangle_{\mathbb{H}})$ for all $x \in \mathbb{H}$. In what follows we assume that in addition $g \in \text{ran}(\Gamma^X) \subseteq \mathbb{H}$. In this situation there exists $f \in \mathbb{H}$ such that

$$\langle g, x \rangle_{\mathbb{H}} = \mathbb{P}(Y \langle X, x \rangle_{\mathbb{H}}) = \mathbb{P}(\langle X, f \rangle_{\mathbb{H}} \langle X, x \rangle_{\mathbb{H}}) = \langle \Gamma^X f, x \rangle_{\mathbb{H}} \quad \forall x \in \mathbb{H}$$

or in equal \mathbb{P} -a.s. we have $Y = \langle X, f \rangle_{\mathbb{H}} + \xi$ with $\mathbb{P}(\xi \langle X, x \rangle_{\mathbb{H}}) = 0$ for all $x \in \mathbb{H}$. Let us denote by $\mathbb{P}_{\Gamma^X} \in \mathcal{W}(\mathcal{B}_{\mathbb{H}})$ the marginal distribution of X which is not fully specified given $\Gamma^X \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{H})$ (see [Model §02101.08](#)). Moreover, for $\Gamma^X \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{H})$ and $f \in \mathbb{F}_2 \subseteq \mathbb{H}$, and hence $g := \Gamma^X f \in \mathbb{H}$, we denote by \mathbb{P}_{Γ^X} the joint distribution of (X, Y) without fully specifying the distribution which however is assumed to satisfy $\mathbb{P}_{\Gamma^X}(Y \langle X, x \rangle_{\mathbb{H}}) = \mathbb{P}_{\Gamma^X}(\langle X, f \rangle_{\mathbb{H}} \langle X, x \rangle_{\mathbb{H}})$ for all $x \in \mathbb{H}$. Let $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ be a generalised Fourier series transform as in [Notation §01102.07](#) which is fixed and *known* in advance. Following [Model §02101.08](#) $\Gamma_{\bullet}^X := U \Gamma^X U^* \in \mathbb{L}(\ell_2)$ is an operator with kernel (infinite matrix) $\Gamma_{\bullet}^X \in \mathcal{M}(2^{\mathbb{N}})$ satisfying $\Gamma_{\bullet}^X = (\Gamma_{j|j}^X = \langle \Gamma^X u_j, u_j \rangle_{\mathbb{H}} = \mathbb{P}_{\Gamma^X}(\langle X, u_j \rangle_{\mathbb{H}} \langle X, u_j \rangle_{\mathbb{H}}))_{j,j \in \mathbb{N}}$. Therefore the stochastic process $\psi_{\bullet} = (\psi_{j|j}(X) := \langle X, u_j \rangle_{\mathbb{H}} \langle X, u_j \rangle_{\mathbb{H}})_{j,j \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_{\mathbb{H}} \otimes 2^{\mathbb{N}})$ fulfils [Assumption §02101.02](#) and $\Gamma_{\bullet}^X = \mathbb{P}_{\Gamma^X}(\psi_{\bullet})$. Moreover, for each $f \in \mathbb{F}_2 \subseteq \mathbb{H}$ the generalised Fourier coefficients $g = (g_j)_{j \in \mathbb{N}} = U g = U \Gamma^X U^* U f = \Gamma_{\bullet}^X f_{\bullet}$, satisfy $g_{\bullet} = \mathbb{P}_{\Gamma^X}(Y \langle X, u_j \rangle_{\mathbb{H}})$. The stochastic process $\psi = (\psi_j := Y \langle X, u_j \rangle_{\mathbb{H}})_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{\mathbb{H}} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}})$ fulfils [Assumption §01101.04](#) and $g_{\bullet} = \Gamma_{\bullet}^X f_{\bullet} = \mathbb{P}_{\Gamma^X}(\psi)$. The observable noisy versions take the form $\hat{g}_{\bullet} = \Gamma_{\bullet}^X f_{\bullet} + n^{-1/2} \hat{\epsilon}_{\bullet}$ and $\hat{\Gamma}_{\bullet}^X = \Gamma_{\bullet}^X + k^{-1/2} \hat{\eta}_{\bullet}$, or in equal (02.03) with error processes

$$\begin{aligned} \hat{\epsilon}_{\bullet} &= n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{\Gamma^X})(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{P}_{\Gamma^X}(\psi)) \in \mathcal{M}((\mathcal{B}_{\mathbb{H}} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}}) \quad \text{and} \\ \hat{\eta}_{\bullet} &= k^{1/2}(\hat{\mathbb{P}}_k - \mathbb{P}_{\Gamma^X})(\psi_{\bullet}) = k^{1/2}(\hat{\mathbb{P}}_k(\psi_{\bullet}) - \mathbb{P}_{\Gamma^X}(\psi_{\bullet})) \in \mathcal{M}(\mathcal{B}_{\mathbb{H}}^{\otimes k} \otimes 2^{\mathbb{N}}) \end{aligned}$$

satisfying [Assumption §01101.04](#) and [Assumption §02101.02](#). □

§02102.09 Functional linear instrumental regression. Let $(\mathbb{H}, \langle \cdot, \cdot \rangle_{\mathbb{H}})$ and $(\mathbb{G}, \langle \cdot, \cdot \rangle_{\mathbb{G}})$ be separable Hilbert space equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{H}}$ and $\mathcal{B}_{\mathbb{G}}$, respectively, and let (Z, X, Y) be an $\mathbb{G} \times \mathbb{H} \times \mathbb{R}$ -valued random vector with joint distribution $\mathbb{P}^{Z,X,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}} \otimes \mathcal{B})$. We denote by $\mathbb{P}^Z \in \mathcal{W}(\mathcal{B}_{\mathbb{G}})$, $\mathbb{P}^{Z,X} \in \mathcal{W}(\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}})$, and $\mathbb{P}^{Z,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B})$ the marginal distribution of Z , (Z, X) and (Z, Y) , respectively. We tactically identify Z , X and Y with the coordinate map $\Pi_{\mathbb{G}}$, $\Pi_{\mathbb{H}}$ and $\Pi_{\mathbb{R}}$, respectively, and thus (Z, X, Y) with the identity $\text{id}_{\mathbb{G} \times \mathbb{H} \times \mathbb{R}}$ such that $\mathbb{P} = \mathbb{P}^{Z,X,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}} \otimes \mathcal{B})$. Here and subsequently, we assume that $Y, \|Z\|_{\mathbb{G}}^2, \|X\|_{\mathbb{H}}^2 \in \mathbb{L}_2(\mathbb{P}) = \mathbb{L}_2(\mathbb{G} \times \mathbb{H} \times \mathbb{R}, \mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}} \otimes \mathcal{B}, \mathbb{P})$, $\mathbb{P}(\langle z, Z \rangle_{\mathbb{G}}) = 0$ and $\mathbb{P}(\langle x, X \rangle_{\mathbb{H}}) = 0$ for all $z \in \mathbb{G}$ and $x \in \mathbb{H}$. In this situation (Z, X) admits a *cross-covariance operator* $\Gamma^{Z,X} \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ (see [Model §02101.09](#)) and there is $g \in \mathbb{G}$ satisfying $\langle g, z \rangle_{\mathbb{G}} = \mathbb{P}(Y \langle Z, z \rangle_{\mathbb{G}})$ for all $z \in \mathbb{G}$. In what follows we assume that in addition $g \in \text{ran}(\Gamma^{Z,X}) \subseteq \mathbb{G}$. In this situation there exists $f \in \mathbb{H}$ such that

$$\langle g, z \rangle_{\mathbb{G}} = \mathbb{P}(Y \langle Z, z \rangle_{\mathbb{G}}) = \mathbb{P}(\langle X, f \rangle_{\mathbb{H}} \langle Z, z \rangle_{\mathbb{G}}) = \langle \Gamma^{Z,X} f, z \rangle_{\mathbb{G}}$$

or in equal \mathbb{P} -a.s. we have $Y = \langle X, f \rangle_{\mathbb{H}} + \xi$ with $\mathbb{P}(\xi \langle Z, z \rangle_{\mathbb{G}}) = 0$ for all $z \in \mathbb{G}$. Let us denote by $\mathbb{P}_{\Gamma^{Z,X}} \in \mathcal{W}(\mathcal{B}_{\mathbb{G}} \otimes \mathcal{B}_{\mathbb{H}})$ the marginal distribution of (Z, X) which is not fully specified given $\Gamma^{Z,X} \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$ (see [Model §02101.09](#)). Moreover, for $\Gamma^{Z,X} \in \mathbb{T} \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$ and $f \in \mathbb{F}_2 \subseteq \mathbb{H}$, and hence $g := \Gamma^{Z,X} f \in \mathbb{G}$, we denote by $\mathbb{P}_{\Gamma^{Z,X}}$ the joint distribution of (Z, X, Y) without fully specifying the distribution which however is assumed to satisfy $\mathbb{P}_{\Gamma^{Z,X}}(Y \langle Z, z \rangle_{\mathbb{G}}) = \mathbb{P}_{\Gamma^{Z,X}}(\langle X, f \rangle_{\mathbb{H}} \langle Z, z \rangle_{\mathbb{G}})$

for all $z \in \mathbb{G}$. Let $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ and $V \in \mathbb{L}(\mathbb{G}, \ell_2)$ be generalised Fourier series transforms as in **Notation** §01102.07 which are fixed and *known* in advance. Following **Model** §02101.09 $\Gamma_{\cdot}^{Z,X} := V\Gamma^{Z,X}U^* \in \mathbb{L}(\ell_2)$ is an operator with kernel (infinite matrix) $\Gamma_{\cdot}^{Z,X} \in \mathcal{M}(2^{\mathbb{N}})$ satisfying $\Gamma_{\cdot}^{Z,X} = (\Gamma_{j\bar{l}_i}^{Z,X} = \langle \Gamma^{Z,X} u_i, v_j \rangle_{\mathbb{G}} = \mathbb{P}_{\Gamma^{Z,X}}(\langle X, u_i \rangle_{\mathbb{H}} \langle Z, v_j \rangle_{\mathbb{G}}))_{j, \bar{l}_i \in \mathbb{N}}$. Therefore the stochastic process $\psi_{\cdot} = (\psi_{j\bar{l}_i}(Z, X) := \langle X, u_i \rangle_{\mathbb{H}} \langle Z, v_j \rangle_{\mathbb{G}})_{j, \bar{l}_i \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_e \otimes \mathcal{B}_h \otimes 2^{\mathbb{N}})$ fulfils Assumption §02101.02 and $\Gamma_{\cdot}^{Z,X} = \mathbb{P}_{\Gamma^{Z,X}}(\psi_{\cdot})$. Moreover, for each $f \in \mathbb{F}_2 \subseteq \mathbb{H}$ the generalised Fourier coefficients $g = (g_j)_{j \in \mathbb{N}} = Vf = V\Gamma^{Z,X}U^*Uf = \Gamma_{\cdot}^{Z,X}f$, satisfy $g = \mathbb{P}_{\Gamma^{Z,X}}(Y \langle Z, v \rangle_{\mathbb{G}})$. The stochastic process $\psi = (\psi_j(Z, Y) := Y \langle Z, v_j \rangle_{\mathbb{G}})_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_e \otimes \mathcal{B}) \otimes 2^{\mathbb{N}})$ fulfils Assumption §01101.04 and $g = \Gamma_{\cdot}^{Z,X}f = \mathbb{P}_{\Gamma^{Z,X}}(\psi)$. The observable noisy versions take the form $\hat{g} = \Gamma_{\cdot}^{Z,X}f + n^{-1/2}\hat{\epsilon}$ and $\hat{\Gamma}_{\cdot}^{Z,X} = \Gamma_{\cdot}^{Z,X} + k^{-1/2}\hat{\eta}_{\cdot}$, or in equal (02.03) with error processes

$$\begin{aligned} \hat{\epsilon} &= n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{\Gamma^{Z,X}})(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{P}_{\Gamma^{Z,X}}(\psi)) \in \mathcal{M}((\mathcal{B}_e \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}}) \quad \text{and} \\ \hat{\eta}_{\cdot} &= k^{1/2}(\hat{\mathbb{P}}_k - \mathbb{P}_{\Gamma^{Z,X}})(\psi_{\cdot}) = k^{1/2}(\hat{\mathbb{P}}_k(\psi_{\cdot}) - \mathbb{P}_{\Gamma^{Z,X}}(\psi_{\cdot})) \in \mathcal{M}((\mathcal{B}_e \otimes \mathcal{B}_h)^{\otimes k} \otimes 2^{\mathbb{N}}) \end{aligned}$$

satisfying Assumption §01101.04 and Assumption §02101.02. \square

§02102.10 **Regression with unknown design.** Consider the measure space $(\mathcal{D}, \mathcal{B}_{\mathcal{D}}, \lambda_{\mathcal{D}})$ where $\lambda_{\mathcal{D}}$ denotes the restriction of the Lebesgue measure to the Borel- σ -algebra $\mathcal{B}_{\mathcal{D}}$ over $\mathcal{D} \in \mathcal{B}$, and the *real* Hilbert space $\mathbb{L}_2(\lambda_{\mathcal{D}}) := \mathbb{L}_2(\mathcal{D}, \mathcal{B}_{\mathcal{D}}, \lambda_{\mathcal{D}})$ of square Lebesgue-integrable real-valued functions. Let (X, Y) be a $\mathcal{D} \times \mathbb{R}$ -valued random vector. We assume in what follows that the marginal distribution $\mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{\mathcal{D}})$ of the regressor X admits a Lebesgue density $\varphi \in \mathbb{L}_1(\lambda_{\mathcal{D}})$, that is $\mathbb{P}^X = \varphi \lambda_{\mathcal{D}}$, which is *not known* in advance. Moreover, let the joint distribution of (X, Y) be given by $\mathbb{P}_{f|X}^{X,Y} := \varphi \lambda_{\mathcal{D}} \odot \mathbb{P}_{f(X)}^{\xi}$ without fully specifying the error distribution $\mathbb{P}^{\xi} \in \mathcal{W}(\mathcal{B})$ and thus the regular conditional distribution $\mathbb{P}_f^{Y|X} = \mathbb{P}_{f(X)}^{\xi}$ (compare **Model** §01105.10). We tactically identify X and Y with the coordinate map $\Pi_{\mathcal{D}}$ and $\Pi_{\mathbb{R}}$, respectively, and thus (X, Y) with the identity $\text{id}_{\mathcal{D} \times \mathbb{R}}$ such that $\mathbb{P}_{f|X} = \mathbb{P}_{f|X}^{X,Y} \in \mathcal{W}(\mathcal{B}_{\mathcal{D}} \otimes \mathcal{B})$. In addition we assume that $\varphi \in \mathbb{L}_{\infty}(\lambda_{\mathcal{D}})$, and hence $M_{\varphi} \in \mathbb{L}(\mathbb{L}_2(\lambda_{\mathcal{D}}))$, $\mathbb{P}^{\xi} \in \mathbb{P}_{\{0\} \times \mathbb{R}_{\geq 0}} \subseteq \mathcal{W}_2(\mathcal{B})$, i.e. ξ has mean zero and a finite second moment, and $f \in \mathbb{L}_2(\lambda_{\mathcal{D}})$, then $g = M_{\varphi}f \in \mathbb{L}_2(\lambda_{\mathcal{D}})$ for each $h \in \mathbb{L}_2(\lambda_{\mathcal{D}})$ satisfies

$$\mathbb{P}_{f|X}(Yh(X)) = \mathbb{P}^X(\mathbb{P}_f^{Y|X}(Y)h) = \varphi \lambda_{\mathcal{D}}(fh) = \lambda_{\mathcal{D}}(\varphi fh) = \langle M_{\varphi}f, h \rangle_{\mathbb{L}_2(\lambda_{\mathcal{D}})} = \langle g, h \rangle_{\mathbb{L}_2(\lambda_{\mathcal{D}})} \in \mathbb{R}.$$

Let $U \in \mathbb{L}(\mathbb{L}_2(\lambda_{\mathcal{D}}), \ell_2)$ be a generalised Fourier series transform as in **Notation** §01102.07 which is fixed and *known in advance*. Evidently, we have $M_{\cdot}^{\varphi} := UM_{\varphi}U^* \in \mathbb{L}(\ell_2) \subseteq \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$ and for each $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{\mathcal{D}})$ and $g := M_{\varphi}f \in \mathbb{L}_2(\lambda_{\mathcal{D}})$ the generalised Fourier coefficients $f_{\cdot} = (f_j)_{j \in \mathbb{N}} = Uf$ and $g_{\cdot} = (g_j)_{j \in \mathbb{N}} = Ug = M_{\cdot}^{\varphi}f_{\cdot}$ for each $j \in \mathbb{N}$ satisfy

$$g_j = \langle M_{j\bar{l}_i}^{\varphi}, f \rangle_{\ell_2} = \langle M_{j\bar{l}_i}^{\varphi}, f_{\cdot} \rangle_{\ell_2} = \langle M_{\varphi}f, U^* \mathbf{1}^{j\bar{l}_i} \rangle_{\mathbb{L}_2(\lambda_{\mathcal{D}})} = \langle M_{\varphi}f, u_j \rangle_{\mathbb{L}_2(\lambda_{\mathcal{D}})} = \mathbb{P}_{f|X}(Y u_j(X)) \in \mathbb{R}$$

The stochastic process $\psi = (\psi_j(X, Y) := Y u_j(X))_{j \in \mathbb{N}} \in \mathcal{M}((\mathcal{B}_{\mathcal{D}} \otimes \mathcal{B}) \otimes 2^{\mathbb{N}})$ fulfils Assumption §01101.04 and $g_{\cdot} = M_{\cdot}^{\varphi}f_{\cdot} = \mathbb{P}_{f|X}(\psi)$ (compare **Model** §01105.10). Moreover, considering the marginal distribution $\mathbb{P}_{\varphi} = \varphi \lambda_{\mathcal{D}}$ of X we have $M_{\cdot}^{\varphi} := UM_{\varphi}U^* \in \mathbb{L}(\ell_2) \subseteq \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$ satisfying $M_{\cdot}^{\varphi} = (M_{j\bar{l}_i}^{\varphi} = \langle M_{\cdot}^{\varphi} u_i, u_j \rangle_{\mathbb{G}} = \mathbb{P}_{\varphi}(u_i u_j))_{j, \bar{l}_i \in \mathbb{N}}$. Therefore the stochastic process $\psi_{\cdot} = (\psi_{j\bar{l}_i} := u_j(X) u_j(X))_{j, \bar{l}_i \in \mathbb{N}} \in \mathcal{M}(\mathcal{B}_{\mathcal{D}} \otimes 2^{\mathbb{N}})$ fulfils Assumption §02101.02 and $M_{\cdot}^{\varphi} = \mathbb{P}_{\varphi}(\psi_{\cdot})$ (compare **Model** §02101.10). The observable noisy versions take the form $\hat{g}_{\cdot} = M_{\cdot}^{\varphi}f_{\cdot} + n^{-1/2}\hat{\epsilon}$ and $\hat{M}_{\cdot}^{\varphi} = M_{\cdot}^{\varphi} + n^{-1/2}\hat{\eta}_{\cdot}$, or in equal (02.03) with error processes

$$\begin{aligned} \hat{\epsilon} &= n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{f|X})(\psi) = n^{1/2}(\hat{\mathbb{P}}_n(\psi) - \mathbb{P}_{f|X}(\psi)) \in \mathcal{M}((\mathcal{B}_{\mathcal{D}} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{N}}) \quad \text{and} \\ \hat{\eta}_{\cdot} &= k^{1/2}(\hat{\mathbb{P}}_k - \mathbb{P}_{\varphi})(\psi_{\cdot}) = k^{1/2}(\hat{\mathbb{P}}_k(\psi_{\cdot}) - \mathbb{P}_{\varphi}(\psi_{\cdot})) \in \mathcal{M}(\mathcal{B}_{\mathcal{D}}^{\otimes k} \otimes 2^{\mathbb{N}}) \end{aligned}$$

satisfying Assumption §01101.04 and Assumption §02101.02. \square

§02|03 Noisy diagonal operator

§02|03.01 **Notation.** Under Assumption §02|00.01 we consider the reconstruction of $\theta = U\theta \in \mathbb{J}$ (or in equal $\theta = U^*\theta \in \mathbb{H}$) from noisy versions of $Vg = g = \mathfrak{s}\theta \in \mathbb{J}$ and $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ or in equal $M_{\mathfrak{s}} = VTU^* \in \mathbb{L}^m(\mathbb{J})$. \square

§02|03.02 **Noisy diagonal operator.** Let $\dot{\eta} = (\dot{\eta}_j)_{j \in \mathcal{J}}$ be a stochastic process satisfying Assumption §01|01.04 with mean zero and let $k \in \mathbb{N}$ be a sample size. The stochastic process $\widehat{\mathfrak{s}} = \mathfrak{s} + k^{-1/2}\dot{\eta}$ with mean function $\mathfrak{s} \in \mathcal{J}$ is called a *noisy version* of $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ and hence the diagonal operator $M_{\widehat{\mathfrak{s}}} \in \mathbb{L}^m(\mathbb{J})$, or *noisy diagonal operator* for short. We denote by $\mathbb{P}_{\mathfrak{s}}^k$ the distribution of $\widehat{\mathfrak{s}}$. If $\dot{\eta}$ admits a covariance function (possibly depending on \mathfrak{s}), say $\text{cov}_{\mathfrak{s}} \in \mathcal{J}^2$, then we eventually write $\dot{\eta} \sim P_{(0, \text{cov}_{\mathfrak{s}})}$ and $\widehat{\mathfrak{s}} \sim P_{(\mathfrak{s}, k^{-1}\text{cov}_{\mathfrak{s}})}$ for short. \square

§02|03.03 **Comment.** Similar to a noisy image (Definition §01|02.03) we consider Empirical mean model §01|02.04, Sequence model §01|02.05 or Gaussian sequence model §01|02.06. Examples are provided in Subsubsection §01|02|01. \square

§02|03|01 Examples of empirical mean models

§02|03.04 **Covariance operator under second order stationarity.** Consider the *complex* Hilbert spaces $\mathbb{L}_2(\lambda_{[0,1]})$ and $\mathbb{J} := \ell_2(\mathbb{Z})$. Let $(\mathbb{L}_2(\lambda_{[0,1]}), \langle \cdot, \cdot \rangle_{\mathbb{L}_2(\lambda_{[0,1]})})$ be equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}$ and X be a $\mathbb{L}_2(\lambda_{[0,1]})$ -valued *real random function* (by the usual embedding of real-valued functions as in Notation §01|02.10). We tactically identify X with the identity $\text{id}_{\mathbb{L}_2(\lambda_{[0,1]})}$ on $\mathbb{L}_2(\lambda_{[0,1]})$ such that X is defined on the measure space $(\mathbb{L}_2(\lambda_{[0,1]}), \mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}, \mathbb{P})$ and $X \sim \mathbb{P} = \mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})})$. Here and subsequently, we assume that $\|X\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2 \in \mathbb{L}_2(\mathbb{P})$ and $\mathbb{P}(\langle X, x \rangle_{\mathbb{L}_2(\lambda_{[0,1]})}) = 0$ for all $x \in \mathbb{L}_2(\lambda_{[0,1]})$. In this situation X admits a *covariance operator* $\Gamma^X \in \mathbb{L}^{\mathbb{L}_2(\lambda_{[0,1]})}$ (see Remark §01|01.07). Moreover, let X be *second order stationary*, i.e. there exists $c^X \in \mathcal{M}(\mathcal{B}_{[0,1]})$ such that

$$\text{cov}_{t,s} = \text{Cov}(X(t), X(s)) = c^X(t - s - \lfloor t - s \rfloor), \quad \forall s, t \in [0, 1].$$

Evidently, since $\|X\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2 \in \mathbb{L}_2(\mathbb{P})$ we have

$$\begin{aligned} \|c^X\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2 &= \lambda_{[0,1]}(|c^X|^2) = \int_{[0,1]} |\text{Cov}(X(0), X(t))|^2 \lambda_{[0,1]}(dt) \\ &\leq \mathbb{P}(|X(0)|^2) \int_{[0,1]} \mathbb{P}(|X(t)|^2) \lambda_{[0,1]}(dt) = \mathbb{P}(|X(0)|^2) \mathbb{P}(\|X\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2) \in \mathbb{R}_{\geq 0} \end{aligned}$$

and hence $c^X \in \mathbb{L}_2(\lambda_{[0,1]})$ too. Furthermore, the *covariance operator* $\Gamma^X \in \mathbb{L}^{\mathbb{L}_2(\lambda_{[0,1]})}$ equals a *circular additive convolution* (see Notation §01|04.11), since for all $x, y \in \mathbb{L}_2(\lambda_{[0,1]})$

$$\begin{aligned} \langle \Gamma^X x, y \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} &= \text{Cov}(\langle X, y \rangle_{\mathbb{L}_2(\lambda_{[0,1]})}, \langle X, x \rangle_{\mathbb{L}_2(\lambda_{[0,1]})}) = \mathbb{P}(\lambda_{[0,1]}(X\bar{y}) \overline{\lambda_{[0,1]}(X\bar{x})}) \\ &= \int_{[0,1]} \int_{[0,1]} \bar{y}(t) \text{Cov}(X(t), X(s)) x(s) \lambda_{[0,1]}(ds) \lambda_{[0,1]}(dt) \\ &= \int_{[0,1]} \bar{y}(t) \int_{[0,1]} c^X(t - s - \lfloor t - s \rfloor) x(s) \lambda_{[0,1]}(ds) \lambda_{[0,1]}(dt) \\ &= \int_{[0,1]} (c^X \otimes x)(t) \bar{y}(t) \lambda_{[0,1]}(dt) = \langle \otimes_{c^X} x, y \rangle_{\mathbb{L}_2(\lambda_{[0,1]})}, \end{aligned}$$

hence $\otimes_{c^X} = \Gamma^X \in \mathbb{L}^{\mathbb{L}_2(\lambda_{[0,1]})}$ in short. Let $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2(\mathbb{Z}))$ be the *Fourier-series transform* with $h \mapsto Fh := h_{\bullet} = \lambda_{[0,1]}(h\bar{e}_{\bullet})$ and exponential basis $e_{\bullet} := (e_j)_{j \in \mathbb{Z}}$ (see Notations §01|02.10

and §01102.12). Since $c^X \in \mathbb{L}_2(\lambda_{[0,1]})$ we denote $c^X = Fc^X$. Then the *circular convolution theorem* states $(\otimes_{c^X} h)_\bullet = F(\otimes_{c^X} h) = (Fc^X)(Fh) = c^X h_\bullet$. Consequently, (c^X, F, F) is an eigen value decomposition of $\otimes_{c^X} \in \mathbb{L}^{\mathbb{Z}}(\mathbb{L}_2(\lambda_{[0,1]}))$ with $c^X \in \ell_2(\mathbb{Z}) \subseteq \ell_\infty(\mathbb{Z})$, and thus $\Gamma^X = \otimes_{c^X} \in \mathbb{L}^{F, F}(\mathbb{L}^{\mathbb{Z}}(\mathbb{L}_2(\mathbb{Z}))) = F^*(\mathbb{L}^{\mathbb{Z}}(\ell_2(\mathbb{Z})))F$. Let us denote by $\mathbb{P}_{c^X} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})})$ the distribution of X which is not fully specified given $c^X \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]})$. We consider the statistical product experiment

$$((\mathbb{L}_2(\lambda_{[0,1]}))^k, \mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}^{\otimes k}, \mathbb{P}_{\mathbb{D}_2}^{\otimes k} = (\mathbb{P}_{c^X}^{\otimes k})_{c^X \in \mathbb{D}_2}).$$

The stochastic process $\psi_\bullet = (\psi_j(X) := |\langle X, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})}|^2)_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})} \otimes 2^{\mathbb{Z}})$ fulfils Assumption §01101.04 and $c^X = \mathbb{P}_{c^X}(\psi_\bullet)$ since for each $j \in \mathbb{Z}$ we have

$$\begin{aligned} \mathbb{P}_{c^X}(|\langle X, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})}|^2) &= \langle \Gamma^X e_j, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \langle \otimes_{c^X} e_j, e_j \rangle_{\mathbb{L}_2(\lambda_{[0,1]})} = \langle c^X F e_j, F e_j \rangle_{\ell_2(\mathbb{Z})} \\ &= \langle c^X \mathbb{1}^{\{j\}}, \mathbb{1}^{\{j\}} \rangle_{\ell_2(\mathbb{Z})} = c_j^X. \end{aligned}$$

Similar to an Empirical mean model §01102.04 we define $\widehat{c}^X = (\widehat{c}_j^X := \widehat{\mathbb{P}}_k(\psi_j))_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}^{\otimes k} \otimes 2^{\mathbb{Z}})$. By construction $c^X = \mathbb{P}_{c^X}(\psi_\bullet) \in \mathcal{M}(2^{\mathbb{Z}})$ is the mean sequence of \widehat{c}^X . For each $j \in \mathbb{Z}$ the statistic $\dot{\eta}_j := k^{1/2}(\widehat{\mathbb{P}}_k(\psi_j) - \mathbb{P}_{c^X}(\psi_j)) \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}^{\otimes k})$ is centred, i.e. $\dot{\eta}_j \in \mathbb{L}_1(\mathbb{P}_{c^X}^{\otimes k})$ with $\mathbb{P}_{c^X}^{\otimes k}(\dot{\eta}_j) = 0$, and exploiting $\psi_\bullet \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})} \otimes 2^{\mathbb{Z}})$ the stochastic process

$$\dot{\eta}_\bullet = (\dot{\eta}_j)_{j \in \mathbb{Z}} = k^{1/2}(\widehat{\mathbb{P}}_k - \mathbb{P}_{c^X})(\psi_\bullet) = k^{1/2}(\widehat{\mathbb{P}}_k(\psi_\bullet) - \mathbb{P}_{c^X}(\psi_\bullet)) \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}^{\otimes k} \otimes 2^{\mathbb{Z}})$$

satisfies Assumption §01101.04 and by construction $\widehat{c}^X = c^X + k^{-1/2}\dot{\eta}_\bullet$ is a noisy version of c^X . \square

§02103.05 **Cross-covariance operator under second order stationarity.** Consider the *complex* Hilbert spaces $\mathbb{L}_2(\lambda_{[0,1]})$ and $\mathbb{J} := \ell_2(\mathbb{Z})$. Let $(\mathbb{L}_2(\lambda_{[0,1]}), \langle \cdot, \cdot \rangle_{\mathbb{L}_2(\lambda_{[0,1]})})$ be equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}$ and let Z and X be a $\mathbb{L}_2(\lambda_{[0,1]})$ -valued *real random function* (by the usual embedding of real-valued functions as in **Notation** §01102.10). Then (Z, X) is an $(\mathbb{L}_2(\lambda_{[0,1]})^2, \mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}^{\otimes 2})$ -valued *random function*. We denote by $\mathbb{P}^Z, \mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})})$ the marginal distribution of Z and X , respectively, and by $\mathbb{P}^{Z, X} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}^{\otimes 2})$ the joint distribution of (Z, X) . We tactically take Z and X as coordinate map, and thus identify (Z, X) with the identity $\text{id}_{(\mathbb{L}_2(\lambda_{[0,1]})^2)}$ such that $\mathbb{P} = \mathbb{P}^{Z, X} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{[0,1]})}^{\otimes 2})$. Here and subsequently, we assume that $\|Z\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2 \in \mathcal{L}_2(\mathbb{P})$, $\|X\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2 \in \mathcal{L}_2(\mathbb{P})$, $\mathbb{P}(\langle Z, z \rangle_{\mathbb{L}_2(\lambda_{[0,1]})}) = 0$ and $\mathbb{P}(\langle X, x \rangle_{\mathbb{L}_2(\lambda_{[0,1]})}) = 0$ for all $z, x \in \mathbb{L}_2(\lambda_{[0,1]})$. In this situation (Z, X) admits a *cross-covariance operator* $\Gamma^{Z, X} \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}))$ (see Model §02101.09). Moreover, let (Z, X) be *second order stationary*, i.e. there exists $c^{Z, X} \in \mathcal{M}(\mathcal{B}_{[0,1]})$ such that

$$\text{cov}_{t,s}^{Z, X} = \text{Cov}(Z(t), X(s)) = c^{Z, X}(t - s - \lfloor t - s \rfloor), \quad \forall s, t \in [0, 1].$$

Evidently, since $\|X\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2 \in \mathbb{L}_2(\mathbb{P})$ we have

$$\begin{aligned} \|c^{Z, X}\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2 &= \lambda_{[0,1]}(|c^{Z, X}|^2) = \int_{[0,1]} |\text{Cov}(Z(0), X(t))|^2 \lambda_{[0,1]}(dt) \\ &\leq \mathbb{P}(|Z(0)|^2) \int_{[0,1]} \mathbb{P}(|X(t)|^2) \lambda_{[0,1]}(dt) = \mathbb{P}(|Z(0)|^2) \mathbb{P}(\|X\|_{\mathbb{L}_2(\lambda_{[0,1]})}^2) \in \mathbb{R}_{\geq 0} \end{aligned}$$

and hence $c^{Z, X} \in \mathbb{L}_2(\lambda_{[0,1]})$ too. Furthermore, the *cross-covariance operator* $\Gamma^{Z, X} \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}))$ equals

a *circular additive convolution* (see [Notation §01104.11](#)), since for all $x, y \in \mathbb{L}_2(\lambda_{0,1})$

$$\begin{aligned} \langle \Gamma^{ZX} x, z \rangle_{\mathbb{L}_2(\lambda_{0,1})} &= \text{Cov}(\langle Z, z \rangle_{\mathbb{L}_2(\lambda_{0,1})}, \langle X, x \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = \mathbb{P}(\lambda_{0,1})(Z \bar{z} \overline{\lambda_{0,1}(X \bar{x})}) \\ &= \int_{[0,1)} \int_{[0,1)} \bar{z}(t) \text{Cov}(Z(t), X(s)) x(s) \lambda_{0,1}(ds) \lambda_{0,1}(dt) \\ &= \int_{[0,1)} \bar{z}(t) \int_{[0,1)} c^{ZX}(t-s - \lfloor t-s \rfloor) x(s) \lambda_{0,1}(ds) \lambda_{0,1}(dt) \\ &= \int_{[0,1)} (c^{ZX} \otimes x)(t) \bar{z}(t) \lambda_{0,1}(dt) = \langle \otimes_{c^{ZX}} x, z \rangle_{\mathbb{L}_2(\lambda_{0,1})}, \end{aligned}$$

hence $\otimes_{c^{ZX}} = \Gamma^{ZX} \in \mathbb{L}(\mathbb{L}_2(\lambda_{0,1}))$ in short. Let $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{0,1}), \ell_2(\mathbb{Z}))$ be the *Fourier-series transform* with $h \mapsto Fh := \mathbf{h}_\bullet = \lambda_{0,1}(h \bar{\mathbf{e}}_\bullet)$ and exponential basis $\mathbf{e}_\bullet := (e_j)_{j \in \mathbb{Z}}$ (see [Notations §01102.10](#) and [§01102.12](#)). Since $c^{ZX} \in \mathbb{L}_2(\lambda_{0,1})$ we denote $\mathbf{c}_\bullet^{ZX} = F c^{ZX}$. Then the *circular convolution theorem* states $(\otimes_{c^{ZX}} h)_\bullet = F(\otimes_{c^{ZX}} h) = (F c^{ZX})(Fh) = \mathbf{c}_\bullet^{ZX} \mathbf{h}_\bullet$. Consequently, $(\mathbf{c}_\bullet^{ZX}, F, F)$ is a singular value decomposition of $\otimes_{c^{ZX}} \in \mathbb{L}(\mathbb{L}_2(\lambda_{0,1}))$ with $\mathbf{c}_\bullet^{ZX} \in \ell_2(\mathbb{Z}) \subseteq \ell_\infty(\mathbb{Z})$, and thus $\Gamma^{ZX} = \otimes_{c^{ZX}} \in \mathbb{L}^{\mathbb{F}, \mathbb{F}}(\mathbb{L}^{\mathbb{M}}(\ell_2(\mathbb{Z}))) = \mathbb{F}^*(\mathbb{L}^{\mathbb{M}}(\ell_2(\mathbb{Z})))\mathbb{F}$. Let us denote by $\mathbb{P}_{c^{ZX}} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})})$ the joint distribution of (Z, X) which is not fully specified given $c^{ZX} \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$. We consider the statistical product experiment

$$((\mathbb{L}_2(\lambda_{0,1}))^{2k}, \mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^{\otimes 2k}, \mathbb{P}_{\mathbb{D}_2}^{\otimes k} = (\mathbb{P}_{c^{ZX}}^{\otimes k})_{c^{ZX} \in \mathbb{D}_2}).$$

The stochastic process $\psi_\bullet = (\psi_j(Z, X) := \langle Z, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})}, \langle e_j, X \rangle_{\mathbb{L}_2(\lambda_{0,1})})_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^{\otimes 2} \otimes 2^{\mathbb{Z}})$ fulfils Assumption [§01101.04](#) and $\mathbf{c}_\bullet^{ZX} = \mathbb{P}_{c^{ZX}}(\psi_\bullet)$ since for each $j \in \mathbb{Z}$ we have

$$\begin{aligned} \mathbb{P}_{c^{ZX}}(\langle Z, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})}, \langle e_j, X \rangle_{\mathbb{L}_2(\lambda_{0,1})}) &= \text{Cov}(\langle Z, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})}, \langle X, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})}) \\ &= \langle \Gamma^{ZX} e_j, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})} = \langle \otimes_{c^{ZX}} e_j, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})} = \langle \mathbf{c}_\bullet^{ZX} F e_j, F e_j \rangle_{\ell_2(\mathbb{Z})} = \mathbf{c}_j^{ZX}. \end{aligned}$$

Similar to an Empirical mean model [§01102.04](#) we define $\widehat{\mathbf{c}}_\bullet^{ZX} = (\widehat{\mathbf{c}}_j^{ZX} := \widehat{\mathbb{P}}_k(\psi_j))_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^{\otimes 2k} \otimes 2^{\mathbb{Z}})$. By construction $\mathbf{c}_\bullet^{ZX} = \mathbb{P}_{c^{ZX}}(\psi_\bullet) \in \mathcal{M}(2^{\mathbb{Z}})$ is the mean sequence of $\widehat{\mathbf{c}}_\bullet^{ZX}$. For each $j \in \mathbb{Z}$ the statistic $\dot{\eta}_j := k^{1/2}(\widehat{\mathbb{P}}_k(\psi_j) - \mathbb{P}_{c^{ZX}}(\psi_j)) \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^{\otimes 2k})$ is centred, i.e. $\dot{\eta}_j \in \mathbb{L}_1(\mathbb{P}_{c^{ZX}}^{\otimes k})$ with $\mathbb{P}_{c^{ZX}}^{\otimes k}(\dot{\eta}_j) = 0$, and exploiting $\psi_\bullet \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^{\otimes 2} \otimes 2^{\mathbb{Z}})$ the stochastic process

$$\dot{\eta}_\bullet = (\dot{\eta}_j)_{j \in \mathbb{Z}} = k^{1/2}(\widehat{\mathbb{P}}_k - \mathbb{P}_{c^{ZX}})(\psi_\bullet) = k^{1/2}(\widehat{\mathbb{P}}_k(\psi_\bullet) - \mathbb{P}_{c^{ZX}}(\psi_\bullet)) \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^{\otimes 2k} \otimes 2^{\mathbb{Z}})$$

satisfies Assumption [§01101.04](#) and by construction $\widehat{\mathbf{c}}_\bullet^{ZX} = \mathbf{c}_\bullet^{ZX} + k^{-1/2} \dot{\eta}_\bullet$ is a noisy version of \mathbf{c}_\bullet^{ZX} . \square

§02|04 Diagonal statistical inverse problem with noisy operator

[§02104.01](#) **Assumption.** For $\mathbb{J} = \mathbb{L}_2(\nu)$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$, fixed and presumed to be known in advance, the operator $T \in \mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J})) \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$ and hence $M_\mathfrak{s} = VTU^* \in \mathbb{L}^{\mathbb{M}}(\mathbb{J})$ or in equal $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ is *not known* in advance where $g = VT\theta = M_\mathfrak{s}\theta = \mathfrak{s}\theta \in \mathbb{J}$ or in equal $g \in \mathbb{J}\mathfrak{s}$. \square

[§02104.02](#) **Notation.** Under Assumption [§02104.01](#) given $g \in \mathbb{J}\mathfrak{s}$ for $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ we consider the reconstruction of $\theta = U\theta \in \mathbb{J}$ (or in equal $\theta = U^*\theta \in \mathbb{H}$) from a noisy version of $g = Vg = M_\mathfrak{s}\theta = \mathfrak{s}\theta \in \mathbb{J}$ and a noisy version of $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$. Keep in mind, that we identify the equivalence class and its representative g . \square

§02104.03 **Diagonal statistical inverse problem with noisy operator.** Consider as in Definition §01102.03 stochastic processes $\dot{\boldsymbol{\epsilon}} = (\dot{\boldsymbol{\epsilon}}_j)_{j \in \mathcal{J}}$ and $\dot{\boldsymbol{\eta}} = (\dot{\boldsymbol{\eta}}_j)_{j \in \mathcal{J}}$ satisfying Assumption §01101.04 with mean zero and sample sizes $n, k \in \mathbb{N}$. Under Assumption §02104.01 where $\boldsymbol{s}_\bullet \in \mathbb{L}_\infty(\nu)$ is *not known* in advanced, the observable noisy image and operator, respectively, has \mathbb{J} -mean $\boldsymbol{g}_\bullet = \boldsymbol{s}_\bullet \boldsymbol{\theta}$ and mean-function $\boldsymbol{s}_\bullet \in \mathbb{L}_\infty(\nu)$, and takes the form $\widehat{\boldsymbol{g}}_\bullet = \boldsymbol{g}_\bullet + n^{-1/2} \dot{\boldsymbol{\epsilon}}_\bullet$ and $\widehat{\boldsymbol{s}}_\bullet = \boldsymbol{s}_\bullet + k^{-1/2} \dot{\boldsymbol{\eta}}_\bullet$ or in equal

$$\widehat{\boldsymbol{g}}_j = \boldsymbol{s}_j \boldsymbol{\theta}_j + n^{-1/2} \dot{\boldsymbol{\epsilon}}_j \quad \text{and} \quad \widehat{\boldsymbol{s}}_j = \boldsymbol{s}_j + k^{-1/2} \dot{\boldsymbol{\eta}}_j, \quad j \in \mathcal{J}. \quad (02.06)$$

We denote by $\mathbb{P}_{\boldsymbol{\theta}_\bullet}^{n,k}$ the joint distribution of $(\widehat{\boldsymbol{g}}_\bullet, \widehat{\boldsymbol{s}}_\bullet)$. The reconstruction of $\boldsymbol{\theta} \in \mathbb{J}$ (in equal $\boldsymbol{\theta} = \mathbf{U}^* \boldsymbol{\theta} \in \mathbb{H}$) from a noisy version $(\widehat{\boldsymbol{g}}_\bullet, \widehat{\boldsymbol{s}}_\bullet) \sim \mathbb{P}_{\boldsymbol{\theta}_\bullet}^{n,k}$ of the image $\boldsymbol{g}_\bullet = \boldsymbol{s}_\bullet \boldsymbol{\theta} \in \mathbb{J}$ and $\boldsymbol{s}_\bullet \in \mathbb{L}_\infty(\nu)$ is called a *diagonal statistical inverse problem with noisy operator*. \square

§02104.04 **Diagonal inverse empirical mean model (dieMM) with noisy operator.** Consider the reconstruction of $\boldsymbol{\theta} \in \mathbb{J}$ (in equal $\boldsymbol{\theta} = \mathbf{U}^* \boldsymbol{\theta} \in \mathbb{H}$) in an Empirical mean model as in §01102.04. Under Assumption §02104.01, where $\mathbb{M}_\bullet \in \mathbb{L}(\mathbb{J})$ with $\boldsymbol{s}_\bullet \in \mathbb{L}_\infty(\nu)$ is *not known* in advance, the observable noisy image has \mathbb{J} -mean $\mathbb{V}g = \boldsymbol{g}_\bullet = \boldsymbol{s}_\bullet \boldsymbol{\theta} \in \mathbb{J}$ and the observable noisy diagonal operator has mean function $\boldsymbol{s}_\bullet \in \mathbb{L}_\infty(\nu)$, and takes each the form of an Empirical mean model as in §01102.04. More precisely, for each $\boldsymbol{\theta} \in \Theta \subseteq \mathbb{H}$ and $\boldsymbol{s}_\bullet \in \mathcal{S} \subseteq \mathbb{L}_\infty(\nu)$ let $\mathbb{P}_{\boldsymbol{\theta}_\bullet} \in \mathcal{W}(\mathcal{Z})$ be a probability measure on a measurable space $(\mathcal{Z}, \mathcal{Z})$. Similar to §01102.04 consider stochastic processes $\psi_j^\ominus, \psi_j^\ominus \in \mathcal{Z} \otimes \mathcal{J}$ which in addition for all $\boldsymbol{\theta} \in \Theta$ and $\boldsymbol{s}_\bullet \in \mathcal{S}$ satisfy $\psi_j^\ominus, \psi_j^\ominus \in \mathcal{L}_1(\mathbb{P}_{\boldsymbol{\theta}_\bullet})$ for each $j \in \mathcal{J}$ and $\mathbb{P}_{\boldsymbol{\theta}_\bullet}(\psi_j^\ominus) = \boldsymbol{g}_\bullet = \boldsymbol{s}_\bullet \boldsymbol{\theta}$ and $\mathbb{P}_{\boldsymbol{\theta}_\bullet}(\psi_j^\ominus) = \boldsymbol{s}_\bullet$. The observable noisy versions take the form $\widehat{\boldsymbol{g}}_\bullet = \boldsymbol{s}_\bullet \boldsymbol{\theta}_\bullet + n^{-1/2} \dot{\boldsymbol{\epsilon}}_\bullet$ and $\widehat{\boldsymbol{s}}_\bullet = \boldsymbol{s}_\bullet + k^{-1/2} \dot{\boldsymbol{\eta}}_\bullet$, or in equal (02.06) with error processes

$$\begin{aligned} \dot{\boldsymbol{\epsilon}}_\bullet &= n^{1/2}(\widehat{\mathbb{P}}_n - \mathbb{P}_{\boldsymbol{\theta}_\bullet})(\psi_j^\ominus) = n^{1/2}(\widehat{\mathbb{P}}_n(\psi_j^\ominus) - \mathbb{P}_{\boldsymbol{\theta}_\bullet}(\psi_j^\ominus)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J}) \quad \text{and} \\ \dot{\boldsymbol{\eta}}_\bullet &= k^{1/2}(\widehat{\mathbb{P}}_k - \mathbb{P}_{\boldsymbol{\theta}_\bullet})(\psi_j^\ominus) = k^{1/2}(\widehat{\mathbb{P}}_k(\psi_j^\ominus) - \mathbb{P}_{\boldsymbol{\theta}_\bullet}(\psi_j^\ominus)) \in \mathcal{M}(\mathcal{Z}^{\otimes k} \otimes \mathcal{J}) \end{aligned}$$

satisfying Assumption §01101.04. \square

§02104.05 **Diagonal inverse sequence model (diSM) with noisy operator.** Consider $\mathbb{J} = \ell_2 = \mathbb{L}_2(\nu_{\mathbb{N}})$ as in §01101.14. Let $\dot{\boldsymbol{\epsilon}} = (\dot{\boldsymbol{\epsilon}}_j)_{j \in \mathbb{N}}$ and $\dot{\boldsymbol{\eta}} = (\dot{\boldsymbol{\eta}}_j)_{j \in \mathbb{N}}$ be real-valued stochastic processes satisfying Assumption §01101.04 with mean zero and let $n, k \in \mathbb{N}$ be sample sizes. Under Assumption §02104.01, where $\mathbb{M}_\bullet \in \mathbb{L}(\mathbb{J})$ with $\boldsymbol{s}_\bullet \in \mathbb{L}_\infty(\nu)$ is *not known* in advance, the observable noisy image has ℓ_2 -mean $\boldsymbol{g}_\bullet = \mathbb{T}_\bullet \boldsymbol{\theta}$ and the observable noisy operator has mean function $\boldsymbol{s}_\bullet \in \mathbb{L}_\infty(\nu)$, and take both the form of a Sequence model as in §01102.05, that is $\widehat{\boldsymbol{g}}_\bullet = \mathbb{T}_\bullet \boldsymbol{\theta}_\bullet + n^{-1/2} \dot{\boldsymbol{\epsilon}}_\bullet$ and $\widehat{\boldsymbol{s}}_\bullet = \boldsymbol{s}_\bullet + k^{-1/2} \dot{\boldsymbol{\eta}}_\bullet$ or in equal

$$\widehat{\boldsymbol{g}}_j = \boldsymbol{s}_j \boldsymbol{\theta}_j + n^{-1/2} \dot{\boldsymbol{\epsilon}}_j \quad \text{and} \quad \widehat{\boldsymbol{s}}_j = \boldsymbol{s}_j + k^{-1/2} \dot{\boldsymbol{\eta}}_j, \quad j \in \mathbb{N}. \quad (02.07)$$

We denote by $\mathbb{P}_{\boldsymbol{\theta}_\bullet}^{n,k}$ the joint distribution of $(\widehat{\boldsymbol{g}}_\bullet, \widehat{\boldsymbol{s}}_\bullet)$. \square

§02104.06 **Gaussian diagonal inverse sequence model (GdiSM) with noisy operator.** Consider Gaussian white noise processes $\dot{\mathbb{B}}_\bullet := (\dot{\mathbb{B}}_j)_{j \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{\mathbb{W}}_\bullet := (\dot{\mathbb{W}}_j)_{j \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}$. The observable noisy versions $\widehat{\boldsymbol{g}}_\bullet = \boldsymbol{g}_\bullet + n^{-1/2} \dot{\mathbb{B}}_\bullet$ with ℓ_2 -mean $\boldsymbol{g}_\bullet = \mathbb{T}_\bullet \boldsymbol{\theta}$ and $\widehat{\boldsymbol{s}}_\bullet = \boldsymbol{s}_\bullet + k^{-1/2} \dot{\mathbb{W}}_\bullet$ with mean function $\boldsymbol{s}_\bullet \in \mathbb{L}_\infty(\nu)$ take both the form of a Gaussian sequence model as in §01102.06, that is

$$\begin{aligned} \widehat{\boldsymbol{g}}_j &= \boldsymbol{s}_j \boldsymbol{\theta}_j + n^{-1/2} \dot{\mathbb{B}}_j \quad \text{and} \quad \widehat{\boldsymbol{s}}_j = \boldsymbol{s}_j + k^{-1/2} \dot{\mathbb{W}}_j, \quad j \in \mathbb{N} \\ &\quad \text{with} \quad (\dot{\mathbb{B}}_j)_{j \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}} \quad \text{and} \quad (\dot{\mathbb{W}}_j)_{j \in \mathbb{N}} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}. \end{aligned} \quad (02.08)$$

We denote by $\mathbb{N}_{\boldsymbol{\theta}_\bullet}^{n,k}$ the joint distribution of the stochastic process $(\widehat{\boldsymbol{g}}_\bullet, \widehat{\boldsymbol{s}}_\bullet)$. \square

§02|04|01 Examples of diagonal inverse empirical mean models with noisy operator

§02|04.07 **Circular density deconvolution with unknown error density.** Similar to Model §01|04.12 consider the *complex* Hilbert spaces $\mathbb{L}_2(\lambda_{[0,1]})$ and $\mathbb{J} := \ell_2(\mathbb{Z})$. Let $\mathbb{D}_1 \subseteq \mathbb{L}_1(\lambda_{[0,1]})$ and $\mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]}) \subseteq \mathbb{L}_1(\lambda_{[0,1]})$ be sets of Lebesgue densities on $([0, 1], \mathcal{B}_{[0,1]})$ (by the usual embedding of real-valued functions as in **Notation** §01|02.10). We denote for each density $\mathfrak{p} \in \mathbb{L}_1(\lambda_{[0,1]})$ by $\mathbb{P}_{\mathfrak{p}} := \mathfrak{p}\lambda_{[0,1]} \in \mathcal{W}(\mathcal{B}_{[0,1]})$ the associated probability measure. Given a Lebesgue error density $\mathfrak{q} \in \mathbb{D}_1$ which is *not known* anymore for each Lebesgue density $\mathfrak{p} \in \mathbb{D}_2$ we consider the Lebesgue density $g = \mathfrak{q} \circledast \mathfrak{p} \in \mathbb{L}_2(\lambda_{[0,1]})$ (see **Notation** §01|04.11) and denote by $\mathbb{P}_{\mathfrak{p}|\mathfrak{q}} := (\mathfrak{q} \circledast \mathfrak{p})\lambda_{[0,1]} = g\lambda_{[0,1]} \in \mathcal{W}(\mathcal{B}_{[0,1]})$ the associated probability measure. We consider the statistical product experiment

$$([0, 1]^{n+k}, \mathcal{B}_{[0,1]}^{\otimes(n+k)}, \mathbb{P}_{\mathbb{D}_2 \times \mathbb{D}_1}^{n \otimes k} := (\mathbb{P}_{\mathfrak{p}|\mathfrak{q}}^{\otimes n} \otimes \mathbb{P}_{\mathfrak{q}}^{\otimes k})_{\mathfrak{p} \in \mathbb{D}_2, \mathfrak{q} \in \mathbb{D}_1}).$$

Let $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{[0,1]}), \ell_2(\mathbb{Z}))$ be the Fourier-series transform (see **Notation** §01|02.12). Evidently, for $g \in \mathbb{L}_2(\lambda_{[0,1]}) \subseteq \mathbb{L}_1(\lambda_{[0,1]})$ its Fourier-series $\mathfrak{g} = (g_j)_{j \in \mathbb{Z}} = Fg$ satisfies $g_j = \lambda_{[0,1]}(g\bar{e}_j) = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(\bar{e}_j)$ for each $j \in \mathbb{Z}$. Moreover, considering the Fourier-series $\mathfrak{p} = (\mathfrak{p}_j)_{j \in \mathbb{Z}} = F\mathfrak{p}$ of $\mathfrak{p} \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{[0,1]})$ by the *circular convolution theorem* we have $\mathfrak{g} = F(\mathfrak{q} \circledast \mathfrak{p}) = \mathfrak{q} \cdot \mathfrak{p}$ with $\mathfrak{q}_j = \lambda_{[0,1]}(\mathfrak{q}\bar{e}_j) = \mathbb{P}_{\mathfrak{q}}(\bar{e}_j) \in \ell_\infty(\mathbb{Z})$ and $\mathfrak{p}_j = F\mathfrak{p} \in \ell_2(\mathbb{Z})$ (see **Notation** §01|04.11). Moreover, the stochastic process $\bar{e}_\cdot = (\bar{e}_j)_{j \in \mathbb{Z}}$ on $([0, 1], \mathcal{B}_{[0,1]})$ is $(\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}})$ - \mathcal{B} -measurable, i.e. $\bar{e}_\cdot \in \mathcal{M}(\mathcal{B}_{[0,1]} \otimes 2^{\mathbb{Z}})$ for short (compare Model §01|04.12). We define $\hat{\mathfrak{g}}_\cdot = (\hat{\mathfrak{g}}_j := \hat{\mathbb{P}}_n(\bar{e}_j))_{j \in \mathbb{Z}} = \hat{\mathbb{P}}_n(\bar{e}_\cdot) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes n} \otimes 2^{\mathbb{Z}})$ and $\hat{\mathfrak{q}}_\cdot = (\hat{\mathfrak{q}}_j := \hat{\mathbb{P}}_k(\bar{e}_j))_{j \in \mathbb{Z}} = \hat{\mathbb{P}}_k(\bar{e}_\cdot) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes k} \otimes 2^{\mathbb{Z}})$ similar to an Empirical mean model §01|02.04 where by construction $\mathfrak{g}_j = \mathfrak{q}_j \mathfrak{p}_j = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(\bar{e}_j)$ is the $\ell_2(\mathbb{Z})$ -mean of $\hat{\mathfrak{g}}_\cdot$ and $\mathfrak{q}_j = \mathbb{P}_{\mathfrak{q}}(\bar{e}_j) \in \ell_\infty(\mathbb{Z})$ is the mean sequence of $\hat{\mathfrak{q}}_\cdot$. The observable noisy versions take the form $\hat{\mathfrak{g}}_\cdot = \mathfrak{q}_\cdot \mathfrak{p}_\cdot + n^{-1/2} \hat{\epsilon}_\cdot$ and $\hat{\mathfrak{q}}_\cdot = \mathfrak{q}_\cdot + k^{-1/2} \hat{\eta}_\cdot$, or in equal (02.06) with error processes

$$\begin{aligned} \hat{\epsilon}_\cdot &= n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}})(\bar{e}_\cdot) = n^{1/2}(\hat{\mathbb{P}}_n(\bar{e}_\cdot) - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(\bar{e}_\cdot)) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes n} \otimes 2^{\mathbb{Z}}) \quad \text{and} \\ \hat{\eta}_\cdot &= k^{1/2}(\hat{\mathbb{P}}_k - \mathbb{P}_{\mathfrak{q}})(\bar{e}_\cdot) = k^{1/2}(\hat{\mathbb{P}}_k(\bar{e}_\cdot) - \mathbb{P}_{\mathfrak{q}}(\bar{e}_\cdot)) \in \mathcal{M}(\mathcal{B}_{[0,1]}^{\otimes k} \otimes 2^{\mathbb{Z}}) \end{aligned}$$

satisfying Assumption §01|01.04. □

§02|04.08 **Density additive deconvolution on \mathbb{R} with unknown error density.** Similar to Model §01|04.15 consider the *complex* Hilbert space $\mathbb{L}_2 = \mathbb{L}_2(\lambda)$. Let $\mathbb{D}_1 \subseteq \mathbb{L}_1$ and $\mathbb{D}_2 \subseteq \mathbb{L}_2 \cap \mathbb{L}_1$ be sets of Lebesgue densities on $(\mathbb{R}, \mathcal{B})$ (by the usual embedding of real-valued functions as in **Notation** §01|02.10). We denote for each density $\mathfrak{p} \in \mathbb{L}_1$ by $\mathbb{P}_{\mathfrak{p}} := \mathfrak{p}\lambda \in \mathcal{W}(\mathcal{B})$ the associated probability measure. Given a Lebesgue density $\mathfrak{q} \in \mathbb{D}_1$ which is *not known* anymore for each Lebesgue density $\mathfrak{p} \in \mathbb{D}_2$ we consider the Lebesgue density $g = *_q \mathfrak{p} = \mathfrak{q} * \mathfrak{p} \in \mathbb{L}_2 \cap \mathbb{L}_1$ (see **Notation** §01|04.14) and denote by $\mathbb{P}_{\mathfrak{p}|\mathfrak{q}} := (\mathfrak{q} * \mathfrak{p})\lambda = g\lambda \in \mathcal{W}(\mathcal{B})$ the associated probability measure. We consider the statistical product experiment

$$(\mathbb{R}^{n+k}, \mathcal{B}^{\otimes(n+k)}, \mathbb{P}_{\mathbb{D}_2 \times \mathbb{D}_1}^{n \otimes k} := (\mathbb{P}_{\mathfrak{p}|\mathfrak{q}}^{\otimes n} \otimes \mathbb{P}_{\mathfrak{q}}^{\otimes k})_{\mathfrak{p} \in \mathbb{D}_2, \mathfrak{q} \in \mathbb{D}_1}).$$

Let $F \in \mathbb{L}(\mathbb{L}_2)$ be the Fourier-Plancherel transform (see **Notation** §01|02.15). Evidently, for $g \in \mathbb{L}_2 \cap \mathbb{L}_1$ its Fourier-Plancherel transform $\mathfrak{g} = (g_j)_{j \in \mathbb{R}} = Fg$ satisfies $g_j = \lambda(g\bar{e}_j) = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(\bar{e}_j)$ for all $j \in \mathbb{R}$. Moreover, considering the Fourier-Plancherel transform $\mathfrak{p} = (\mathfrak{p}_j)_{j \in \mathbb{R}} = F\mathfrak{p}$ of $\mathfrak{p} \in \mathbb{D}_2 \subseteq \mathbb{L}_2 \cap \mathbb{L}_1$ by the *additive convolution theorem* we have $\mathfrak{g}_j = F(*_q \mathfrak{p}) = \lambda(\mathfrak{q}\bar{e}_j)(F\mathfrak{p}) = \mathfrak{q}_j \mathfrak{p}_j$ λ -a.s. with $\mathfrak{q}_j = \lambda(\mathfrak{q}\bar{e}_j) = \mathbb{P}_{\mathfrak{q}}(\bar{e}_j) \in \mathbb{L}_\infty$ and $\mathfrak{p}_j = F\mathfrak{p} \in \mathbb{L}_2$ (see **Notation** §01|04.14). Moreover, the complex-valued stochastic process $\bar{e}_\cdot = (\bar{e}_j)_{j \in \mathbb{R}}$ on $(\mathbb{R}, \mathcal{B})$ is \mathcal{B}^2 - \mathcal{B} -measurable, i.e. $\bar{e}_\cdot \in \mathcal{M}(\mathcal{B}^2)$ for short (compare Model §01|04.15). We define $\hat{\mathfrak{g}}_\cdot = (\hat{\mathfrak{g}}_j := \hat{\mathbb{P}}_n(\bar{e}_j))_{j \in \mathbb{R}} = \hat{\mathbb{P}}_n(\bar{e}_\cdot) \in \mathcal{M}(\mathcal{B}^{\otimes n} \otimes \mathcal{B})$ and $\hat{\mathfrak{q}}_\cdot = (\hat{\mathfrak{q}}_j := \hat{\mathbb{P}}_k(\bar{e}_j))_{j \in \mathbb{R}} = \hat{\mathbb{P}}_k(\bar{e}_\cdot) \in \mathcal{M}(\mathcal{B}^{\otimes k} \otimes \mathcal{B})$ similar to an Empirical mean model §01|02.04

where by construction $g = \mathfrak{q} \cdot \mathfrak{p} = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(\bar{\mathfrak{e}})$ is the \mathbb{L}_2 -mean of \widehat{g} and $\mathfrak{q} \cdot = \mathbb{P}_{\mathfrak{q}}(\bar{\mathfrak{e}}) \in \mathbb{L}_\infty$ is the mean function of $\widehat{\mathfrak{q}} \cdot$. The observable noisy versions take the form $\widehat{g} = \mathfrak{q} \cdot \mathfrak{p} \cdot + n^{-1/2} \dot{\mathfrak{e}} \cdot$ and $\widehat{\mathfrak{q}} \cdot = \mathfrak{q} \cdot + k^{-1/2} \dot{\mathfrak{n}} \cdot$, or in equal (02.06) with error processes

$$\begin{aligned} \dot{\mathfrak{e}} \cdot &= n^{1/2}(\widehat{\mathbb{P}}_n - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}})(\bar{\mathfrak{e}}) = n^{1/2}(\widehat{\mathbb{P}}_n(\bar{\mathfrak{e}}) - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(\bar{\mathfrak{e}})) \in \mathcal{M}(\mathcal{B}^{\otimes n} \otimes \mathcal{B}) \quad \text{and} \\ \dot{\mathfrak{n}} \cdot &= k^{1/2}(\widehat{\mathbb{P}}_k - \mathbb{P}_{\mathfrak{q}})(\bar{\mathfrak{e}}) = k^{1/2}(\widehat{\mathbb{P}}_k(\bar{\mathfrak{e}}) - \mathbb{P}_{\mathfrak{q}}(\bar{\mathfrak{e}})) \in \mathcal{M}(\mathcal{B}^{\otimes k} \otimes \mathcal{B}) \end{aligned}$$

satisfying Assumption §01101.04. □

§02104.09 **Density multiplicative deconvolution on $\mathbb{R}_{>0}$ with unknown error density.** Consider the *complex* Hilbert spaces $\mathbb{L}_2(x^{2c-1}) = \mathbb{L}_2(\mathbb{R}_{>0}, \mathcal{B}_{>0}, x^{2c-1} \lambda_{>0})$ and $\mathbb{L}_2 = \mathbb{L}_2(\lambda)$ as in Model §01104.17. Let $\mathbb{D}_1 \subseteq \mathbb{L}_1(x^{c-1})$ and $\mathbb{D}_2 \subseteq \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ be sets of Lebesgue-densities on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ (by the usual embedding of real-valued functions as in **Notation** §01102.10). We denote for each Lebesgue density \mathfrak{p} on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ by $\text{colno}\mathbb{P}_{\mathfrak{p}} := \mathfrak{p} \lambda_{>0} \in \mathcal{W}(\mathcal{B}_{>0})$ the associated probability measure. Given a Lebesgue density $\mathfrak{q} \in \mathbb{L}_1(x^{c-1})$ which is *not known* anymore for each Lebesgue density $\mathfrak{p} \in \mathbb{D}_2$ we consider the Lebesgue density $g = \boxtimes_{\mathfrak{q}} \mathfrak{p} = \mathfrak{q} \boxtimes \mathfrak{p} \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ (see **Notation** §01104.16) and denote by $\mathbb{P}_{\mathfrak{p}|\mathfrak{q}} := (\mathfrak{q} \boxtimes \mathfrak{p}) \lambda_{>0} = g \lambda_{>0} \in \mathcal{W}(\mathcal{B}_{>0})$ the associated probability measure. We consider the statistical product experiment

$$(\mathbb{R}_{>0}^{n+k}, \mathcal{B}_{>0}^{\otimes(n+k)}, \mathbb{P}_{\mathbb{D}_2 \times \mathbb{D}_1}^{n \otimes k} := (\mathbb{P}_{\mathfrak{p}|\mathfrak{q}}^{\otimes n} \otimes \mathbb{P}_{\mathfrak{q}}^{\otimes k})_{\mathfrak{p} \in \mathbb{D}_2, \mathfrak{q} \in \mathbb{D}_1}).$$

Let $M_c \in \mathbb{L}(\mathbb{L}_2(x^{2c-1}), \mathbb{L}_2)$ be the Mellin transform (see **Notation** §01102.17). Evidently, for $g \in \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ its Mellin transform $g \cdot = (g_j)_{j \in \mathbb{R}} = M_c g$ satisfies $g_j = x^{c-1} \lambda_{>0}(\bar{x}_j g) = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(x^{c-1} \bar{x}_j)$ for all $j \in \mathbb{R}$. Moreover, considering the Mellin transform $\mathfrak{p} \cdot = (\mathfrak{p}_j)_{j \in \mathbb{R}} = M_c \mathfrak{p}$ of $\mathfrak{p} \in \mathbb{D}_2 \subseteq \mathbb{L}_1(x^{c-1}) \cap \mathbb{L}_2(x^{2c-1})$ by the *multiplicative convolution theorem* we have $g \cdot = M_c(\boxtimes_{\mathfrak{q}} \mathfrak{p}) = x^{c-1} \lambda_{>0}(\bar{x} \mathfrak{q})(M_c \mathfrak{p}) = \mathfrak{q} \cdot \mathfrak{p} \cdot$ λ -a.s. with $\mathfrak{q} \cdot = x^{c-1} \lambda_{>0}(\bar{x} \mathfrak{q}) = \mathbb{P}_{\mathfrak{q}}(x^{c-1} \bar{x}) \in \mathbb{L}_\infty$ and $\mathfrak{p} \cdot = M_c \mathfrak{p} \in \mathbb{L}_2$ (see **Notation** §01104.16). Moreover, the complex-valued stochastic process $x^{c-1} \bar{x} \cdot = (x^{c-1} \bar{x}_j)_{j \in \mathbb{R}} \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ on $(\mathbb{R}_{>0}, \mathcal{B}_{>0})$ is $\mathcal{B}_{>0} \otimes \mathcal{B}$ - \mathcal{B} -measurable, i.e. $x^{c-1} \bar{x} \cdot \in \mathcal{M}(\mathcal{B}_{>0} \otimes \mathcal{B})$ for short. We define $\widehat{g} \cdot = (\widehat{g}_j := \widehat{\mathbb{P}}_n(x^{c-1} \bar{x}_j))_{j \in \mathbb{R}} = \widehat{\mathbb{P}}_n(x^{c-1} \bar{x} \cdot) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes n} \otimes \mathcal{B})$ and $\widehat{\mathfrak{q}} \cdot = (\widehat{\mathfrak{q}}_j := \widehat{\mathbb{P}}_k(x^{c-1} \bar{x}_j))_{j \in \mathbb{R}} = \widehat{\mathbb{P}}_k(x^{c-1} \bar{x} \cdot) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes k} \otimes \mathcal{B})$ similar to an Empirical mean model §01102.04 where by construction $g = \mathfrak{q} \cdot \mathfrak{p} \cdot = \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(x^{c-1} \bar{x} \cdot)$ is the \mathbb{L}_2 -mean of $\widehat{g} \cdot$ and $\mathfrak{q} \cdot = \mathbb{P}_{\mathfrak{q}}(x^{c-1} \bar{x} \cdot) \in \mathbb{L}_\infty$ is the mean function of $\widehat{\mathfrak{q}} \cdot$. The observable noisy versions take the form $\widehat{g} \cdot = \mathfrak{q} \cdot \mathfrak{p} \cdot + n^{-1/2} \dot{\mathfrak{e}} \cdot$ and $\widehat{\mathfrak{q}} \cdot = \mathfrak{q} \cdot + k^{-1/2} \dot{\mathfrak{n}} \cdot$, or in equal (02.06) with error processes

$$\begin{aligned} \dot{\mathfrak{e}} \cdot &= n^{1/2}(\widehat{\mathbb{P}}_n - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}})(x^{c-1} \bar{x} \cdot) = n^{1/2}(\widehat{\mathbb{P}}_n(x^{c-1} \bar{x} \cdot) - \mathbb{P}_{\mathfrak{p}|\mathfrak{q}}(x^{c-1} \bar{x} \cdot)) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes n} \otimes \mathcal{B}) \quad \text{and} \\ \dot{\mathfrak{n}} \cdot &= k^{1/2}(\widehat{\mathbb{P}}_k - \mathbb{P}_{\mathfrak{q}})(x^{c-1} \bar{x} \cdot) = k^{1/2}(\widehat{\mathbb{P}}_k(x^{c-1} \bar{x} \cdot) - \mathbb{P}_{\mathfrak{q}}(x^{c-1} \bar{x} \cdot)) \in \mathcal{M}(\mathcal{B}_{>0}^{\otimes k} \otimes \mathcal{B}) \end{aligned}$$

satisfying Assumption §01101.04. □

§02104.10 **Functional linear regression under second order stationarity.** Consider the *complex* Hilbert spaces $\mathbb{L}_2(\lambda_{(0,1)})$ and $\mathbb{J} := \ell_2(\mathbb{Z})$. Let $(\mathbb{L}_2(\lambda_{(0,1)}), \langle \cdot, \cdot \rangle_{\mathbb{L}_2(\lambda_{(0,1)})})$ be equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{L}_2(\lambda_{(0,1)})}$ and let (X, Y) be an $\mathbb{L}_2(\lambda_{(0,1)}) \times \mathbb{R}$ -valued random vector with joint distribution $\mathbb{P}^{X,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{(0,1)})} \otimes \mathcal{B})$. We denote by $\mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{(0,1)})})$ the marginal distribution of the *real random function* X (by the usual embedding of real-valued functions as in **Notation** §01102.10). We tactically identify X and Y with the coordinate map $\Pi_{\mathbb{L}_2(\lambda_{(0,1)})}$ and $\Pi_{\mathbb{R}}$, respectively, and thus (X, Y) with the identity $\text{id}_{\mathbb{L}_2(\lambda_{(0,1)}) \times \mathbb{R}}$ such that $\mathbb{P} = \mathbb{P}^{X,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{(0,1)})} \otimes \mathcal{B})$. Here and subsequently, we assume that $Y, \|X\|_{\mathbb{L}_2(\lambda_{(0,1)})}^2 \in \mathbb{L}_2(\mathbb{P}) = \mathbb{L}_2(\mathbb{L}_2(\lambda_{(0,1)}) \times \mathbb{R}, \mathcal{B}_{\mathbb{L}_2(\lambda_{(0,1)})} \otimes \mathcal{B}, \mathbb{P})$, $\mathbb{P}(\langle x, X \rangle_{\mathbb{L}_2(\lambda_{(0,1)})} = 0$ for all $x \in \mathbb{L}_2(\lambda_{(0,1)})$, and that X is *second order stationary* as in Model §02103.04. In this situation X admits a *covariance operator* $\Gamma^X \in \mathbb{L}^{\mathbb{Z}}(\mathbb{L}_2(\lambda_{(0,1)}))$ which equals a *circular additive convolution*, that is $\otimes_{c^X} = \Gamma^X \in \mathbb{L}^{\mathbb{Z}}(\mathbb{L}_2(\lambda_{(0,1)}))$ with $c^X \in \mathbb{L}_2(\lambda_{(0,1)})$ (see Model §02103.04) and there is $g \in \mathbb{L}_2(\lambda_{(0,1)})$ satisfying $\langle g, x \rangle_{\mathbb{L}_2(\lambda_{(0,1)})} =$

$\mathbb{P}(Y\langle X, x \rangle_{\mathbb{L}_2(\lambda_{0,1})})$ for all $x \in \mathbb{L}_2(\lambda_{0,1})$. In what follows we assume that in addition $g \in \text{ran}(\Gamma^X) \subseteq \mathbb{L}_2(\lambda_{0,1})$. In this situation there exists $f \in \mathbb{L}_2(\lambda_{0,1})$ such that

$$\langle g, x \rangle_{\mathbb{L}_2(\lambda_{0,1})} = \mathbb{P}(Y\langle X, x \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = \mathbb{P}(\langle f, X \rangle_{\mathbb{L}_2(\lambda_{0,1})} \langle X, x \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = \langle \Gamma^X f, x \rangle_{\mathbb{L}_2(\lambda_{0,1})} \quad \forall x \in \mathbb{L}_2(\lambda_{0,1})$$

or in equal \mathbb{P} -a.s. we have $Y = \langle f, X \rangle_{\mathbb{L}_2(\lambda_{0,1})} + \xi$ with $\mathbb{P}(\xi \langle X, x \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = 0$ for all $x \in \mathbb{L}_2(\lambda_{0,1})$. Let us denote by $\mathbb{P}_{c^X} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})})$ the marginal distribution of X which is not fully specified given $c^X \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$ (see Model §02103.04). Moreover, for $c^X \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$ and $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$, and hence $g := \otimes_{c^X} f \in \mathbb{L}_2(\lambda_{0,1})$, we denote by $\mathbb{P}_{f|c^X}$ the joint distribution of (X, Y) without fully specifying the distribution which however is assumed to satisfy $\mathbb{P}_{f|c^X}(Y\langle X, x \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = \mathbb{P}_{c^X}(\langle f, X \rangle_{\mathbb{L}_2(\lambda_{0,1})} \langle X, x \rangle_{\mathbb{L}_2(\lambda_{0,1})})$ for all $x \in \mathbb{L}_2(\lambda_{0,1})$. Let $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{0,1}), \ell_2(\mathbb{Z}))$ be the *Fourier-series transform* with $h \mapsto Fh := \underline{h}_\bullet = \lambda_{0,1}(h\bar{e}_\bullet)$ and exponential basis $e_\bullet := (e_j)_{j \in \mathbb{Z}}$ (see Notations §01102.10 and §01102.12). Following Model §02103.04 $M_{c^X} = F \otimes_{c^X} F^* = F \Gamma^X F^* \in \mathbb{L}(\ell_2(\mathbb{Z}))$ is a multiplication operator with $c_\bullet^X \in \ell_2(\mathbb{Z}) \subseteq \ell_\infty(\mathbb{Z})$ satisfying $c_\bullet^X = (c_j^X = \mathbb{P}_{c^X}(|\langle X, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})}|^2))_{j \in \mathbb{Z}}$. Therefore the complex-valued stochastic process $|\lambda_{0,1}(X\bar{e}_\bullet)|^2 = (|\lambda_{0,1}(X\bar{e}_j)|^2 = |\langle X, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})}|^2)_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})} \otimes 2^{\mathbb{Z}})$ fulfils Assumption §01101.04 and $c_\bullet^X = \mathbb{P}_{c^X}(|\lambda_{0,1}(X\bar{e}_\bullet)|^2)$. Moreover, for each $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$ the Fourier coefficients $\underline{g}_\bullet = (g_j)_{j \in \mathbb{Z}} = Fg = F \Gamma^X F^* F f = M_{c^X} \underline{f}_\bullet = c_\bullet^X \underline{f}_\bullet$, satisfy $\underline{g}_\bullet = (g_j = \mathbb{P}_{f|c^X}(Y\langle X, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})}))_{j \in \mathbb{Z}}$. The complex-valued stochastic process $Y\lambda_{0,1}(X\bar{e}_\bullet) = (Y\lambda_{0,1}(X\bar{e}_j) = Y\langle X, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})})_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})} \otimes \mathcal{B}) \otimes 2^{\mathbb{Z}}$ fulfils Assumption §01101.04 and $\underline{g}_\bullet = c_\bullet^X \underline{f}_\bullet = \mathbb{P}_{f|c^X}(Y\lambda_{0,1}(X\bar{e}_\bullet))$. The observable noisy versions take the form $\hat{\underline{g}}_\bullet = c_\bullet^X \underline{f}_\bullet + n^{-1/2} \dot{\underline{\epsilon}}_\bullet$ and $\hat{c}_\bullet^X = c_\bullet^X + k^{-1/2} \dot{\underline{\eta}}_\bullet$, or in equal (02.06) with error processes

$$\dot{\underline{\epsilon}}_\bullet = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{f|c^X})(|\lambda_{0,1}(X\bar{e}_\bullet)|^2) \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{Z}} \quad \text{and}$$

$$\dot{\underline{\eta}}_\bullet = k^{1/2}(\hat{\mathbb{P}}_k - \mathbb{P}_{c^X})(Y\lambda_{0,1}(X\bar{e}_\bullet)) \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^{\otimes k} \otimes 2^{\mathbb{Z}})$$

satisfying Assumption §01101.04. □

§02104.11 **Functional linear instrumental regression under second order stationarity.** Consider the complex Hilbert spaces $\mathbb{L}_2(\lambda_{0,1})$ and $\mathbb{J} := \ell_2(\mathbb{Z})$. Let $(\mathbb{L}_2(\lambda_{0,1}), \langle \cdot, \cdot \rangle_{\mathbb{L}_2(\lambda_{0,1})})$ be equipped with its Borel- σ -algebra $\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}$ and let (Z, X, Y) be an $\mathbb{L}_2(\lambda_{0,1})^2 \times \mathbb{R}$ -valued random vector with joint distribution $\mathbb{P}^{Z,X,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^2 \otimes \mathcal{B})$. We denote by $\mathbb{P}^Z, \mathbb{P}^X \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})})$ the marginal distributions of the *real random functions* Z and X (by the usual embedding of real-valued functions as in Notation §01102.10). Moreover, denote by $\mathbb{P}^{Z,X} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^2)$, and $\mathbb{P}^{Z,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})} \otimes \mathcal{B})$ the marginal distribution of (Z, X) and (Z, Y) , respectively. We tactically take Z, X and Y as coordinate maps and thus identify (Z, X, Y) with the identity $\text{id}_{\mathbb{L}_2(\lambda_{0,1})^2 \times \mathbb{R}}$ such that $\mathbb{P} = \mathbb{P}^{Z,X,Y} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^2 \otimes \mathcal{B})$. Here and subsequently, given $\mathbb{L}_2(\mathbb{P}) := \mathbb{L}_2((\mathbb{L}_2(\lambda_{0,1})^2 \times \mathbb{R}, \mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^2 \otimes \mathcal{B}, \mathbb{P})$ we assume that $Y, \|Z\|_{\mathbb{L}_2(\lambda_{0,1})}^2, \|X\|_{\mathbb{L}_2(\lambda_{0,1})}^2 \in \mathbb{L}_2(\mathbb{P})$, $\mathbb{P}(\langle Z, z \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = 0$, $\mathbb{P}(\langle x, X \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = 0$ for all $z, x \in \mathbb{L}_2(\lambda_{0,1})$, and that (Z, X) is *second order stationary* as in Model §02103.05. In this situation (Z, X) admits a *cross-covariance operator* $\Gamma^{ZX} \in \mathbb{L}(\mathbb{L}_2(\lambda_{0,1}))$ which equals a *circular additive convolution*, that is $\otimes_{c^{ZX}} = \Gamma^{ZX} \in \mathbb{L}(\mathbb{L}_2(\lambda_{0,1}))$ with $c^{ZX} \in \mathbb{L}_2(\lambda_{0,1})$ (see Model §02101.09) and there is $g \in \mathbb{L}_2(\lambda_{0,1})$ satisfying $\langle g, z \rangle_{\mathbb{L}_2(\lambda_{0,1})} = \mathbb{P}(Y\langle Z, z \rangle_{\mathbb{L}_2(\lambda_{0,1})})$ for all $z \in \mathbb{L}_2(\lambda_{0,1})$. In what follows we assume that in addition $g \in \text{ran}(\Gamma^{ZX}) \subseteq \mathbb{L}_2(\lambda_{0,1})$. In this situation there exists $f \in \mathbb{L}_2(\lambda_{0,1})$ such that

$$\langle g, z \rangle_{\mathbb{L}_2(\lambda_{0,1})} = \mathbb{P}(Y\langle Z, z \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = \mathbb{P}(\langle f, X \rangle_{\mathbb{L}_2(\lambda_{0,1})} \langle Z, z \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = \langle \Gamma^{ZX} f, z \rangle_{\mathbb{L}_2(\lambda_{0,1})} \quad \forall z \in \mathbb{L}_2(\lambda_{0,1})$$

or in equal \mathbb{P} -a.s. we have $Y = \langle f, X \rangle_{\mathbb{L}_2(\lambda_{0,1})} + \xi$ with $\mathbb{P}(\xi \langle Z, z \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = 0$ for all $z \in \mathbb{L}_2(\lambda_{0,1})$. Let us denote by $\mathbb{P}_{c^{ZX}} \in \mathcal{W}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^2)$ the marginal distribution of (Z, X) which is not fully specified given $c^{ZX} \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$ (see Model §02103.05). Moreover, for $c^{ZX} \in \mathbb{D}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$ and $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$, and hence $g := \otimes_{c^{ZX}} f \in \mathbb{L}_2(\lambda_{0,1})$, we denote by $\mathbb{P}_{f|c^{ZX}}$ the joint distribution of (Z, X, Y) without fully specifying the distribution which however is assumed to satisfy

$\mathbb{P}_{f|c^{z,x}}(Y \langle Z, z \rangle_{\mathbb{L}_2(\lambda_{0,1})}) = \mathbb{P}_{c^{z,x}}(\langle f, X \rangle_{\mathbb{L}_2(\lambda_{0,1})} \langle Z, z \rangle_{\mathbb{L}_2(\lambda_{0,1})})$ for all $z \in \mathbb{L}_2(\lambda_{0,1})$. Let $F \in \mathbb{L}(\mathbb{L}_2(\lambda_{0,1}), \ell_2(\mathbb{Z}))$ be the *Fourier-series transform* with $h \mapsto Fh := h_\bullet = \lambda_{0,1}(h\bar{e}_\bullet)$ and exponential basis $e_\bullet := (e_j)_{j \in \mathbb{Z}}$ (see *Notations* §01102.10 and §01102.12). Similar to Model §02103.05 $M_{c^{z,x}} = F \otimes_{c^{z,x}} F^* = F \Gamma^{Z,X} F^* \in \mathbb{L}(\ell_2(\mathbb{Z}))$ is a multiplication operator with $c_\bullet^{z,x} \in \ell_2(\mathbb{Z}) \subseteq \ell_\infty(\mathbb{Z})$ which satisfies $c_\bullet^{z,x} = (c_j^{z,x} = \mathbb{P}_{c^{z,x}}(\langle Z, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})} \langle e_j, X \rangle_{\mathbb{L}_2(\lambda_{0,1})}))_{j \in \mathbb{Z}}$. Therefore the complex-valued stochastic process $\lambda_{0,1}(Z\bar{e}_\bullet)\lambda_{0,1}(\bar{X}e_\bullet) = (\lambda_{0,1}(Z\bar{e}_j)\lambda_{0,1}(\bar{X}e_j) = \langle Z, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})} \langle e_j, X \rangle_{\mathbb{L}_2(\lambda_{0,1})})_{j \in \mathbb{Z}} \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})} \otimes 2^{\mathbb{Z}})$ fulfils Assumption §01101.04 and $c_\bullet^{z,x} = \mathbb{P}_{c^{z,x}}(\lambda_{0,1}(Z\bar{e}_\bullet)\lambda_{0,1}(\bar{X}e_\bullet))$. Moreover, for each $f \in \mathbb{F}_2 \subseteq \mathbb{L}_2(\lambda_{0,1})$ the Fourier coefficients $g = (g_j)_{j \in \mathbb{Z}} = Fg = F \Gamma^{Z,X} F^* F f = M_{c^{z,x}} f = c_\bullet^{z,x} f$, satisfy $g = (g_j = \mathbb{P}_{f|c^{z,x}}(Y \langle Z, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})}))_{j \in \mathbb{Z}}$. The complex-valued stochastic process $Y \lambda_{0,1}(Z\bar{e}_\bullet) = (Y \lambda_{0,1}(Z\bar{e}_j) = Y \langle Z, e_j \rangle_{\mathbb{L}_2(\lambda_{0,1})})_{j \in \mathbb{Z}} \in \mathcal{M}((\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})} \otimes \mathcal{B}) \otimes 2^{\mathbb{Z}})$ fulfils Assumption §01101.04 and $g = c_\bullet^{z,x} f = \mathbb{P}_{f|c^{z,x}}(Y \lambda_{0,1}(Z\bar{e}_\bullet))$. The observable noisy versions take the form $\hat{g}_\bullet = c_\bullet^{z,x} f_\bullet + n^{-1/2} \dot{\epsilon}_\bullet$ and $\hat{c}_\bullet^{z,x} = c_\bullet^{z,x} + k^{-1/2} \dot{\eta}_\bullet$, or in equal (02.06) with error processes

$$\begin{aligned} \dot{\epsilon}_\bullet &= n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{f|c^{z,x}})(\lambda_{0,1}(Z\bar{e}_\bullet)\lambda_{0,1}(\bar{X}e_\bullet)) \in \mathcal{M}((\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})} \otimes \mathcal{B})^{\otimes n} \otimes 2^{\mathbb{Z}}) \quad \text{and} \\ \dot{\eta}_\bullet &= k^{1/2}(\hat{\mathbb{P}}_k - \mathbb{P}_{c^x})(Y \lambda_{0,1}(Z\bar{e}_\bullet)) \in \mathcal{M}(\mathcal{B}_{\mathbb{L}_2(\lambda_{0,1})}^{\otimes k} \otimes 2^{\mathbb{Z}}) \end{aligned}$$

satisfying Assumption §01101.04. □

Chapter 2

Regularisation of inverse problems

Given $g = T_{\alpha}\theta$ the regularised reconstruction of θ in a direct problem and an inverse problem with linear operator T_{α} in a diagonal or general case is presented.

Overview

§03	Ill-posed inverse problems	39
§04	Regularisation by orthogonal projection	41
§04 01	Weighted norms and inner products	42
§04 02	Direct problem	42
§04 02 01	Global and maximal global \mathfrak{v} -error	43
§04 02 02	Local and maximal local ϕ -error	44
§04 03	Diagonal inverse problem	45
§04 03 01	Global and maximal global \mathfrak{v} -error	45
§04 03 02	Local and maximal local ϕ -error	46
§05	(Generalised) linear Galerkin approach	47
§05 01	Linear Galerkin approach	47
§05 01 01	Global and maximal global \mathfrak{v} -error	50
§05 01 02	Global and maximal global ϕ -error	51
§05 02	Generalised linear Galerkin approach	52
§05 02 01	Global and maximal global \mathfrak{v} -error	53
§05 02 02	Global and maximal global ϕ -error	53
§06	Spectral regularisation	54
§06 01	(Generalised) Tikhonov regularisation	55
§06 02	Spectral regularisation	57
§06 02 01	Maximal global \mathfrak{v} -error	57
§06 02 02	Maximal local ϕ -error	60

§03 Ill-posed inverse problems

Let \mathbb{H} and \mathbb{G} be separable Hilbert spaces over $\mathbb{K} \in \{\mathbb{R}, \mathbb{C}\}$ endowed with inner product $\langle \cdot, \cdot \rangle_{\mathbb{H}}$ and $\langle \cdot, \cdot \rangle_{\mathbb{G}}$ and induced norm $\|\cdot\|_{\mathbb{H}}$ and $\|\cdot\|_{\mathbb{G}}$, respectively. Consider a linear bounded operator $T : \mathbb{H} \rightarrow \mathbb{G}$, for short $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$.

§03|00.01 **Definition.** Given $g \in \mathbb{G}$ the reconstruction of a solution $\theta \in \mathbb{H}$ of the equation $g = T\theta$ is called *inverse problem*. □

§03|00.02 **Definition** (Hadamard [1932]). An inverse problem $g = T\theta$ is called *well-posed* if (a) a solution θ *exists*, (b) the solution θ is *unique*, and (c) the solution depends continuously on g . An inverse problem which is not well-posed is called *ill-posed*. □

For a broader overview on inverse problems we refer the reader to the monograph by Kress [1989] or Engl et al. [2000].

§03100.03 **Property (Existence and identification).** *There exists a unique solution of the equation $g = T\theta$ if and only if the following two conditions are satisfied*

(existence) g belongs to the range $\text{ran}(T)$ of T ,

(identification) The operator T is injective, i.e., its null space $\ker(T) = \{0\}$ is trivial. \square

§03100.04 **Remark.** If there does not exist a solution typically one might consider a least-square solution which exists if and only if $g \in \text{ran}(T) \oplus \ker(T^*)$. A least-square solution with minimal norm, if it exists, could be recovered, in case the solution is not unique. Nevertheless, the main issue is often the stability of the inverse problem. More precisely, if the solution does not depend continuously on g , i.e., the inverse T^{-1} of T is not continuous, a reconstruction $\hat{\theta} = T^{-1}\hat{g}$ may be far from the solution θ even if the noisy version \hat{g} is close to g . \square

§03100.05 **Property.** Denote by $\Pi_{\overline{\text{ran}(T)}}$ the orthogonal projection onto the closure $\overline{\text{ran}(T)}$ of the range of T . For each $g \in \mathbb{G}$ the following assertions are equivalent (i) θ minimises $h \mapsto \|g - Th\|_{\mathbb{G}}$ over \mathbb{H} (least square solution); (ii) $\Pi_{\overline{\text{ran}(T)}}g = T\theta$; (iii) $T^*g = T^*T\theta$ (normal equation). \square

§03100.06 **Remark.** We note that $g \in \text{ran}(T) \oplus \text{ran}(T)^\perp$ implies $\Pi_{\overline{\text{ran}(T)}}g \in \text{ran}(T)$ and hence the preimage $T^{-1}(\{\Pi_{\overline{\text{ran}(T)}}g\}) = \{h \in \mathbb{H} : Th = \Pi_{\overline{\text{ran}(T)}}g\}$ is not empty. More precisely, due to **Property** §03100.05 it equals the *set of least square solutions*, i.e. $T^{-1}(\{\Pi_{\overline{\text{ran}(T)}}g\}) = \{\theta \in \mathbb{H} : T^*g = T^*T\theta\}$. \square

§03100.07 **Notation.** In the sequel keep in mind that for each $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ its restriction $T_{\text{res}} : \ker(T)^\perp \rightarrow \text{ran}(T)$ is bijective and thus has an inverse $T_{\text{res}}^{-1} : \text{ran}(T) \rightarrow \ker(T)^\perp$. Here and subsequently we identify T and T_{res} . \square

§03100.08 **Definition.** For $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ the *Moore-Penrose inverse* (generalised or pseudo inverse) T^\dagger is the unique linear extension of $T^{-1} : \text{ran}(T) \rightarrow \ker(T)^\perp$ to the domain $\text{dom}(T^\dagger) := \text{ran}(T) \oplus \text{ran}(T)^\perp$ with $\ker(T^\dagger) = \text{ran}(T)^\perp$ satisfying $T^\dagger g := T^{-1}\Pi_{\overline{\text{ran}(T)}}g$ for any $g \in \text{dom}(T^\dagger)$. \square

§03100.09 **Remark.** We note that $TT^\dagger T = T$, $T^\dagger TT^\dagger = T^\dagger$, $T^\dagger T = \Pi_{\ker(T)^\perp}$ and $TT^\dagger g = \Pi_{\overline{\text{ran}(T)}}g$ for any $g \in \text{dom}(T^\dagger)$. If T is injective, and hence T^*T , then $T^*T : \mathbb{H} \rightarrow \text{ran}(T^*T)$ is invertible, which in turn, for any $g \in \text{ran}(T) \oplus \text{ran}(T)^\perp$, implies that $(T^*T)^\dagger T^*g$ is the unique solution of the normal equation, and thus $T^{-1}(\{\Pi_{\overline{\text{ran}(T)}}g\}) = \{T^\dagger g\} = \{(T^*T)^\dagger T^*g\}$. If T is invertible then $T^\dagger = T^{-1}$. \square

§03100.10 **Property.** For each $g \in \text{dom}(T^\dagger)$, $T^\dagger g$ belongs to $T_{\text{res}}^{-1}(\{\Pi_{\overline{\text{ran}(T)}}g\})$ and, hence is a least square solution. Moreover, $T^\dagger g$ is the unique least square solution with minimal $\|\cdot\|_{\mathbb{H}}$ -norm, that is, $\|T^\dagger g\|_{\mathbb{H}} = \inf\{\|h\|_{\mathbb{H}} : h \in T_{\text{res}}^{-1}(\{\Pi_{\overline{\text{ran}(T)}}g\})\}$. \square

We eventually approximate the operator T by sequence $(T^m)_{m \in \mathbb{N}}$ of operators in $\mathbb{L}(\mathbb{H}, \mathbb{G})$, where for each m the operator $T^m \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ has a finite dimensional image. If $\|T^m - T\|_{\mathbb{L}(\mathbb{H}, \mathbb{G})} = o(1)$ as $m \rightarrow \infty$, then T is compact (reference), i.e. $T \in \mathbb{K}(\mathbb{H}, \mathbb{G})$ for short, and the inverse problem is generally ill-posed due to the next property.

§03100.11 **Property.** If \mathbb{H} and \mathbb{G} are infinite dimensional and $T \in \mathbb{K}(\mathbb{H}, \mathbb{G})$ is injective, then

$$\inf\{\|Th\|_{\mathbb{G}} : \|h\|_{\mathbb{H}} = 1, h \in \mathbb{H}\} = 0,$$

which in turn implies that $T_{\text{res}}^{-1} : \text{ran}(T) \rightarrow \mathbb{H}$, and hence also T^\dagger , is not continuous. \square

Coming back to the reconstruction of $\theta \in \mathbb{H}$ from a noisy version of the image $g = T\theta \in \mathbb{G}$ and eventually a noisy version of the operator $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ as introduced in **Chapter 1**. Throughout this manuscript introducing the measure space $(\mathcal{J}, \mathcal{J}, \nu)$ with index set \mathcal{J} being contained in \mathbb{N} , \mathbb{Z} or \mathbb{R} , surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{L}_2(\nu))$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{L}_2(\nu))$, which are fixed and presumed to be known in advance, we write $g = A\theta$ with $A := VT \in \mathbb{L}(\mathbb{H}, \mathbb{L}_2(\nu))$, $VTU^* \in \mathbb{L}(\mathbb{L}_2(\nu))$, $g = (g_j)_{j \in \mathcal{J}} := Vg \in \mathbb{L}_2(\nu)$ and $\theta = (\theta_j)_{j \in \mathcal{J}} := U\theta \in \mathbb{L}_2(\nu)$. Concerning the operator VTU^* we distinguish two cases, in **Section §04** it behaves like a multiplication, i.e. $VTU^* = M_{\mathfrak{s}} \in \mathbb{L}(\mathbb{J})$ for some $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$, while in **Section §05** and **Section §06** we consider the non-diagonal case $VTU^* = T_{\cdot, \cdot} \in \mathbb{L}(\ell_2)$.

§04 Regularisation by orthogonal projection

§04100.01 **Notation (Reminder)**. Consider the measure space $(\mathcal{J}, \mathcal{J}, \nu)$ and the Hilbert space $\mathbb{J} = \mathbb{L}_2(\nu)$ as in **Notation §01101.01**. For $w \in \mathbb{R}^{\mathcal{J}}$ define the multiplication map $M_w : \mathbb{R}^{\mathcal{J}} \rightarrow \mathbb{R}^{\mathcal{J}}$ with $a \mapsto M_w a := w \cdot a := (w_j a_j)_{j \in \mathcal{J}}$. If $w \in \mathcal{M}(\mathcal{J})$, i.e. w is \mathcal{J} - \mathcal{B} -measurable, then we have $M_w : \mathcal{M}(\mathcal{J}) \rightarrow \mathcal{M}(\mathcal{J})$ too. If in addition $w \in \mathbb{L}_\infty(\nu)$ then we have also $M_w \in \mathbb{L}(\mathbb{J})$ identifying again equivalence classes and representatives. We set $\mathbb{L}(\mathbb{J}) := \{M_w : w \in \mathbb{L}_\infty(\nu)\} \subseteq \mathbb{L}(\mathbb{J})$ noting that $\|M_w\|_{\mathbb{L}(\mathbb{J})} = \sup \{\|w \cdot a\|_{\mathbb{J}} : \|a\|_{\mathbb{J}} \leq 1\} \leq \|w\|_{\mathbb{L}_\infty(\nu)}$ for each $M_w \in \mathbb{L}(\mathbb{J})$ (see **Notation §01104.01**). Finally, given surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ we define $\mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J})) := V^*(\mathbb{L}(\mathbb{J}))U := \{V^* M_w U \in \mathbb{L}(\mathbb{H}, \mathbb{G}) : M_w \in \mathbb{L}(\mathbb{J})\}$. As a consequence, for each $T \in \mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J}))$ we have $VTU^* = M_w \in \mathbb{L}(\mathbb{J})$ for some $w \in \mathbb{L}_\infty(\nu)$. \square

§04100.02 **Notation (Reminder (see §01104.02))**. For $A \in \mathcal{J}$ we denote by $\mathbb{1}_A = (\mathbb{1}_j^A)_{j \in \mathcal{J}}$ the indicator function where for each $j \in \mathcal{J}$, $\mathbb{1}_j^A = 1$ if $j \in A$ and $\mathbb{1}_j^A = 0$ otherwise. Obviously, $\mathbb{1}_A$ is \mathcal{J} - \mathcal{B} -measurable, i.e. $\mathbb{1}_A \in \mathcal{M}(\mathcal{J})$, and it belongs to $\mathbb{L}_\infty(\nu)$, and to $\mathbb{L}_2(\nu)$ whenever $\nu(A) \in \mathbb{R}_{\geq 0}$. Since $\{j\} \in \mathcal{J}$ we have $\mathbb{1}^{\{j\}} \in \mathcal{J}$ and $\mathbb{1}^{\{j\}} \in \mathbb{L}_\infty(\nu)$. Obviously, we have $\mathbb{1} = \mathbb{1}^{\mathcal{J}} \in \mathbb{L}_\infty(\nu)$ and $M_{\mathbb{1}} \in \mathbb{L}(\mathbb{J})$. For each $w \in \mathbb{L}_\infty(\nu)$ set $\mathbb{J}w := \{\{a \cdot w\}_\nu : a \in \mathbb{L}_2(\nu)\} = \{a \cdot w : a \in \mathbb{J} = \mathbb{L}_2(\nu)\}$ and hence in particular $\mathbb{J}\mathbb{1}_A = \{a \cdot \mathbb{1}_A : a \in \mathbb{J}\}$. Given $0 = (0)_{j \in \mathcal{J}}$ for $w \in \mathcal{M}(\mathcal{J})$ we write further $\mathcal{N}_w := \{w = 0\} := \{j \in \mathcal{J} : w_j = 0\} \in \mathcal{J}$, and denote by $\text{dom}(M_w) = \{a \in \mathbb{J} : a \cdot w \in \mathbb{J}\}$, $\text{ran}(M_w) = \{a \cdot w : a \in \text{dom}(M_w) \subseteq \mathbb{J}\}$ and $\text{ker}(M_w) = \{a \in \mathbb{J} : \{a \cdot w\}_\nu = 0\}$, respectively, the domain, range and nullspace of $M_w : \mathbb{J} \supseteq \text{dom}(M_w) \rightarrow \mathbb{J}$. We write $w \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$, if $w \in \mathcal{M}(\mathcal{J})$ and $\nu(\mathcal{N}_w) = 0$. Similarly, for $w \in \mathcal{M}(\mathcal{J})$ with $\nu(\{w \leq 0\}) = 0$ we write $w \in \mathcal{M}_{> 0, \nu}(\mathcal{J})$. For $w \in \mathcal{M}(\mathcal{J})$ we denote its Moore-Penrose inverse by $w^\dagger := w^{-1} \mathbb{1}_{\mathcal{N}_w^c} \in \mathcal{M}(\mathcal{J})$ meaning $w_j^\dagger := w_j^{-1}$ if $j \in \mathcal{N}_w^c$ and $w_j^\dagger := 0$ if $j \in \mathcal{N}_w$. Obviously, we have $w^\dagger w \cdot w^\dagger = w^\dagger$, $w \cdot w^\dagger w = w$ and $w \cdot w^\dagger = w^\dagger w = \mathbb{1}_{\mathcal{N}_w^c}$. \square

§04100.03 **Property**. For each $w \in \mathcal{L}_\infty(\nu)$ the multiplication $M_w \in \mathbb{L}(\mathbb{J})$ is a linear bounded operator. Keeping $\mathcal{N}_w = \{w = 0\} \in \mathcal{J}$ in mind its range and null space is given by $\text{ran}(M_w) = \mathbb{J}w$ and $\text{ker}(M_w) = \mathbb{J}\mathbb{1}_{\mathcal{N}_w^c} = \text{ran}(M_{\mathbb{1}_{\mathcal{N}_w^c}})$, respectively. $M_w \in \mathbb{L}(\mathbb{J})$ is consequently *injective* if and only if $w \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$, i.e. $w \in \mathcal{M}(\mathcal{J})$ and $\nu(\mathcal{N}_w) = 0$. If in addition $w \in \mathcal{M}_{> 0, \nu}(\mathcal{J}) \cap \mathcal{L}_\infty(\nu)$ then the multiplication $M_w \in \mathbb{L}(\mathbb{J}) \subseteq \mathbb{L}(\mathbb{J})$ is a positive semi-definite operator, which is *injective* if and only if $w \in \mathcal{M}_{> 0, \nu}(\mathcal{J})$. For each $A \in \mathcal{J}$ setting $A^c := \mathcal{J} \setminus A \in \mathcal{J}$ the range and null space of the multiplication $M_{\mathbb{1}_A} \in \mathbb{L}(\mathbb{J}) \subseteq \mathbb{L}(\mathbb{J})$ is given by $\text{ran}(M_{\mathbb{1}_A}) = \mathbb{J}\mathbb{1}_A$ and $\text{ker}(M_{\mathbb{1}_A}) = \mathbb{J}\mathbb{1}_{A^c}$, respectively. Obviously, we have $M_{\mathbb{1}_A}^2 = M_{\mathbb{1}_A}$ and hence $M_{\mathbb{1}_A}$ is an *orthogonal projection* and $\mathbb{J} = \mathbb{J}\mathbb{1}_A \oplus \mathbb{J}\mathbb{1}_{A^c}$. Moreover, the map $M_{\mathbb{1}} = \text{id}_{\mathbb{J}}$ equals the identity on \mathbb{J} . \square

§04100.04 **Assumption**. For $\mathbb{J} = \mathbb{L}_2(\nu)$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$, fixed and presumed to be *known* in advance, let $T \in \mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J})) \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$ and hence $VTU^* = M_{\mathfrak{s}} \in \mathbb{L}(\mathbb{J})$ for some $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$, let $g \in \text{dom}(M_{\mathfrak{s}})$, and hence $\mathfrak{s}^\dagger g \in \mathbb{J} = \mathbb{L}_2(\nu)$. \square

In the sequel we consider first in **Subsection** §04|02 the direct problem, that is $\mathfrak{s} = \mathbf{1}$, and then secondly in **Subsection** §04|03 the diagonal inverse problem, that is $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$.

§04|01 Weighted norms and inner products

§04|01.01 **Notation (Reminder (see §01|00.01)).** Extending the real line by the points $-\infty$ and $+\infty$ we define $\overline{\mathbb{R}} := \mathbb{R} \cup \{\pm\infty\}$ and denote by $\overline{\mathcal{B}}$ the Borel- σ -field over $\overline{\mathbb{R}}$ where the trace of $\overline{\mathcal{B}}_{\mathbb{R}} = \overline{\mathcal{B}} \cap \mathbb{R}$ over \mathbb{R} equals \mathcal{B} . Thereby, each $a_\cdot \in \mathcal{M}(\mathcal{J})$ is in a canonical way also \mathcal{J} - $\overline{\mathcal{B}}$ measurable, $a_\cdot \in \overline{\mathcal{M}}(\mathcal{J})$ for short. For $w_\cdot \in \overline{\mathcal{M}}(\mathcal{J})$ and hence $w_\cdot^2 \in \overline{\mathcal{M}}_{\geq 0}(\mathcal{J})$, consider the measure $w_\cdot^2 \nu$ on $(\mathcal{J}, \mathcal{J})$, i.e., $w_\cdot^2 = dw_\cdot^2 \nu / d\nu$ is the Radon-Nikodym density of $w_\cdot^2 \nu$ with respect to ν . We write shortly $\langle \cdot, \cdot \rangle_w := \langle \cdot, \cdot \rangle_{\mathbb{L}_2(w_\cdot^2 \nu)}$ and $\|\cdot\|_w := \|\cdot\|_{\mathbb{L}_2(w_\cdot^2 \nu)}$. For $w_\cdot \in \mathcal{M}(\mathcal{J})$ with Moore-Penrose inverse $w_\cdot^\dagger := w_\cdot^{-1} \mathbf{1}_\cdot^{\mathcal{N}_w} \in \mathcal{M}(\mathcal{J})$ we set $\mathbb{J}^w := \mathbb{L}_2(\nu) := \text{dom}(M_{w_\cdot})$ and write $w_\cdot^{2\dagger} := (w_\cdot^\dagger)^2 = (w_\cdot^2)^\dagger$ for short. \square

§04|01.02 **Property.** Let $w_\cdot \in \mathcal{M}(\mathcal{J})$. Then $w_\cdot^2 \nu \in \mathcal{M}_\sigma(\mathcal{J})$ is a σ -finite measure satisfying $w_\cdot^2 \nu(|a_\cdot|^2) = \nu(|w_\cdot a_\cdot|^2)$ for each $a_\cdot \in \mathcal{L}_2(w_\cdot^2 \nu)$, and $\mathbb{L}_2(w_\cdot^2 \nu)$ endowed with inner product $\langle \cdot, \cdot \rangle_w = \langle \cdot, \cdot \rangle_{\mathbb{L}_2(w_\cdot^2 \nu)} = \langle M_{w_\cdot} \cdot, M_{w_\cdot} \cdot \rangle_{\mathbb{L}_2(\nu)}$ is a separable Hilbert space. If in addition $w_\cdot \in \mathcal{M}(\mathcal{J}) \cap \mathcal{L}_\infty(\nu)$, then

$$\mathcal{L}_2(w_\cdot^{2\dagger} \nu) = \mathcal{L}_2(\nu) w_\cdot + \overline{\mathcal{M}}(\mathcal{J}) \mathbf{1}_\cdot^{\mathcal{N}_w} = \{w_\cdot h_\cdot : h_\cdot \in \mathcal{L}_2(\nu)\} + \{h_\cdot \mathbf{1}_\cdot^{\mathcal{N}_w} : h_\cdot \in \overline{\mathcal{M}}(\mathcal{J})\}. \quad (04.01)$$

Indeed, for each $h_\cdot \in \overline{\mathcal{M}}(\mathcal{J})$ consider the decomposition $h_\cdot = w_\cdot w_\cdot^\dagger h_\cdot + h_\cdot \mathbf{1}_\cdot^{\mathcal{N}_w}$. The claim follows immediately from the equivalence of $h_\cdot \in \mathcal{L}_2(w_\cdot^{2\dagger} \nu)$ and $w_\cdot^\dagger h_\cdot \in \mathcal{L}_2(\nu)$. Since $w_\cdot \in \mathcal{L}_\infty(\nu)$ the map $M_{w_\cdot} : \mathcal{L}_2(\nu) \rightarrow \mathcal{L}_2(\nu)$ is well-defined, and (similar to (04.01))

$$\text{dom}(M_{w_\cdot}) = \{h_\cdot \in \mathcal{L}_2(\nu) : w_\cdot^\dagger h_\cdot \in \mathcal{L}_2(\nu)\} = \mathcal{L}_2(\nu) w_\cdot + \mathcal{L}_2(\nu) \mathbf{1}_\cdot^{\mathcal{N}_w} \subseteq \mathcal{L}_2(w_\cdot^{2\dagger} \nu).$$

Consequently, if in addition $\mathcal{N}_w = \emptyset$, then $\text{dom}(M_{w_\cdot}) = \mathcal{L}_2(w_\cdot^{2\dagger} \nu)$. If $w_\cdot \in \mathbb{L}_\infty(\nu)$ then $M_{w_\cdot} \in \mathbb{L}(\mathbb{J})$, and (with Moore-Penrose inverse w_\cdot^\dagger of a representative $w_\cdot \in \mathcal{M}(\mathcal{J})$) $M_{w_\cdot^\dagger} : \mathbb{J} \supseteq \text{dom}(M_{w_\cdot}) \rightarrow \mathbb{J}$. Moreover, we have $\text{dom}(M_{w_\cdot}) = \mathbb{J}$, $\text{ran}(M_{w_\cdot}) = \mathbb{J} w_\cdot$, and $\ker(M_{w_\cdot}) = \mathbb{J} \mathbf{1}_\cdot^{\mathcal{N}_w}$ (see **Property** §04|00.03). Therewith, it follows $\text{dom}(M_{w_\cdot}) = \mathbb{J} w_\cdot \oplus \mathbb{J} \mathbf{1}_\cdot^{\mathcal{N}_w}$. Consequently, if in addition $\nu(\mathcal{N}_w) = 0$, then $\mathbb{J}^w = \mathbb{L}_2(\nu) = \text{dom}(M_{w_\cdot}) = \mathbb{J} w_\cdot = \mathbb{L}_2(w_\cdot^{2\dagger} \nu)$. The last equality follows from (04.01) since both measures $w_\cdot^{2\dagger} \nu$ and ν share the same null sets (i.e. they mutually dominate each other). \square

§04|02 Direct problem

We assume throughout this subsection that Assumption §04|00.04 is satisfied with $\mathfrak{s} = \mathbf{1} \in \mathbb{L}_\infty(\nu)$.

§04|02.01 **Notation.** For a non-empty and generally non-finite subset \mathcal{J} of \mathbb{N} , \mathbb{Z} or \mathbb{R} and $m \in \mathbb{N}$ we set $\llbracket m \rrbracket := [-m, m] \cap \mathcal{J}$ and assuming $\llbracket m \rrbracket \in \mathcal{J}$ we write shortly $\mathbf{1}_\cdot^m = (\mathbf{1}_j^m)_{j \in \mathcal{J}} := \mathbf{1}_\cdot^{\llbracket m \rrbracket} \in \mathcal{M}(\mathcal{J})$. Furthermore, we define $\mathbf{1}_\cdot^{m\perp} := \mathbf{1}_\cdot - \mathbf{1}_\cdot^m \in \mathcal{M}(\mathcal{J})$. \square

§04|02.02 **Property.** For each $m \in \mathbb{N}$, $M_{\mathbf{1}_\cdot^m} \in \mathbb{L}(\mathbb{J})$ and $M_{\mathbf{1}_\cdot^{m\perp}} \in \mathbb{L}(\mathbb{J})$ is the *orthogonal projection* onto the linear subspace $\mathbb{J} \mathbf{1}_\cdot^m \subseteq \mathbb{J}$ and its orthogonal complement $\mathbb{J} \mathbf{1}_\cdot^{m\perp} = (\mathbb{J} \mathbf{1}_\cdot^m)^\perp \subseteq \mathbb{J}$, respectively, that is $\mathbb{J} = \mathbb{J} \mathbf{1}_\cdot^m \oplus \mathbb{J} \mathbf{1}_\cdot^{m\perp}$. We have point-wise $\mathbf{1}_\cdot^m - \mathbf{1}_\cdot = o(1)$ as $m \rightarrow \infty$ meaning that for each $j \in \mathcal{J}$ holds $\mathbf{1}_j^m - \mathbf{1}_j = o(1)$ as $m \rightarrow \infty$. Considering the orthogonal projection $M_{\mathbf{1}_\cdot^m} \in \mathbb{L}(\mathbb{J})$ and the identity $M_{\mathbf{1}_\cdot} = \text{id}_{\mathbb{J}} \in \mathbb{L}(\mathbb{J})$ point-wise convergence $M_{\mathbf{1}_\cdot^m} - \text{id}_{\mathbb{J}} = o(1)$ as $m \rightarrow \infty$ holds too, that is, $\|(M_{\mathbf{1}_\cdot^m} - \text{id}_{\mathbb{J}})a_\cdot\|_{\mathbb{J}} = \|(\mathbf{1}_\cdot^m - \mathbf{1}_\cdot)a_\cdot\|_{\mathbb{J}} = \|\mathbf{1}_\cdot^{m\perp} a_\cdot\|_{\mathbb{J}} = o(1)$ as $m \rightarrow \infty$ for all $a_\cdot \in \mathbb{J}$. \square

§04|02.03 **Orthogonal projection.** Given $m \in \mathbb{N}$ we define for each $\theta = U\theta \in \mathbb{J}$ its orthogonal projection $\theta^m := \theta \mathbf{1}_\cdot^m \in \mathbb{J} \mathbf{1}_\cdot^m$ (and $\theta^m := U^* \theta^m \in \mathbb{H}$). \square

§04|02|01 Global and maximal global \mathfrak{v} -error

We shall measure first globally the accuracy of the orthogonal projection $\theta^m := \theta \mathbb{1}_m^m$ of $\theta \in \mathbb{J}$.

§04|02.04 **Property.** If $\mathfrak{v}_\bullet \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ (i.e. $\mathfrak{v}_\bullet \in \mathcal{M}(\mathcal{J})$ and $\nu(\mathcal{N}_0) = 0$) and $\theta \in \mathbb{L}_2(\mathfrak{v}_\bullet^2 \nu)$ (i.e. $\|\theta\|_{\mathfrak{v}_\bullet}^2 = \mathfrak{v}_\bullet^2 \nu(\theta^2) \in \mathbb{R}_{>0}$), then for each $m \in \mathbb{N}$ we have $\theta^m \in \mathbb{L}_2(\mathfrak{v}_\bullet^2 \nu)$ too, since $\|\theta^m\|_{\mathfrak{v}_\bullet}^2 = \mathfrak{v}_\bullet^2 \nu(\theta^2 \mathbb{1}_m) \leq \mathfrak{v}_\bullet^2 \nu(\theta^2)$. Moreover, it holds $\|\theta^m - \theta\|_{\mathfrak{v}_\bullet}^2 = \|\theta \mathbb{1}_m^{\perp} - \theta\|_{\mathfrak{v}_\bullet}^2 = \mathfrak{v}_\bullet^2 \nu(\theta^2 \mathbb{1}_m^{\perp}) \leq \mathfrak{v}_\bullet^2 \nu(\theta^2) \in \mathbb{R}_{>0}$ and $\|\theta^m - \theta\|_{\mathfrak{v}_\bullet}^2 = o(1)$ as $m \rightarrow \infty$ by dominated convergence. \square

§04|02.05 **Comment.** We assume throughout this chapter that the Hilbert space $\mathbb{J} = \mathbb{L}_2(\mathcal{J}, \mathcal{J}, \nu)$ and the surjective partial isometry $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ is fixed and known in advance. Considering a \mathfrak{v} -error means the weight sequences $\mathfrak{v}_\bullet \in \mathcal{M}(\mathcal{J})$ is also fixed and known in advance. Consequently, the condition $\mathfrak{v}_\bullet \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ does not impose an additional restriction. \square

§04|02.06 **Global \mathfrak{v} -error.** Given $\mathfrak{v}_\bullet \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$, $m \in \mathbb{N}$, a solution $\theta = U\theta \in \mathbb{L}_2(\mathfrak{v}_\bullet^2 \nu)$ and its orthogonal projection $\theta^m = \theta \mathbb{1}_m^m \in \mathbb{J} \mathbb{1}_m^m$ we call $\|\theta^m - \theta\|_{\mathfrak{v}_\bullet} = \|\theta \mathbb{1}_m^{\perp}\|_{\mathfrak{v}_\bullet} \in \mathbb{R}_{>0}$ *global \mathfrak{v} -error*. \square

§04|02.07 **Assumption.** Consider weights $\mathfrak{a}_\bullet, \mathfrak{v}_\bullet \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ (i.e. $\mathfrak{a}_\bullet, \mathfrak{v}_\bullet \in \mathcal{M}(\mathcal{J})$ and $\nu(\mathcal{N}_0) = 0 = \nu(\mathcal{N}_0)$), such that $\mathfrak{a}_\bullet \in \mathbb{L}_\infty(\nu)$ and $(\mathfrak{a}\mathfrak{v})_\bullet := (\mathfrak{a}_j \mathfrak{v}_j)_{j \in \mathcal{J}} = \mathfrak{a}_\bullet \mathfrak{v}_\bullet \in \mathbb{L}_\infty(\nu)$. We write $(\mathfrak{a}\mathfrak{v})_{(m)} := \|(\mathfrak{a}\mathfrak{v})_\bullet \mathbb{1}_m^{\perp}\|_{\mathbb{L}_\infty(\nu)} \in \mathbb{R}_{>0}$ for each $m \in \mathbb{N}$. \square

§04|02.08 **Notation.** For sequences $a_n, b_n \in (\mathbb{K})^{\mathbb{N}}$ taking its values in $\mathbb{K} \in \{\mathbb{R}, \mathbb{R}_{>0}, \mathbb{R}_{\geq 0}, \mathbb{Q}, \mathbb{Z}, \dots\}$ we write $a_\bullet \in (\mathbb{K})_{\nearrow}^{\mathbb{N}}$ and $b_\bullet \in (\mathbb{K})_{\searrow}^{\mathbb{N}}$ if a_\bullet and b_\bullet , respectively, is monotonically *non-decreasing* and *non-increasing*. If in addition $a_n \rightarrow \infty$ and $b_n \rightarrow 0$ as $n \rightarrow \infty$, then we write $a_\bullet \in (\mathbb{K})_{\nearrow \infty}^{\mathbb{N}}$ and $b_\bullet \in (\mathbb{K})_{\searrow 0}^{\mathbb{N}}$ for short. For $w_\bullet \in \mathbb{L}_\infty(\nu)$ we set $w_{(0)} := \|w_\bullet\|_{\mathbb{L}_\infty(\nu)}$ and $w_{(s)} = (w_{(j)} := \|w_\bullet \mathbb{1}_j^{\perp}\|_{\mathbb{L}_\infty(\nu)})_{j \in \mathbb{N}}$, where by construction $w_{(s)} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$. \square

§04|02.09 **Reminder.** Under Assumption §04|02.07 we have $\mathbb{J}^{\mathfrak{a}} = \mathbb{L}_2(\nu) = \text{dom}(M_{\mathfrak{a}}) = \mathbb{J} \mathfrak{a}_\bullet = \mathbb{L}_2(\mathfrak{a}^{2\ddagger} \nu)$ and the three measures ν , $\mathfrak{a}^{2\ddagger} \nu$ and $\mathfrak{v}_\bullet^2 \nu$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04|01.02). Consequently, $\mathbb{J}^{\mathfrak{a}} \subseteq \mathbb{J} = \mathbb{L}_2(\nu)$ and if $h_\bullet \in \mathbb{L}_2(\mathfrak{a}^{2\ddagger} \nu)$ satisfies $\mathfrak{v}_\bullet^2 \nu(h_\bullet^2) \in \mathbb{R}_{>0}$, for example, then $h_\bullet \in \mathbb{L}_2(\mathfrak{v}_\bullet^2 \nu)$ too. \square

§04|02.10 **Notation.** Under Assumption §04|02.07 and given a constant $r \in \mathbb{R}_{>0}$ we consider $\mathbb{J}^{\mathfrak{a}} = \mathbb{L}_2(\nu) = \mathbb{L}_2(\mathfrak{a}^{2\ddagger} \nu)$ endowed with $\|\cdot\|_{\mathfrak{a}^\ddagger} := \|\cdot\|_{\mathbb{J}^{\mathfrak{a}}} := \|\cdot\|_{\mathbb{L}_2(\mathfrak{a}^{2\ddagger} \nu)}$ and the ellipsoid

$$\mathbb{J}^{\mathfrak{a}, r} := \{h_\bullet \in \mathbb{J}^{\mathfrak{a}}: \|h_\bullet\|_{\mathfrak{a}^\ddagger}^2 = \mathfrak{a}^{2\ddagger} \nu(h_\bullet^2) = \nu(\mathfrak{a}^{2\ddagger} h_\bullet^2) \leq r^2\} \subseteq \mathbb{J}^{\mathfrak{a}}. \quad \square$$

§04|02.11 **Property.** Under Assumption §04|02.07 we have $\mathbb{J}^{\mathfrak{a}} \subseteq \mathbb{L}_2(\mathfrak{v}_\bullet^2 \nu)$. Indeed, for each $h_\bullet \in \mathbb{J}^{\mathfrak{a}}$ (i.e., $\|h_\bullet\|_{\mathfrak{a}^\ddagger} \in \mathbb{R}_{>0}$) follows $\|h_\bullet\|_{\mathfrak{v}_\bullet}^2 = \nu(h_\bullet^2 \mathfrak{a}^{2\ddagger} (\mathfrak{a}\mathfrak{v})_\bullet^2) \leq \|h_\bullet\|_{\mathfrak{a}^\ddagger}^2 \|(\mathfrak{a}\mathfrak{v})_\bullet\|_{\mathbb{L}_\infty(\nu)}^2 \in \mathbb{R}_{>0}$. \square

§04|02.12 **Abstract smoothness condition.** Under Assumption §04|02.07 a solution $\theta \in \mathbb{J}$ satisfies an *abstract smoothness condition* if there is $r \in \mathbb{R}_{>0}$ such that $\theta \in \mathbb{J}^{\mathfrak{a}, r} \subseteq \mathbb{J}^{\mathfrak{a}}$. \square

§04|02.13 **Lemma.** Under Assumption §04|02.07 for each $m \in \mathbb{N}$ and solution $\theta \in \mathbb{J}^{\mathfrak{a}, r} \subseteq \mathbb{L}_2(\mathfrak{v}_\bullet^2 \nu)$ its orthogonal projection $\theta^m := \theta \mathbb{1}_m^m \in \mathbb{J} \mathbb{1}_m^m$ satisfies $\|\theta^m - \theta\|_{\mathfrak{v}_\bullet} = \|\theta \mathbb{1}_m^{\perp}\|_{\mathfrak{v}_\bullet} \leq r (\mathfrak{a}\mathfrak{v})_{(m)}$. \square

§04|02.14 **Proof of Lemma** §04|02.13. Given in the lecture. \square

§04|02.15 **Maximal global \mathfrak{v} -error.** Under Assumption §04|02.07 for $m \in \mathbb{N}$, a solution $\theta = U\theta \in \mathbb{J}^{\mathfrak{a}, r}$ and its orthogonal projection $\theta^m = \theta \mathbb{1}_m^m \in \mathbb{J} \mathbb{1}_m^m$ we call $\sup \{\|\theta^m - \theta\|_{\mathfrak{v}_\bullet}: \theta \in \mathbb{J}^{\mathfrak{a}, r}\}$ *maximal global \mathfrak{v} -error* over the class of solutions $\mathbb{J}^{\mathfrak{a}, r}$. \square

§04|02|02 Local and maximal local ϕ -error

Secondly, we measure locally the accuracy of the orthogonal projection $\theta^m := \theta \mathbb{1}^m$ of $\theta \in \mathbb{J}$.

§04|02.16 **Notation.** For $\phi \in \mathcal{M}(\mathcal{J})$ and $\text{dom}(\phi\nu) := \{h_\bullet \in \mathbb{J} = \mathbb{L}_2(\nu) : \phi h_\bullet \in \mathbb{L}_1(\nu)\}$ we consider the linear functional $\phi\nu : \mathbb{J} \supseteq \text{dom}(\phi\nu) \rightarrow \mathbb{R}$ given by $h_\bullet \mapsto \phi\nu(h_\bullet) := \nu(\phi h_\bullet)$ with a slight abuse of notations. \square

§04|02.17 **Comment.** If $\phi \in \mathbb{J} = \mathbb{L}_2(\nu)$, then it follows $\text{dom}(\phi\nu) = \mathbb{J}$ and $\|\phi\nu\|_{\mathbb{L}(\mathbb{J}, \mathbb{R})} = \|\phi\|_{\mathbb{J}} \in \mathbb{R}_{\geq 0}$. Consequently, we have $\phi\nu \in \mathbb{L}(\mathbb{J}, \mathbb{R})$ and $\phi\nu(h_\bullet) = \langle h_\bullet, \phi \rangle_{\mathbb{J}}$, in other words ϕ is a Fréchet-Riesz representative of the continuous linear functional $\phi\nu$. \square

§04|02.18 **Property.** If $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ (i.e. $\phi \in \mathcal{M}(\mathcal{J})$ and $\nu(\mathcal{N}_\phi) = 0$) and $\theta \in \text{dom}(\phi\nu)$ (i.e. $\theta\phi \in \mathbb{L}_1(\nu)$), then for each $m \in \mathbb{N}$ we have $\theta^m \in \text{dom}(\phi\nu)$ too, since $\|\phi\theta^m\|_{\mathbb{L}_1(\nu)} = \nu(|\phi\theta|\mathbb{1}^m) \leq \nu(|\phi\theta|)$. Moreover, it holds

$$|\phi\nu(\theta) - \phi\nu(\theta^m)| \leq |\phi|\nu(|\theta^m - \theta|) = |\phi|\nu(|\theta|\mathbb{1}^{m\perp}) \leq \nu(|\phi\theta|) \in \mathbb{R}_{\geq 0}$$

and $|\phi\nu(\theta) - \phi\nu(\theta^m)| = o(1)$ as $m \rightarrow \infty$ by dominated convergence. \square

§04|02.19 **Comment.** We assume throughout this chapter that the Hilbert space $\mathbb{J} = \mathbb{L}_2(\mathcal{J}, \mathcal{J}, \nu)$ and the surjective partial isometry $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ is fixed and known in advance (Assumption §04|00.04). Considering a ϕ -error means the linear function $\phi\nu$ and hence in equal $\phi \in \mathcal{J}$ is also fixed and known in advance. Consequently, the condition $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ does not impose an additional restriction. \square

§04|02.20 **Local ϕ -error.** Given $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$, $m \in \mathbb{N}$, a solution $\theta = U\theta \in \text{dom}(\phi\nu)$ and its orthogonal projection $\theta^m = \theta\mathbb{1}^m \in \mathbb{J}\mathbb{1}^m$ we call $|\phi\nu(\theta) - \phi\nu(\theta^m)| = |\phi\nu(\theta\mathbb{1}^{m\perp})| \in \mathbb{R}_{\geq 0}$ *local ϕ -error*. \square

§04|02.21 **Assumption.** Consider $\phi, \alpha \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ (i.e. $\phi, \alpha \in \mathcal{M}(\mathcal{J})$ and $\nu(\mathcal{N}_\phi) = 0 = \nu(\mathcal{N}_\alpha)$), such that $\alpha \in \mathbb{L}_\infty(\nu)$ and $(\alpha\phi)_\bullet := (\alpha_j\phi_j)_{j \in \mathcal{J}} = \alpha_\bullet\phi_\bullet \in \mathbb{L}_2(\nu)$ and hence $\|\alpha_\bullet\mathbb{1}^{m\perp}\|_\phi = \|(\alpha\phi)_\bullet\mathbb{1}^{m\perp}\|_{\mathbb{L}_2(\nu)} = o(1)$ as $m \rightarrow \infty$. \square

§04|02.22 **Reminder.** Under Assumption §04|02.21 we have $\mathbb{J}^\alpha = \mathbb{L}_2(\nu) = \text{dom}(M_\alpha) = \mathbb{J}\alpha_\bullet = \mathbb{L}_2(\alpha^{2\uparrow}\nu)$ and the three measures ν , $|\phi|\nu$ and $\alpha^{2\uparrow}\nu$ dominate mutually each other (see **Property** §04|01.02). Consequently, $\mathbb{J}^\alpha \subseteq \mathbb{J} = \mathbb{L}_2(\nu)$ and if $h_\bullet \in \mathbb{L}_2(\alpha^{2\uparrow}\nu)$ satisfies $\nu(|\phi h_\bullet|) \in \mathbb{R}_{\geq 0}$, for example, then $h_\bullet \in \mathbb{L}_1(|\phi|\nu)$ too. \square

§04|02.23 **Property.** Under Assumption §04|02.21 we have $\mathbb{J}^\alpha \subseteq \text{dom}(\phi\nu)$. Indeed, for each $h_\bullet \in \mathbb{J}^\alpha$, i.e. $\|h_\bullet\|_{\alpha^\dagger} \in \mathbb{R}_{\geq 0}$, we have $\|\phi h_\bullet\|_{\mathbb{L}_1(\nu)} = \nu(|h_\bullet\alpha^\dagger(\alpha\phi)_\bullet|) \leq \|h_\bullet\|_{\alpha^\dagger} \|(\alpha\phi)_\bullet\|_{\mathbb{L}_2(\nu)} \in \mathbb{R}_{\geq 0}$. \square

§04|02.24 **Notation (Reminder).** Under Assumption §04|02.21 a solution $\theta = U\theta \in \mathbb{J}$ satisfies an abstract smoothness condition if there is $r \in \mathbb{R}_{> 0}$ such that $\theta \in \mathbb{J}^{\alpha, r} = \{h_\bullet \in \mathbb{J}^\alpha : \|h_\bullet\|_{\alpha^\dagger}^2 \leq r^2\} \subseteq \mathbb{J}^\alpha$ where $\|\cdot\|_{\alpha^\dagger} = \|\cdot\|_{\mathbb{J}^\alpha} := \|\cdot\|_{\mathbb{L}_2(\alpha^{2\uparrow}\nu)}$ (see **Definition** §04|02.12). Since $(\alpha\phi)_\bullet \in \mathbb{L}_2(\nu)$ we have $\|\alpha_\bullet\mathbb{1}^{m\perp}\|_\phi = \|(\alpha\phi)_\bullet\mathbb{1}^{m\perp}\|_{\mathbb{L}_2(\nu)} = o(1)$ as $m \rightarrow \infty$ by dominated convergence. \square

§04|02.25 **Lemma.** Under Assumption §04|02.21 for each $m \in \mathbb{N}$ and $\theta \in \mathbb{J}^{\alpha, r} \subseteq \text{dom}(\phi\nu)$ its orthogonal projection $\theta^m := \theta\mathbb{1}^m \in \mathbb{J}\mathbb{1}^m$ of satisfies $|\phi\nu(\theta) - \phi\nu(\theta^m)| = |\phi\nu(\theta\mathbb{1}^{m\perp})| \leq \nu(|\phi\theta|\mathbb{1}^{m\perp}) \leq r \|\alpha_\bullet\mathbb{1}^{m\perp}\|_\phi$.

§04|02.26 **Proof of Lemma** §04|02.25. Given in the lecture. \square

§04|02.27 **Maximal local ϕ -error.** Under Assumption §04|02.21 for $m \in \mathbb{N}$, a solution $\theta = U\theta \in \mathbb{J}^{\alpha, r}$ and its orthogonal projection $\theta^m = \theta\mathbb{1}^m \in \mathbb{J}\mathbb{1}^m$ we call $\sup \{|\phi\nu(\theta) - \phi\nu(\theta^m)| : \theta \in \mathbb{J}^{\alpha, r}\}$ *maximal local ϕ -error* over the class of solutions $\mathbb{J}^{\alpha, r}$. \square

§04|03 Diagonal inverse problem

We assume throughout this subsection that Assumption §04|00.04 is satisfied with $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$.

§04|03.01 **Reminder.** Under Assumption §04|00.04 we consider $T \in \mathbb{L}^{\mathbb{U},\mathbb{V}}(\mathbb{L}^{\mathbb{U}}(\mathbb{J})) \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$, and hence $VTU^* = M_{\mathfrak{s}} \in \mathbb{L}^{\mathbb{U}}(\mathbb{J})$ and $g = M_{\mathfrak{s}}\theta = \mathfrak{s}\theta \in \mathbb{J}$ for some $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$. Due to **Property** §04|01.02 the Moore-Penrose inverse of $M_{\mathfrak{s}} \in \mathbb{L}^{\mathbb{U}}(\mathbb{J})$ satisfies $M_{\mathfrak{s}}^\dagger = M_{\mathfrak{s}^\dagger} : \mathbb{J} \supseteq \text{dom}(M_{\mathfrak{s}}) \rightarrow \mathbb{J}$ with $\text{dom}(M_{\mathfrak{s}}) = \mathbb{J}\mathfrak{s} \oplus \mathbb{J}\mathbb{1}_{\mathfrak{s}}^{\mathcal{N}} = \mathbb{J}^{\mathfrak{s}}$. For each $m \in \mathbb{N}$, $M_{\mathbb{1}^m} \in \mathbb{L}^{\mathbb{U}}(\mathbb{J})$ and $M_{\mathbb{1}^m}^\perp \in \mathbb{L}^{\mathbb{U}}(\mathbb{J})$ is the *orthogonal projection* onto the linear subspace $\mathbb{J}\mathbb{1}^m \subseteq \mathbb{J}$ and its orthogonal complement $\mathbb{J}\mathbb{1}^{m\perp} = (\mathbb{J}\mathbb{1}^m)^\perp \subseteq \mathbb{J}$, respectively, that is $\mathbb{J} = \mathbb{J}\mathbb{1}^m \oplus \mathbb{J}\mathbb{1}^{m\perp}$ (see **Property** §04|02.02). Given $g \in \mathbb{J}$ we call $\theta \in \mathbb{J}$ satisfying $\|g - \mathfrak{s}\theta\|_{\mathbb{J}} = \inf \{\|g - \mathfrak{s}h\|_{\mathbb{J}} : h \in \mathbb{J}\}$ a least squares solution, if it exists (see **Property** §03|00.05). \square

§04|03.02 **Property.** For $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ and each $g \in \text{dom}(M_{\mathfrak{s}}) = \mathbb{J}\mathfrak{s} \oplus \mathbb{J}\mathbb{1}_{\mathfrak{s}}^{\mathcal{N}}$ is $\theta = M_{\mathfrak{s}}g = \mathfrak{s}^\dagger g$ the unique least square solution with minimal $\|\cdot\|_{\mathbb{J}}$ -norm in the set $\mathfrak{s}^\dagger g + \mathbb{J}\mathbb{1}_{\mathfrak{s}}^{\mathcal{N}}$ of all least square solutions. If in addition $\nu(\mathcal{N}_{\mathfrak{s}}) = 0$ (i.e. $M_{\mathfrak{s}}$ is injective), then $\theta = \mathfrak{s}^\dagger g$ is the *unique* least square solution. Given $m \in \mathbb{N}$ for each $g \in \text{dom}(M_{\mathfrak{s}})$ we have $g\mathbb{1}^m \in \text{dom}(M_{\mathfrak{s}})$ too. In particular, for $\theta = \mathfrak{s}^\dagger g$ it follows $\theta\mathbb{1}^m = (\mathfrak{s}^\dagger g)\mathbb{1}^m = \mathfrak{s}^\dagger(g\mathbb{1}^m) \in \mathbb{J}\mathbb{1}^m$. \square

§04|03.03 **Orthogonal projection.** Given $m \in \mathbb{N}$ we define for each $g \in \text{dom}(M_{\mathfrak{s}})$ and $\theta = \mathfrak{s}^\dagger g \in \mathbb{J}$ the orthogonal projections $g^m = g\mathbb{1}^m \in \mathbb{J}\mathbb{1}^m$ and $\theta^m = (\mathfrak{s}^\dagger g)\mathbb{1}^m = \mathfrak{s}^\dagger g^m \in \mathbb{J}\mathbb{1}^m$. \square

§04|03.04 **Assumption.** Consider weights $\mathfrak{a}, \mathfrak{t} \in \mathcal{M}_{>0,\nu}(\mathcal{J}) \cap \mathcal{L}_\infty(\nu)$ and hence $(\mathfrak{t}\mathfrak{a}) := \mathfrak{t}\mathfrak{a} \in \mathcal{L}_\infty(\nu)$. \square

§04|03.05 **Link condition.** Given weights $\mathfrak{t} \in \mathcal{M}_{>0,\nu}(\mathcal{J}) \cap \mathcal{L}_\infty(\nu)$, an operator $M_{\mathfrak{s}} \in \mathbb{L}^{\mathbb{U}}(\mathbb{J})$, and hence $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$, satisfies a *link condition* if there is $d \in \mathbb{R}_{\geq 1}$ such that

$$M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t},d} := \{M_{\mathfrak{w}} \in \mathbb{L}^{\mathbb{U}}(\mathbb{J}) : |\mathfrak{w}_j| \leq d\mathfrak{t}_j \wedge \mathfrak{t}_j \leq d|\mathfrak{w}_j| \nu\text{-a.e.}\}. \quad \square$$

§04|03.06 **Property.** Given weights $\mathfrak{t} \in \mathcal{M}_{>0,\nu}(\mathcal{J}) \cap \mathcal{L}_\infty(\nu)$ and introducing $\|\cdot\|_{\mathfrak{t}} := \|M_{\mathfrak{t}}\cdot\|_{\mathbb{J}}$ we have

$$\mathbb{M}_{\mathfrak{t},d} = \{M \in \mathbb{L}^{\mathbb{U}}(\mathbb{J}) : d^{-1}\|h\|_{\mathfrak{t}} \leq \|Mh\|_{\mathbb{J}} \leq d\|h\|_{\mathfrak{t}}, \forall h \in \mathbb{J}\}$$

moreover, for each $M \in \mathbb{M}_{\mathfrak{t},d}$ and for all $s \in \mathbb{R}$ (exploiting $M_{\mathfrak{t}}^s = M_{\mathfrak{t}}$) holds

$$d^{-|s|}\|h\|_{\mathfrak{t}^s} \leq \|M^s h\|_{\mathbb{J}} \leq d^{|s|}\|h\|_{\mathfrak{t}^s}, \quad \forall h \in \text{dom}(M_{\mathfrak{t}}). \quad \square$$

§04|03.07 **Comment.** Given $M \in \mathbb{M}_{\mathfrak{t},d}$ there exists $\mathfrak{w} \in \mathcal{M}_{\neq 0,\nu}(\mathcal{J}) \cap \mathcal{L}_\infty(\nu)$ such that $M = M_{\mathfrak{w}}$. Consequently, we have $\nu(\mathcal{N}_{\mathfrak{w}}) = 0$ and hence for each $s \in \mathbb{R}$ the value \mathfrak{w}_j^s is well-defined for all $j \in \mathcal{N}_{\mathfrak{w}}^c$, and thus \mathfrak{w}^s is ν -a.e. defined. In particular it follows $\mathfrak{w}^\dagger = \mathfrak{w}^{-1}$ ν -a.e., and hence $M^\dagger = M_{\mathfrak{w}^{-1}}$. Similarly, we have $M^s = M_{\mathfrak{w}^s}$ for each $s \in \mathbb{R}$. \square

§04|03.08 **Property.** Under Assumption §04|03.04 let $\theta \in \mathbb{J}$ and $M_{\mathfrak{s}} \in \mathbb{L}^{\mathbb{U}}(\mathbb{J})$ satisfy, respectively, an abstract smoothness condition $\theta \in \mathbb{J}^{\mathfrak{a},r}$ as in **Definition** §04|02.12 and a link condition $M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t},d}$ as in **Definition** §04|03.05, then $g = \mathfrak{s}\theta \in \mathbb{J}$ fulfils an abstract smoothness condition $g \in \mathbb{J}^{(\mathfrak{t}\mathfrak{a}),dr}$, since

$$d^{-2}\|g\|_{(\mathfrak{t}\mathfrak{a})^\dagger}^2 = d^{-2}(\mathfrak{t}\mathfrak{a})^{2\dagger}\nu(g^2) \leq \mathfrak{a}^{2\dagger}\mathfrak{s}^{2\dagger}\nu(g^2) = \mathfrak{a}^{2\dagger}\nu(\mathfrak{s}^{2\dagger}g^2) = \mathfrak{a}^{2\dagger}\nu(\theta^2) = \|\theta\|_{\mathfrak{a}^\dagger}^2 \leq r^2. \quad \square$$

§04|03|01 Global and maximal global ν -error

We measure similar to **Subsubsection** §04|02|01 first globally the accuracy of the orthogonal projection $\theta^m = \mathfrak{s}^\dagger g^m \in \mathbb{J}\mathbb{1}^m$ of $\theta = \mathfrak{s}^\dagger g \in \mathbb{J}$.

§04|03.09 **Property (Global v-error).** If $\mathbf{v}_\bullet \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\theta_\bullet = \mathfrak{s}_\bullet^\dagger g_\bullet \in \mathbb{L}_2(\mathbf{v}_\bullet^2 \nu)$, then for each $m \in \mathbb{N}$ we have $\theta^m \in \mathbb{L}_2(\mathbf{v}_\bullet^2 \nu)$ too, since $\|\theta^m\|_{\mathbf{v}_\bullet}^2 = \mathbf{v}_\bullet^2 \nu(\theta^2 \mathbb{1}^m) \leq \mathbf{v}_\bullet^2 \nu(\theta^2)$. Moreover, it holds

$$\|\theta^m - \theta_\bullet\|_{\mathbf{v}_\bullet}^2 = \|\theta_\bullet \mathbb{1}^{m\perp}\|_{\mathbf{v}_\bullet}^2 = \mathbf{v}_\bullet^2 \nu(\theta_\bullet^2 \mathbb{1}^{m\perp}) \leq \mathbf{v}_\bullet^2 \nu(\theta_\bullet^2) \in \mathbb{R}_{\geq 0}$$

and $\|\theta^m - \theta_\bullet\|_{\mathbf{v}_\bullet}^2 = o(1)$ as $m \rightarrow \infty$ by dominated convergence. \square

§04|03.10 **Assumption.** Consider weights $\mathbf{a}_\bullet, \mathbf{v}_\bullet \in \mathcal{M}_{> 0, \nu}(\mathcal{J})$ such that $\mathbf{a}_\bullet \in \mathbb{L}_\infty(\nu)$ and $(\mathbf{a}\mathbf{v})_\bullet := \mathbf{a}_\bullet \mathbf{v}_\bullet \in \mathbb{L}_\infty(\nu)$. We write $(\mathbf{a}\mathbf{v})_{(\bullet)} = ((\mathbf{a}\mathbf{v})_{(m)}) := \|(\mathbf{a}\mathbf{v})_\bullet \mathbb{1}^{m\perp}\|_{\mathbb{L}_\infty(\nu)}\|_{m \in \mathbb{N}}$, where by construction $(\mathbf{a}\mathbf{v})_{(\bullet)} \in (\mathbb{R}_{> 0})_{\searrow}^{\mathbb{N}}$ (compare **Notation** §04|02.08). \square

§04|03.11 **Reminder.** Under Assumption §04|03.10 we have $\mathbb{J}^{\mathbf{a}} = \mathbb{L}_2(\nu) = \text{dom}(M_{\mathbf{a}}) = \mathbb{J}\mathbf{a}_\bullet = \mathbb{L}_2(\mathbf{a}_\bullet^{2\ddagger} \nu)$ and the three measures $\nu, \mathbf{a}_\bullet^{2\ddagger} \nu$ and $\mathbf{v}_\bullet^2 \nu$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04|01.02). Consequently, $\mathbb{J}^{\mathbf{a}} \subseteq \mathbb{J} = \mathbb{L}_2(\nu)$ and if $h_\bullet \in \mathbb{L}_2(\mathbf{a}_\bullet^{2\ddagger} \nu)$ satisfies $\mathbf{v}_\bullet^2 \nu(h_\bullet^2) \in \mathbb{R}_{> 0}$, for example, then $h_\bullet \in \mathbb{L}_2(\mathbf{v}_\bullet^2 \nu)$ too. Moreover under Assumption §04|03.10 we have $\mathbb{J}^{\mathbf{a}, \mathbf{r}} \subseteq \mathbb{J}^{\mathbf{a}} \subseteq \mathbb{L}_2(\mathbf{v}_\bullet^2 \nu)$ (see **Definition** §04|02.12 and **Property** §04|02.11). \square

§04|03.12 **Property (Maximal global v-error).** Under Assumption §04|03.10 for each $m \in \mathbb{N}$ and for each solution $\theta_\bullet = \mathfrak{s}_\bullet^\dagger g_\bullet \in \mathbb{J}^{\mathbf{a}, \mathbf{r}} \subseteq \mathbb{L}_2(\mathbf{v}_\bullet^2 \nu)$ its orthogonal projection $\theta^m := \theta_\bullet \mathbb{1}^m = \mathfrak{s}_\bullet^\dagger g^m \in \mathbb{J}\mathbb{1}^m$ satisfies

$$\|\theta^m - \theta_\bullet\|_{\mathbf{v}_\bullet} = \|\theta_\bullet \mathbb{1}^{m\perp}\|_{\mathbf{v}_\bullet} \leq r(\mathbf{a}\mathbf{v})_{(m)}$$

as shown in **Proof** §04|02.14. \square

§04|03|02 Local and maximal local ϕ -error

Secondly, we measure locally the accuracy of the orthogonal projection $\theta^m = \mathfrak{s}_\bullet^\dagger g^m \in \mathbb{J}\mathbb{1}^m$ of $\theta_\bullet = \mathfrak{s}_\bullet^\dagger g_\bullet \in \mathbb{J}$.

§04|03.13 **Property (Local ϕ -error).** If $\phi_\bullet \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\theta_\bullet = \mathfrak{s}_\bullet^\dagger g_\bullet \in \text{dom}(\phi\nu)$, then for each $m \in \mathbb{N}$ we have $\theta^m \in \text{dom}(\phi\nu)$ too, since $\|\phi_\bullet \theta^m\|_{\mathbb{L}_1(\nu)} = \nu(|\phi_\bullet \theta^m| \mathbb{1}^m) \leq \nu(|\phi_\bullet \theta|)$. Moreover, it holds

$$|\phi\nu(\theta) - \phi\nu(\theta^m)| \leq |\phi_\bullet| \nu(|\theta^m - \theta|) = |\phi_\bullet| \nu(|\theta| \mathbb{1}^{m\perp}) \leq \nu(|\phi_\bullet \theta|) \in \mathbb{R}_{\geq 0}$$

and $|\phi\nu(\theta) - \phi\nu(\theta^m)| = o(1)$ as $m \rightarrow \infty$ by dominated convergence. \square

§04|03.14 **Assumption.** Consider $\mathbf{a}_\bullet, \phi_\bullet \in \mathcal{M}_{> 0, \nu}(\mathcal{J})$ such that $\mathbf{a}_\bullet \in \mathbb{L}_\infty(\nu)$ and $(\mathbf{a}\phi)_\bullet := (\mathbf{a}_j \phi_j)_{j \in \mathcal{J}} = \mathbf{a}_\bullet \phi_\bullet \in \mathbb{L}_2(\nu)$ and hence $\|\mathbf{a}_\bullet \mathbb{1}^{m\perp}\|_{\phi_\bullet} = \|(\mathbf{a}\phi)_\bullet \mathbb{1}^{m\perp}\|_{\mathbb{L}_2(\nu)} = o(1)$ as $m \rightarrow \infty$. \square

§04|03.15 **Reminder.** Under Assumption §04|03.14 we have $\mathbb{J}^{\mathbf{a}} = \mathbb{L}_2(\nu) = \text{dom}(M_{\mathbf{a}}) = \mathbb{J}\mathbf{a}_\bullet = \mathbb{L}_2(\mathbf{a}_\bullet^{2\ddagger} \nu)$ and the three measures $\nu, |\phi_\bullet| \nu$ and $\mathbf{a}_\bullet^{2\ddagger} \nu$ dominate mutually each other (see **Property** §04|01.02). Consequently, $\mathbb{J}^{\mathbf{a}} \subseteq \mathbb{J} = \mathbb{L}_2(\nu)$ and if $h_\bullet \in \mathbb{L}_2(\mathbf{a}_\bullet^{2\ddagger} \nu)$ satisfies $\nu(|\phi_\bullet h_\bullet|) \in \mathbb{R}_{> 0}$, for example, then $h_\bullet \in \mathbb{L}_1(|\phi_\bullet| \nu)$ too. Moreover, under Assumption §04|03.14 we have $\mathbb{J}^{\mathbf{a}, \mathbf{r}} \subseteq \mathbb{J}^{\mathbf{a}} \subseteq \text{dom}(\phi\nu)$ (see **Definition** §04|02.12 and **Property** §04|02.11). \square

§04|03.16 **Property (Maximal local ϕ -error).** Under Assumption §04|03.14 for each $m \in \mathbb{N}$ and for each solution $\theta_\bullet = \mathfrak{s}_\bullet^\dagger g_\bullet \in \mathbb{J}^{\mathbf{a}, \mathbf{r}} \subseteq \text{dom}(\phi\nu)$ its orthogonal projection $\theta^m := \theta_\bullet \mathbb{1}^m = \mathfrak{s}_\bullet^\dagger g^m \in \mathbb{J}\mathbb{1}^m$ satisfies

$$|\phi\nu(\theta - \theta^m)| = |\phi\nu(\theta_\bullet \mathbb{1}^{m\perp})| \leq \nu(|\phi_\bullet \theta| \mathbb{1}^{m\perp}) \leq r \|\mathbf{a}_\bullet \mathbb{1}^{m\perp}\|_{\phi_\bullet}$$

as shown in **Proof** §04|02.26. \square

§05 (Generalised) linear Galerkin approach

§05100.01 **Notation.** Consider $\mathbb{J} = \ell_2 := \mathbb{L}_2(\nu_{\mathbb{N}}) = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ with counting measure $\nu_{\mathbb{N}} := \sum_{j \in \mathbb{N}} \delta_{\{j\}}$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ and $V \in \mathbb{L}(\mathbb{G}, \ell_2)$. For each $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ and $T_{\bullet, \bullet} := VTU^* \in \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$ (compare **Notation** §01104.03) we identify the kernel (infinite dimensional matrix) $T_{\bullet, \bullet} = (T_{j, j_0})_{j, j_0 \in \mathbb{N}} \in \mathbb{R}^{\mathbb{N}^2}$ and the map from ℓ_2 into itself given by

$$a_{\bullet} \mapsto T_{\bullet, \bullet} a_{\bullet} := \left(\sum_{j_0 \in \mathbb{N}} T_{j, j_0} a_{j_0} = \langle T_{j, \bullet}, a_{\bullet} \rangle_{\ell_2} = \nu_{\mathbb{N}}(T_{j, \bullet} a_{\bullet}) \right)_{j \in \mathbb{N}}$$

(compare **Notation** §01105.01). Moreover, we denote by $\mathbb{L}^{\succ}(\ell_2)$ the subset of all strictly positive definite operator in $\mathbb{L}(\ell_2)$. For each $T_{\bullet, \bullet} \in \mathbb{L}^{\succ}(\ell_2)$ we denote its Moore-Penrose inverse by $T_{\bullet, \bullet}^{\dagger} : \ell_2 \supseteq \text{dom}(T_{\bullet, \bullet}^{\dagger}) \rightarrow \ell_2$ (see **Definition** §03100.08). \square

§05100.02 **Notation (Property).** For $m \in \mathbb{N}$ set $\mathbb{1}^{m \perp} := \mathbb{1} - \mathbb{1}^m \in \mathbb{R}^{\mathbb{N}}$ where $0 = \mathbb{1}^{m \perp} \mathbb{1}^m = \mathbb{1}^m \mathbb{1}^{m \perp} \in \mathbb{R}^{\mathbb{N}}$.

(a) For $a_{\bullet} \in \mathbb{R}^{\mathbb{N}}$ introduce its sub-vector $[a_{\bullet}]_m := (a_i)_{i \in \llbracket m \rrbracket} \in \mathbb{R}^m$ where $[a_{\bullet}]_m = [a_{\bullet} \mathbb{1}^m]_m$.

(b) For $A_{\bullet, \bullet} = (A_{j, j_0})_{j, j_0 \in \mathbb{N}} \in \mathbb{R}^{\mathbb{N}^2}$ introduce its sub-matrix $[A_{\bullet, \bullet}]_m := (A_{j, j_0})_{j, j_0 \in \llbracket m \rrbracket} \in \mathbb{R}^{(m, m)}$. Clearly, if we restrict $A_{\bullet, \bullet}^m := M_{\mathbb{1}^m} A_{\bullet, \bullet} M_{\mathbb{1}^m} \in \mathbb{L}(\ell_2)$ with

$$a_{\bullet} \mapsto A_{\bullet, \bullet}^m a_{\bullet} = \left(\mathbb{1}_j^m \sum_{j_0 \in \llbracket m \rrbracket} A_{j, j_0} a_{j_0} = \mathbb{1}_j^m \langle A_{j, \bullet} \mathbb{1}^m, a_{\bullet} \rangle_{\ell_2} = \mathbb{1}_j^m \nu_{\mathbb{N}}(A_{j, \bullet} a_{\bullet} \mathbb{1}^m) \right)_{j \in \mathbb{N}}$$

to an operator from \mathbb{R}^m ($\text{ran}(M_{\mathbb{1}^m}) = \ell_2 \mathbb{1}^m$) to itself, then it is represented by the matrix $[A_{\bullet, \bullet}]_m$. Note that the adjoint $A_{\bullet, \bullet}^* \in \mathbb{L}(\ell_2)$ of $A_{\bullet, \bullet} \in \mathbb{L}(\ell_2)$ and the transposed matrix $[A_{\bullet, \bullet}]_m^* \in \mathbb{R}^{(m, m)}$ of $[A_{\bullet, \bullet}]_m$ satisfy $[A_{\bullet, \bullet}^*]_m = [A_{\bullet, \bullet}]_m^*$. If $[A_{\bullet, \bullet}]_m^{\dagger} \in \mathbb{R}^{(m, m)}$ denotes the Moore-Penrose inverse of $[A_{\bullet, \bullet}]_m$ (as linear map from \mathbb{R}^m into itself), then the Moore-Penrose inverse $A_{\bullet, \bullet}^{m \dagger} \in \mathbb{L}(\ell_2)$ of $A_{\bullet, \bullet}^m$ (see **Definition** §03100.08), restricted to an operator from \mathbb{R}^m to itself can be represented by the matrix $[A_{\bullet, \bullet}]_m^{\dagger}$. In particular, if $[A_{\bullet, \bullet}]_m$ is *regular* (invertible), and hence $[A_{\bullet, \bullet}]_m^{\dagger} = [A_{\bullet, \bullet}]_m^{-1}$, then we have $A_{\bullet, \bullet}^m A_{\bullet, \bullet}^{m \dagger} = M_{\mathbb{1}^m} = A_{\bullet, \bullet}^{m \dagger} A_{\bullet, \bullet}^m$.

(c) Given $M_w \in \mathbb{L}(\ell_2)$, the diagonal matrix $[M_w]_m \in \mathbb{R}^{(m, m)}$ has $[w]_m$ as its diagonal entries. Note that $[M_w]_m^s = [M_{w^s}]_m = [M_w^s]_m$ for all $s \in \mathbb{R}_{>0}$ and $[M_w]_m^{\dagger} = [M_{w^{\dagger}}]_m = [M_w^{\dagger}]_m$.

(d) Keep in mind the Euclidean norm $\|\cdot\|$ of a vector and the weighted norm $\|\cdot\|_{\mathfrak{t}} := \|M_{\mathfrak{t}}\|_{\ell_2}$ with $\mathfrak{t} \in \mathbb{R}_{>0}^{\mathbb{N}}$. For all $a_{\bullet} \in \ell_2 \mathbb{1}^m$ (and $(\mathfrak{t} a)_{\bullet} := \mathfrak{t} a_{\bullet} \in \ell_2 \mathbb{1}^m$) we have

$$\begin{aligned} \|a_{\bullet}\|_{\mathfrak{t}}^2 &= \|M_{\mathfrak{t}} a_{\bullet}\|_{\ell_2}^2 = \|(\mathfrak{t} a)_{\bullet}\|_{\ell_2}^2 = \|(\mathfrak{t} a)_{\bullet} \mathbb{1}^m\|_{\ell_2}^2 = [a_{\bullet}]_m^* [M_{\mathfrak{t}}]_m [a_{\bullet}]_m \\ &= [(\mathfrak{t} a)_{\bullet}]_m^* [(\mathfrak{t} a)_{\bullet}]_m = \|[M_{\mathfrak{t}}]_m [a_{\bullet}]_m\|^2 = \|[(\mathfrak{t} a)_{\bullet}]_m\|^2. \end{aligned}$$

(e) Let $\|A\|_{\text{spec}} := \sup \{\|Ax\| : \|x\| \in [0, 1]\}$ denote the spectral norm of a matrix A . Then we have $\|A_{\bullet, \bullet}^m\|_{\mathbb{L}(\ell_2)} = \|M_{\mathbb{1}^m} A_{\bullet, \bullet} M_{\mathbb{1}^m}\|_{\mathbb{L}(\ell_2)} = \|[A_{\bullet, \bullet}]_m\|_{\text{spec}}$ and for $s \in \mathbb{R}_{>0}$ hence

$$\|(M_{\mathfrak{t}}^m)^s\|_{\mathbb{L}(\ell_2)} = \|M_{\mathbb{1}^m} M_{\mathfrak{t}}^s M_{\mathbb{1}^m}\|_{\mathbb{L}(\ell_2)} = \|[M_{\mathfrak{t}}]_m\|_{\text{spec}} = \|\mathfrak{t}^s \mathbb{1}^m\|_{\ell_{\infty}} = \max \{|\mathfrak{t}_j^s| : j \in \llbracket m \rrbracket\}. \quad \square$$

§05|01 Linear Galerkin approach

§05101.01 **Assumption.** For $\mathbb{J} = \ell_2$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ and $V \in \mathbb{L}(\mathbb{G}, \ell_2)$ fixed and presumed to be *known* in advance, the operator $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ satisfies $T_{\bullet, \bullet} = VTU^* \in \mathbb{L}^{\succ}(\ell_2) \subseteq \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$. Let $g \in \text{dom}(T_{\bullet, \bullet}^{\dagger}) = \text{ran}(T_{\bullet, \bullet})$, and hence $\theta = T_{\bullet, \bullet}^{\dagger} g = T_{\bullet, \bullet}^{-1} g \in \ell_2$. \square

§05101.02 **Linear Galerkin approach.** Let $T_{\bullet,\bullet} \in \mathbb{L}^{\geq}(\ell_2)$ and $g_{\bullet} \in \ell_2$. For $m \in \mathbb{N}$ any element $\theta^m \in \ell_2 \mathbb{1}^m$, i.e. $0 = \theta^m(\mathbb{1} - \mathbb{1}^m) = \theta^m \mathbb{1}^{m\perp}$, satisfying

$$\langle \theta^m, T_{\bullet,\bullet} \theta^m \rangle_{\ell_2} - 2\langle \theta^m, g_{\bullet} \rangle_{\ell_2} \leq \langle a_{\bullet}, T_{\bullet,\bullet} a_{\bullet} \rangle_{\ell_2} - 2\langle a_{\bullet}, g_{\bullet} \rangle_{\ell_2} \quad \text{for all } a_{\bullet} \in \ell_2 \mathbb{1}^m$$

is called a *Galerkin solution* in $\ell_2 \mathbb{1}^m$. □

§05101.03 **Lemma.** Let $T_{\bullet,\bullet} \in \mathbb{L}^{\geq}(\ell_2)$ and $g_{\bullet} \in \ell_2$. (i) For all $m \in \mathbb{N}$ the matrix $[T_{\bullet,\bullet}]_m \in \mathbb{R}^{(m,m)}$ is *strictly positive definite*. (ii) The Galerkin solution $\theta^m \in \ell_2 \mathbb{1}^m$, i.e. $0 = \theta^m(\mathbb{1} - \mathbb{1}^m) = \theta^m \mathbb{1}^{m\perp}$, is *uniquely determined* by $[\theta^m]_m = [T_{\bullet,\bullet}]_m^{-1} [g_{\bullet}]_m$, and hence $\theta^m = T_{\bullet,\bullet}^{\dagger} g_{\bullet}$. (iii) If in addition $g_{\bullet} \in \text{dom}(T_{\bullet,\bullet}^{\dagger})$ and $\theta_{\bullet} := T_{\bullet,\bullet}^{\dagger} g_{\bullet} \in \ell_2$, then the Galerkin solution θ^m *minimises* in $\ell_2 \mathbb{1}^m$ the functional $a_{\bullet} \rightarrow F(a_{\bullet}) = \|T_{\bullet,\bullet}^{1/2}(a_{\bullet} - \theta_{\bullet})\|_{\ell_2}^2$.

§05101.04 **Proof of Lemma §05101.03.** Given in the lecture. □

§05101.05 **Remark.** Consider for $\theta \in \ell_2$ its orthogonal projection $\mathbb{1}^m \theta$ and $\mathbb{1}^{m\perp} \theta$ onto the subspace $\ell_2 \mathbb{1}^m$ and its orthogonal complement $\ell_2 \mathbb{1}^{m\perp} := (\ell_2 \mathbb{1}^m)^{\perp}$, respectively, then the approximation error $\|\mathbb{1}^m \theta - \theta\|_{\ell_2} = \|(\mathbb{1} - \mathbb{1}^m) \theta\|_{\ell_2} = \|\mathbb{1}^{m\perp} \theta\|_{\ell_2}$ converges to zero as $m \rightarrow \infty$ by Lebesgue's dominated convergence theorem. On the other hand, if $g_{\bullet} \in \text{dom}(T_{\bullet,\bullet}^{\dagger})$ and $\theta_{\bullet} := T_{\bullet,\bullet}^{\dagger} g_{\bullet} \in \ell_2$ then the Galerkin solution $\theta^m \in \ell_2 \mathbb{1}^m$ satisfies $[\mathbb{1}^m \theta - \theta^m]_m = -[T_{\bullet,\bullet}]_m^{-1} [T_{\bullet,\bullet}(\mathbb{1} - \mathbb{1}^m) \theta]_m = -[T_{\bullet,\bullet}]_m^{-1} [T_{\bullet,\bullet} \mathbb{1}^{m\perp} \theta]_m$ and, hence it does generally not correspond to the orthogonal projection $\mathbb{1}^{m\perp} \theta = (\mathbb{1} - \mathbb{1}^m) \theta$. Moreover, the approximation error $\sup \{\|\theta^m - \theta_{\bullet}\|_{\ell_2} : m \in \mathbb{N}_{\geq n}\}$ does generally not converge to zero as $n \rightarrow \infty$. However, if

$$\begin{aligned} C_T &:= \sup \left\{ \|T_{\bullet,\bullet}^{\dagger} T_{\bullet,\bullet} M_{\mathbb{1}^{m\perp}}\|_{\mathbb{L}(\ell_2)} : m \in \mathbb{N} \right\} \\ &= \sup \left\{ \|[T_{\bullet,\bullet}]_m^{-1} [T_{\bullet,\bullet} \mathbb{1}^{m\perp} a_{\bullet}]_m\| : \|a_{\bullet}\|_{\ell_2} = 1, a_{\bullet} \in \ell_2, m \in \mathbb{N} \right\} \in \mathbb{R}_{\geq 0} \end{aligned}$$

then $\|\theta^m - \theta_{\bullet}\|_{\ell_2} \leq (1 + C_T) \|\mathbb{1}^{m\perp} \theta_{\bullet}\|_{\ell_2}$ which implies $\sup \{\|\theta^m - \theta_{\bullet}\|_{\ell_2} : m \in \mathbb{N}_{\geq n}\} = o(1)$ as $m \rightarrow \infty$. Here and subsequently, we will restrict ourselves to classes of solutions and operators which ensure the convergence. Obviously, this is a minimal regularity condition for us if we aim to estimate the Galerkin solution. □

§05101.06 **Notation (Reminder §04102.08).** For sequences $a_{\bullet}, b_{\bullet} \in (\mathbb{K})^{\mathbb{N}}$ taking its values in $\mathbb{K} \in \{\mathbb{R}_{>0}, \mathbb{Q}, \mathbb{Z}, \dots\}$ we write $a_{\bullet} \in (\mathbb{K})_{\nearrow}^{\mathbb{N}}$ and $b_{\bullet} \in (\mathbb{K})_{\searrow}^{\mathbb{N}}$ if a_{\bullet} and b_{\bullet} is monotonically, respectively, *non-decreasing* and *non-increasing*. If in addition $a_n \rightarrow \infty$ and $b_n \rightarrow 0$ as $n \rightarrow \infty$, then we write $a_{\bullet} \in (\mathbb{K})_{\uparrow\infty}^{\mathbb{N}}$ and $b_{\bullet} \in (\mathbb{K})_{\downarrow 0}^{\mathbb{N}}$ for short. □

§05101.07 **Property.** If $\mathfrak{t}_{\bullet} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ is monotonically non-increasing, then for all $m \in \mathbb{N}$ we have

$$\|\mathfrak{t}_{\bullet}^{-1} \mathbb{1}^m\|_{\ell_{\infty}}^{-1} = \min \{\mathfrak{t}_j : j \in \llbracket m \rrbracket\} \geq \sup \{\mathfrak{t}_j : j \in \mathbb{N}_{> m}\} = \|\mathfrak{t}_{\bullet} \mathbb{1}^{m\perp}\|_{\ell_{\infty}} = \mathfrak{t}_{(m)},$$

and hence $1 \geq \mathfrak{t}_{(m)} \|\mathfrak{t}_{\bullet}^{-1} \mathbb{1}^m\|_{\ell_{\infty}} = \|\mathfrak{t}_{\bullet} \mathbb{1}^{m\perp}\|_{\ell_{\infty}} \|\mathfrak{t}_{\bullet}^{-1} \mathbb{1}^m\|_{\ell_{\infty}}$. □

§05101.08 **Link condition.** Given weights $\mathfrak{t}_{\bullet} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ an operator $T_{\bullet,\bullet} \in \mathbb{L}^{\geq}(\ell_2)$ satisfies a *link condition* if there is $d \in \mathbb{R}_{\geq 1}$ such that

$$T_{\bullet,\bullet} \in \mathbb{T}_{\mathfrak{t},d}^{\geq} := \{A_{\bullet,\bullet} \in \mathbb{L}^{\geq}(\ell_2) : d^{-1} \|a_{\bullet}\|_{\mathfrak{t}} \leq \|T_{\bullet,\bullet} a_{\bullet}\|_{\ell_2} \leq d \|a_{\bullet}\|_{\mathfrak{t}} \text{ for all } a_{\bullet} \in \ell_2\}$$

and we set $\mathbb{T}_{\mathfrak{t},d} := \{A_{\bullet,\bullet} \in \mathbb{L}^{\geq}(\ell_2) : (A_{\bullet,\bullet}^* A_{\bullet,\bullet})^{1/2} \in \mathbb{T}_{\mathfrak{t},d}^{\geq}\}$.

§05101.09 **Remark.** Note that $\mathfrak{t} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ for each $A_{\cdot,\cdot} \in \mathbb{T}_{\mathfrak{t},d}^{\geq}$ implies $\ker(A_{\cdot,\cdot}) = \{0\}$, i.e. $A_{\cdot,\cdot}$ is injective and hence strictly positive definite. We shall emphasise that for $T_{\cdot,\cdot} \in \mathbb{L}(\ell_2)$ the condition $T_{\cdot,\cdot} \in \mathbb{T}_{\mathfrak{t},d}$ is equivalent to

$$d^{-1} \|a\|_{\mathfrak{t}} \leq \|T_{\cdot,\cdot} a\|_{\ell_2} \leq d \|a\|_{\mathfrak{t}} \quad \text{for all } a \in \ell_2. \quad (05.01)$$

Observe further that $M_s \in \mathbb{L}(\ell_2)$ satisfies the link condition $M_s \in \mathbb{T}_{\mathfrak{t},d}$ as in [Definition §05101.08](#) if and only if $|\mathfrak{s}_j| \leq d \mathfrak{t}_j \wedge \mathfrak{t}_j \leq d |\mathfrak{s}_j|$ ($\nu_{\mathbb{N}}$ -a.e.), i.e. $M_s \in \mathbb{M}_{\mathfrak{t},d}$ as in [Definition §04103.05](#). Thereby, we have $\mathbb{M}_{\mathfrak{t},d} \subseteq \mathbb{T}_{\mathfrak{t},d}$. We shall emphasise, that there are operators satisfying the link condition $\mathbb{T}_{\mathfrak{t},d}^{\geq}$ which do not belong to $\mathbb{M}_{\mathfrak{t},d}$, i.e., they are non-diagonal. Let us briefly give a construction of one of those. We consider a small perturbation of $M_{\mathfrak{t}}$, that is, $T_{\cdot,\cdot} = M_{\mathfrak{t}} + M_{\mathfrak{t}} A_{\cdot,\cdot} M_{\mathfrak{t}}$ where $A_{\cdot,\cdot} \in \mathbb{L}^{\geq}(\ell_2)$ is a positive definite operator with spectral norm $c := \|M_{\mathfrak{t}} A_{\cdot,\cdot}\|_{\mathbb{L}(\ell_2)} < 1$. Obviously, $T_{\cdot,\cdot}$ is strictly positive definite, and $\|T_{\cdot,\cdot} a\|_{\ell_2} \leq \|\text{id}_{\ell_2} + M_{\mathfrak{t}} A_{\cdot,\cdot}\|_{\mathbb{L}(\ell_2)} \|M_{\mathfrak{t}} a\|_{\ell_2} \leq (1+c) \|a\|_{\mathfrak{t}}$. On the other hand, we have $\|(\text{id}_{\ell_2} + M_{\mathfrak{t}} A_{\cdot,\cdot})^{-1}\|_{\mathbb{L}(\ell_2)} = \frac{1}{1 - \|M_{\mathfrak{t}} A_{\cdot,\cdot}\|_{\mathbb{L}(\ell_2)}} = \frac{1}{1-c}$ by the Neumann series argument ??, which in turn implies $\|a\|_{\mathfrak{t}} = \|M_{\mathfrak{t}} a\|_{\ell_2} = \|(\text{id}_{\ell_2} + M_{\mathfrak{t}} A_{\cdot,\cdot})^{-1}\|_{\mathbb{L}(\ell_2)} \|T_{\cdot,\cdot} a\|_{\ell_2} \leq \frac{1}{1-c} \|T_{\cdot,\cdot} a\|_{\ell_2}$. Combining both bounds the operator $T_{\cdot,\cdot}$ satisfies the link condition $T_{\cdot,\cdot} \in \mathbb{T}_{\mathfrak{t},d}^{\geq}$ for all $d \geq \max(1+c, \frac{1}{1-c})$ and is obviously not a multiplication operator, i.e. diagonal. \square

§05101.10 **Property.** If $T_{\cdot,\cdot} \in \mathbb{T}_{\mathfrak{t},d}^{\geq}$ with $\mathfrak{t} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ and $d \in \mathbb{R}_{\geq 1}$ then for all $s \in [-1, 1]$ we have

$$(\text{inequality of Heinz [1951]}) \quad d^{-|s|} \|a\|_{\mathfrak{t}^s} \leq \|T_{\cdot,\cdot}^s a\|_{\ell_2} \leq d^{|s|} \|a\|_{\mathfrak{t}^s} \quad \text{for all } a \in \text{dom}(M_{\mathfrak{t}}). \quad \square$$

§05101.11 **Comment.** Given $T_{\cdot,\cdot} \in \mathbb{T}_{\mathfrak{t},d}^{\geq}$ we have $\ker(T_{\cdot,\cdot}) = \{0\} = \ker(T_{\cdot,\cdot}^*)$ and on $\text{ran}(T_{\cdot,\cdot}) = \text{dom}(T_{\cdot,\cdot}^{\dagger})$ (which is dense in ℓ_2) we have $T_{\cdot,\cdot}^{-1} = T_{\cdot,\cdot}^{\dagger}$. Similarly, for each $s \in \mathbb{R}_{\geq 0}$ on $\text{ran}(T_{\cdot,\cdot}^s) = \text{dom}(T_{\cdot,\cdot}^{s\dagger})$ we have $T_{\cdot,\cdot}^{-s} = T_{\cdot,\cdot}^{s\dagger} = (T_{\cdot,\cdot}^s)^{\dagger}$. \square

§05101.12 **Notation.** Given weights $\mathfrak{a} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ introduce $\ell_2^{\mathfrak{a}} := \ell_2(\mathfrak{a}^{-2}) := \mathbb{L}_2(\mathfrak{a}^{-2} \nu_{\mathbb{N}}) = \ell_2 \mathfrak{a} = \text{ran}(M_{\mathfrak{a}}) = \mathbb{J}^{\mathfrak{a}} \subseteq \ell_2 = \mathbb{J}$ endowed with $\|\cdot\|_{\mathfrak{a}^{-1}} := \|M_{\mathfrak{a}^{-1}} \cdot\|_{\ell_2}$ (as in [Property §04101.02](#)). We assume in the following that $\theta \in \ell_2$ satisfies an abstract smoothness condition ([Definition §04102.12](#)), i.e., there is $r \in \mathbb{R}_{>0}$ such that $\theta \in \ell_2^{\mathfrak{a},r} := \mathbb{J}^{\mathfrak{a},r} = \{h \in \ell_2^{\mathfrak{a}} : \|h\|_{\mathfrak{a}^{-1}} \leq r\} \subseteq \ell_2^{\mathfrak{a}} \subseteq \ell_2$. \square

§05101.13 **Source condition.** Given $T_{\cdot,\cdot} \in \mathbb{L}(\ell_2)$, the solution $\theta \in \ell_2$ satisfies a *source condition*, if there is $s \in \mathbb{R}_{>0}$ such that $\theta \in \text{ran}((T_{\cdot,\cdot}^* T_{\cdot,\cdot})^{s/2})$, that is, $\theta = (T_{\cdot,\cdot}^* T_{\cdot,\cdot})^{s/2} a$ for some $a \in \ell_2$. \square

§05101.14 **Corollary.** For $\mathfrak{a}, \mathfrak{t} \in \mathbb{R}_{>0}$ and $\mathfrak{v} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ set $\mathfrak{t} := \mathfrak{v}^{\mathfrak{t}}, \mathfrak{a} := \mathfrak{v}^{\mathfrak{a}} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$. Consider $\ell_2^{\mathfrak{a}} = \ell_2 \mathfrak{a}$, and assume that $T_{\cdot,\cdot} \in \mathbb{T}_{\mathfrak{t},d}^{\geq}$. If $\mathfrak{a} \leq \mathfrak{t}$ then (i) for any $\theta \in \ell_2^{\mathfrak{a}}$ we have $\theta = T_{\cdot,\cdot}^{\mathfrak{a}/\mathfrak{t}} h$ with $\|h\|_{\ell_2} \leq d^{\mathfrak{a}/\mathfrak{t}} \|\theta\|_{\mathfrak{a}^{-1}}$, and conversely (ii) for any $\theta = T_{\cdot,\cdot}^{\mathfrak{a}/\mathfrak{t}} h$ with $h \in \ell_2$ we obtain $\theta \in \ell_2^{\mathfrak{a}}$ with $\|\theta\|_{\mathfrak{a}^{-1}} \leq d^{\mathfrak{a}/\mathfrak{t}} \|h\|_{\ell_2}$.

§05101.15 **Proof of Corollary §05101.14.** Given in the lecture. \square

§05101.16 **Comment.** Under the assumptions of [Corollary §05101.14](#) if $T_{\cdot,\cdot} \in \mathbb{T}_{\mathfrak{t},d}$ and $\mathfrak{a} \leq \mathfrak{t}$ then (i) for any $\theta \in \ell_2^{\mathfrak{a}}$ we have $\theta = (T_{\cdot,\cdot}^* T_{\cdot,\cdot})^{\mathfrak{a}/(2\mathfrak{t})} h$ with $\|h\|_{\ell_2} \leq d^{\mathfrak{a}/\mathfrak{t}} \|\theta\|_{\mathfrak{a}^{-1}}$, and conversely (ii) for any $\theta = (T_{\cdot,\cdot}^* T_{\cdot,\cdot})^{\mathfrak{a}/(2\mathfrak{t})} h$ with $h \in \ell_2$ we obtain $\theta \in \ell_2^{\mathfrak{a}}$ with $\|\theta\|_{\mathfrak{a}^{-1}} \leq d^{\mathfrak{a}/\mathfrak{t}} \|h\|_{\ell_2}$. \square

§05101.17 **Corollary.** Given $d, r \in \mathbb{R}_{>0}$ and $\mathfrak{t}, \mathfrak{a} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ set $(\mathfrak{t}\mathfrak{a}) := \mathfrak{t}\mathfrak{a} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$. If $T_{\cdot,\cdot} \in \mathbb{T}_{\mathfrak{t},d}$ and $\theta \in \ell_2^{\mathfrak{a},r}$, then we have $g = T_{\cdot,\cdot} \theta \in \ell_2^{(\mathfrak{t}\mathfrak{a}), d^r}$.

§05101.18 **Proof of Corollary §05101.17.** Given in the lecture. \square

§05101.19 **Remark.** Keeping the orthonormal basis $(\mathbf{1}^{\{j\}})_{j \in \mathbb{N}}$ in ℓ_2 in mind ([Notation §01104.02](#)) each $M_{\mathfrak{t}} \in \mathbb{L}(\ell_2)$ with $\mathfrak{t} \in \ell_{\infty}$ is *self-adjoint* with eigensystem $((\mathfrak{t}_j, \mathbf{1}^{\{j\}}))_{j \in \mathbb{N}}$. Indeed, for all $j \in \mathbb{N}$ we have

$M_{\cdot} \mathbb{1}^{\{j\}} = \epsilon_j \mathbb{1}^{\{j\}}$. Recall that $\mathbb{K}(\ell_2)$ denotes the subset of $\mathbb{L}(\ell_2)$ containing all compact operators. If $A_{\cdot} \in \mathbb{K}(\ell_2)$ is compact and in addition self-adjoint, then A_{\cdot} admits an eigensystem $((\epsilon_j, e_j))_{j \in \mathbb{N}}$ where $\epsilon \in \ell_\infty$ contains each eigenvalue of A_{\cdot} repeated according to its multiplicity (with zero as only accumulation point) and $e = (e_j)_{j \in \mathbb{N}}$ is the associated eigenbasis. We denote by $\mathbb{K}^{\geq}(\ell_2)$ the subset of $\mathbb{L}^{\geq}(\ell_2)$ containing all compact operators. If $A_{\cdot} \in \mathbb{K}^{\geq}(\ell_2)$ then we have $\epsilon \in (\mathbb{R}_{>0})_{10}^{\mathbb{N}}$ (possibly after a reordering). \square

§05101.20 **Lemma.** Consider as in Definition §05101.08 a link condition $\mathbb{T}_{\mathfrak{t}, \mathfrak{d}}^{\geq}$ with $\mathfrak{t} \in (\mathbb{R}_{>0})_{\setminus \infty}^{\mathbb{N}}$. Let the operator $T_{\cdot} \in \mathbb{K}^{\geq}(\ell_2)$ admit $((\epsilon_j, e_j))_{j \in \mathbb{N}}$ as eigensystem where $\epsilon \in (\mathbb{R}_{>0})_{\setminus \infty}^{\mathbb{N}}$ contains each eigenvalue of T_{\cdot} repeated according to its multiplicity and the associated eigenbasis $e = (e_j)_{j \in \mathbb{N}}$ does eventually not correspond to the ONB $(\mathbb{1}^{\{j\}})_{j \in \mathbb{N}}$. If $T_{\cdot} \in \mathbb{T}_{\mathfrak{t}, \mathfrak{d}}^{\geq}$, then we have $\mathfrak{d}^{-1} \leq \epsilon_j / \mathfrak{t}_j \leq \mathfrak{d}$ for all $j \in \mathbb{N}$.

§05101.21 **Proof of Lemma §05101.20.** Given in the lecture. \square

§05101.22 **Lemma.** Consider the link condition $T_{\cdot} \in \mathbb{T}_{\mathfrak{t}, \mathfrak{d}}^{\geq}$ as in Definition §05101.08 with $\mathfrak{t} \in (\mathbb{R}_{>0})_{\setminus \infty}^{\mathbb{N}}$. For all $m \in \mathbb{N}$ and $s \in [0, 1]$ we have (i) $\epsilon_{(m)}^s \| [T_{\cdot}]_{\cdot}^{-s} \|_{\text{spec}} \leq (d(d+2))^s \leq (3d^2)^s$, (ii) $\| [T_{\cdot}]_{\cdot}^{-s} [M_{\cdot}]_{\cdot}^s \|_{\text{spec}} \leq (d(d+2))^s \leq (3d^2)^s$ and (iii) $\| [T_{\cdot}]_{\cdot}^s [M_{\cdot}]_{\cdot}^{-s} \|_{\text{spec}} \leq d^s$.

§05101.23 **Proof of Lemma §05101.22.** Given in the lecture. \square

§05|01|01 Global and maximal global v-error

We shall measure first globally the accuracy of the Galerkin solution $\theta^m \in \ell_2 \mathbb{1}^m$ of $\theta = T_{\cdot}^{\dagger} g \in \ell_2$.

§05101.24 **Property (Global v-error).** Consider $\mathfrak{v} \in (\mathbb{R}_{>0})^{\mathbb{N}}$, $T_{\cdot} \in \mathbb{L}^{\geq}(\ell_2)$ and $g \in \text{dom}(T_{\cdot}^{\dagger}) = \text{ran}(T_{\cdot}) \subseteq \ell_2$ and hence $\theta = T_{\cdot}^{\dagger} g = T_{\cdot}^{-1} g \in \ell_2$. Given $m \in \mathbb{N}$ we have $\mathfrak{v}^2 \mathbb{1}^m \in \ell_\infty$ and hence $\ell_2 \mathbb{1}^m \subseteq \ell_2(\mathfrak{v}^2)$. Consequently, denoting by $\theta^m = T_{\cdot}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ a Galerkin solution we have $\theta^m \in \ell_2(\mathfrak{v}^2)$ with

$$\|\theta^m\|_{\mathfrak{v}} \leq \| [M_{\cdot}]_{\cdot} [T_{\cdot}]_{\cdot}^{-1} \|_{\text{spec}} \|g\| \in \mathbb{R}_{>0}.$$

If $C_T := \sup \{ \| [M_{\cdot}]_{\cdot} T_{\cdot}^{m \dagger} T_{\cdot} M_{\cdot}^{\perp} \|_{\mathbb{L}(\ell_2)} : m \in \mathbb{N} \} \in \mathbb{R}_{\geq 0}$ then

$$\|\theta^m - \theta\|_{\mathfrak{v}} \leq (1 + C_T) \| \mathbb{1}^{m \perp} \theta \|_{\ell_2}$$

which implies $\sup \{ \|\theta^j - \theta\|_{\mathfrak{v}} : j \in \llbracket m, \infty \rrbracket \} = o(1)$ as $m \rightarrow \infty$. \square

§05101.25 **Notation (Reminder §04102.08).** For $w \in \ell_\infty$ we set $w_{(0)}^2 := \|w^2\|_{\ell_\infty}$ and $w_{(j)}^2 = (w_{(j)}^2 := \|w^2 \mathbb{1}^{j \perp}\|_{\ell_\infty})_{j \in \mathbb{N}}$, where by construction $w_{(j)}^2 = \sup \{ w_i^2 : i \in \mathbb{N}_{> j} \}$, $j \in \mathbb{Z}_{\geq 0}$ and $w_{(0)}^2 \in (\mathbb{R}_{\geq 0})_{\setminus \infty}^{\mathbb{N}}$. Evidently, if in addition $w_{(j)}^2 \in (\mathbb{R}_{>0})_{\setminus \infty}^{\mathbb{N}}$ then we have $w_{(0)}^2 = (w_{(j)}^2 = w_{(j+1)}^2)_{j \in \mathbb{N}}$. \square

§05101.26 **Assumption.** Consider weights $\mathfrak{t}, \mathfrak{a} \in (\mathbb{R}_{>0})_{\setminus \infty}^{\mathbb{N}}$ and $\mathfrak{v} \in (\mathbb{R}_{10})^{\mathbb{N}}$ such that $(\mathfrak{a}\mathfrak{v})_{\cdot} := \mathfrak{a} \mathfrak{v} \in \ell_\infty$ and $(\mathfrak{t}/\mathfrak{v})_{\cdot} = \mathfrak{t} \mathfrak{v}^{-1} \in \ell_\infty$ are satisfied. In addition there exists $C_{(\mathfrak{t}/\mathfrak{v})} \in (0, 1]$ such that for all $m \in \mathbb{N}$

$$(\mathfrak{t}/\mathfrak{v})_{(m-1)}^2 \geq \min \{ (\mathfrak{t}/\mathfrak{v})_j^2 : j \in \llbracket m \rrbracket \} \geq C_{(\mathfrak{t}/\mathfrak{v})} (\mathfrak{t}/\mathfrak{v})_{(m)}^2 \quad (05.02)$$

or in equal $C_{(\mathfrak{t}/\mathfrak{v})} \|(\mathfrak{t}/\mathfrak{v})_{\cdot}^{-2} \mathbb{1}^m\|_{\ell_\infty} \leq (\mathfrak{t}/\mathfrak{v})_{(m)}^{-2}$. \square

§05101.27 **Reminder.** Under Assumption §05101.26 we have $\mathbb{J}^{\mathfrak{a}} = \ell_2^{\mathfrak{a}} = \text{dom}(M_{\cdot}) = \ell_2 \mathfrak{a} = \ell_2(\mathfrak{a}^{-2})$ and the three measures $\nu_{\mathbb{N}}$, $\mathfrak{a}^{-2} \nu_{\mathbb{N}}$ and $\mathfrak{v}^2 \nu_{\mathbb{N}}$ dominate mutually each other, i.e. they share the same null sets (see Property §04101.02). Consequently, since $(\mathfrak{a}\mathfrak{v})_{\cdot} \in \ell_\infty$ and

$$\|h\|_{\mathfrak{v}} = \|(\mathfrak{a}\mathfrak{v})_{\cdot} \mathfrak{a}^{-1} h\|_{\ell_2} \leq \|(\mathfrak{a}\mathfrak{v})_{\cdot}\|_{\ell_\infty} \|h\|_{\mathfrak{a}^{-1}} \in \mathbb{R}_{\geq 0} \quad \text{for each } h \in \ell_2^{\mathfrak{a}}$$

we have $\ell_2^a \subseteq \ell_2(\mathbf{v}^2)$. Moreover, since $\mathfrak{t}, \mathfrak{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ for each $s \in [0, 1]$ we have $\mathfrak{t}^{1-s}, \mathfrak{t}^s \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and $(\mathfrak{a}\mathfrak{t}^s)_{\bullet} = \mathfrak{a}\mathfrak{t}^s \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$. We note further if in addition $(\mathfrak{t}/\mathbf{v})_{\bullet} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ is satisfied, then Assumption §05101.26 (05.02) is fulfilled with $C_{(\mathfrak{t}/\mathbf{v})} = 1$ due to **Property** §05101.07. \square

§05101.28 **Lemma (Maximal global \mathbf{v} -error)**. *Let Assumption §05101.26, $T_{\bullet, \bullet} \in \mathbb{T}_{\mathfrak{t}, \mathfrak{a}}^{\geq}$, $g_{\bullet} \in \text{dom}(T_{\bullet, \bullet}^{\dagger}) = \text{ran}(T_{\bullet, \bullet}) \subseteq \ell_2$ and $\theta_{\bullet} = T_{\bullet, \bullet}^{\dagger} g_{\bullet} = T_{\bullet, \bullet}^{-1} g_{\bullet} \in \ell_2^{\mathfrak{a}, \mathfrak{r}}$ be satisfied. Given $m \in \mathbb{N}$ denoting by $\theta_{\bullet}^m = T_{\bullet, \bullet}^{m \dagger} g_{\bullet} \in \ell_2 \mathbb{1}^m$ a Galerkin solution for any $s \in [0, 1]$ we obtain*

$$\begin{aligned} \|\theta_{\bullet} - \theta_{\bullet}^m\|_{\mathbf{v}}^2 &\leq (9d^6 C_{(\mathfrak{t}/\mathbf{v})}^{-2} + 1) (\mathfrak{a}\mathbf{v})_{(m)}^2 \|\mathbb{1}^{m \perp} \theta_{\bullet}\|_{\mathfrak{a}^{-1}}^2, & \|\theta_{\bullet}^m\|_{\mathfrak{a}^{-1}} &\leq 3d^3 \|\theta_{\bullet}\|_{\mathfrak{a}^{-1}}, & \text{and} \\ \|\mathbb{T}_{\bullet, \bullet}^s(\theta_{\bullet} - \theta_{\bullet}^m)\|_{\ell_2}^2 &\leq (9d^6 + 1)d^{2s} (\mathfrak{a}\mathfrak{t}^s)_{(m)}^2 \|\mathbb{1}^{m \perp} \theta_{\bullet}\|_{\mathfrak{a}^{-1}}^2. \end{aligned} \quad (05.03)$$

§05101.29 **Proof of Lemma §05101.28**. Given in the lecture. \square

§05|01|02 Global and maximal global ϕ -error

Secondly we measure locally the accuracy of the Galerkin solution $\theta_{\bullet}^m \in \ell_2 \mathbb{1}^m$ of $\theta_{\bullet} = T_{\bullet, \bullet}^{\dagger} g_{\bullet} \in \ell_2$.

§05101.30 **Reminder**. Given $\phi_{\bullet} \in \mathbb{R}_{\searrow 0}^{\mathbb{N}}$ for $\text{dom}(\phi_{\mathbb{N}}) := \{h_{\bullet} \in \ell_2 : \phi_{\bullet} h_{\bullet} \in \ell_1\}$ we consider as in **Notation** §04102.16 the linear functional $\phi_{\mathbb{N}} : \ell_2 \supseteq \text{dom}(\phi_{\mathbb{N}}) \rightarrow \mathbb{R}$ defined by

$$h_{\bullet} \mapsto \phi_{\mathbb{N}}(h_{\bullet}) := \nu_{\mathbb{N}}(\phi_{\bullet} h_{\bullet}) = \sum_{j \in \mathbb{N}} \phi_j h_j.$$

For each $\theta_{\bullet} \in \text{dom}(\phi_{\mathbb{N}})$ and $m \in \mathbb{N}$ by **Property** §04102.18 we have $\theta_{\bullet} \mathbb{1}^m \in \text{dom}(\phi_{\mathbb{N}})$ with

$$|\phi_{\mathbb{N}}(\theta_{\bullet} - \theta_{\bullet} \mathbb{1}^m)| \leq |\phi_{\bullet}|_{\nu_{\mathbb{N}}}(|\theta_{\bullet}| \mathbb{1}^{m \perp}) \leq \nu_{\mathbb{N}}(|\phi_{\bullet}|) \in \mathbb{R}_{>0},$$

and $|\phi_{\mathbb{N}}(\theta_{\bullet} - \theta_{\bullet} \mathbb{1}^m)| = o(1)$ as $m \rightarrow \infty$ by dominated convergence. \square

§05101.31 **Property (Local ϕ -error)**. *Consider $\phi_{\bullet} \in \mathbb{R}_{\searrow 0}^{\mathbb{N}}$, $T_{\bullet, \bullet} \in \mathbb{L}^{\geq}(\ell_2)$ and $g_{\bullet} \in \text{dom}(T_{\bullet, \bullet}^{\dagger}) = \text{ran}(T_{\bullet, \bullet}) \subseteq \ell_2$ and hence $\theta_{\bullet} = T_{\bullet, \bullet}^{\dagger} g_{\bullet} = T_{\bullet, \bullet}^{-1} g_{\bullet} \in \ell_2$. Given $m \in \mathbb{N}$ we have $\phi_{\bullet}^2 \mathbb{1}^m \in \ell_2$ and hence $\ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi_{\mathbb{N}})$. Consequently, denoting by $\theta_{\bullet}^m = T_{\bullet, \bullet}^{m \dagger} g_{\bullet} \in \ell_2 \mathbb{1}^m$ a Galerkin solution we have $\theta_{\bullet}^m \in \text{dom}(\phi_{\mathbb{N}})$ with*

$$\|\phi_{\bullet} \theta_{\bullet}^m\|_{\ell_1} \leq \|[\mathbb{T}_{\bullet, \bullet}]_m^{-1}[\phi_{\bullet}]_m\| \| [g_{\bullet}]_m \| \in \mathbb{R}_{>0}.$$

If $C_T := \sup \{ \|M_{\mathbb{T}^{m \perp}} T_{\bullet, \bullet} T_{\bullet, \bullet}^{m \dagger} \phi_{\bullet}\|_{\ell_2} : m \in \mathbb{N} \} \in \mathbb{R}_{>0}$ then

$$|\phi_{\mathbb{N}}(\theta_{\bullet}^m - \theta_{\bullet})| \leq (1 + C_T) \|\mathbb{1}^{m \perp} \theta_{\bullet}\|_{\ell_2}$$

which implies $\sup \{ |\phi_{\mathbb{N}}(\theta_{\bullet}^j - \theta_{\bullet})| : j \in \llbracket m, \infty \rrbracket \} = o(1)$ as $m \rightarrow \infty$. \square

§05101.32 **Assumption**. Let $\mathfrak{t}, \mathfrak{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and $\phi_{\bullet} \in \mathbb{R}_{\searrow 0}^{\mathbb{N}}$ such that $(\mathfrak{a}\mathfrak{t})_{\bullet} \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ and $(\mathfrak{a}\phi)_{\bullet} \in \ell_2$. \square

§05101.33 **Reminder**. Under Assumption §05101.32 we have $\mathbb{J}^{\mathfrak{a}} = \ell_2^{\mathfrak{a}} = \text{dom}(M_{\mathfrak{a}}) = \ell_2 \mathfrak{a} = \ell_2(\mathfrak{a}^{-2})$ and the three measures $\nu_{\mathbb{N}}$, $\mathfrak{a}^{-2} \nu_{\mathbb{N}}$ and $|\phi_{\bullet}|_{\nu_{\mathbb{N}}}$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04101.02). Consequently, since $(\mathfrak{a}\phi)_{\bullet} \in \ell_2$ and (**Property** §04102.23)

$$\|\phi_{\bullet} h_{\bullet}\|_{\ell_1} = \nu_{\mathbb{N}}(|h_{\bullet} \mathfrak{a}^{-1}(\mathfrak{a}\phi)_{\bullet}|) \leq \|(\mathfrak{a}\phi)_{\bullet}\|_{\ell_2} \|h_{\bullet}\|_{\mathfrak{a}^{-1}} \in \mathbb{R}_{>0} \quad \text{for each } h_{\bullet} \in \ell_2^{\mathfrak{a}}$$

we have $\ell_2^{\mathfrak{a}} \subseteq \text{dom}(\phi_{\mathbb{N}})$. Moreover, from $(\mathfrak{a}\phi)_{\bullet} \in \ell_2$ follows $\|\mathfrak{a} \mathbb{1}^{m \perp}\|_{\phi} = \|(\mathfrak{a}\phi)_{\bullet} \mathbb{1}^{m \perp}\|_{\ell_2} = o(1)$ as $m \rightarrow \infty$. For $s \in [0, 1]$ from $(\mathfrak{a}\mathfrak{t}^s)_{\bullet} = \mathfrak{a}\mathfrak{t}^s \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ follows $(\mathfrak{a}\mathfrak{t}^s)_{(\bullet)} = ((\mathfrak{a}\mathfrak{t}^s)_{(m)}) := (\mathfrak{a}\mathfrak{t}^s)_{m+1} = \|(\mathfrak{a}\mathfrak{t}^s)_{\bullet} \mathbb{1}^{m \perp}\|_{\ell_{\infty}})_{m \in \mathbb{N}} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$. \square

§05101.34 **Lemma (Maximal local ϕ -error).** *Let Assumption §05101.32, $T_{\bullet,\bullet} \in \mathbb{T}_{t,d}^{\geq}$, $g_{\bullet} \in \text{dom}(T_{\bullet,\bullet}^{\dagger}) = \text{ran}(T_{\bullet,\bullet}) \subseteq \ell_2$ and $\theta_{\bullet} = T_{\bullet,\bullet}^{\dagger} g_{\bullet} = T_{\bullet,\bullet}^{-1} g_{\bullet} \in \ell_2^{\alpha,r}$ be satisfied. Given $m \in \mathbb{N}$ denoting by $\theta_{\bullet}^m = T_{\bullet,\bullet}^{m\dagger} g_{\bullet} \in \ell_2 \mathbb{1}^m$ a Galerkin solution for any $s \in [0, 1]$ we obtain*

$$|\phi_{\mathcal{N}}(\theta_{\bullet}^m - \theta_{\bullet})|^2 \leq 3d^3(3d^3 + 1) \|\mathbb{1}^{m\perp} \theta_{\bullet}\|_{\alpha^{-1}}^2 \left(\|\mathbf{a} \mathbb{1}^{m\perp}\|_{\phi}^2 + (\mathbf{a}t^s)_{(m)}^2 \|\mathbf{t}^{-s} \mathbb{1}^m\|_{\phi}^2 \right). \quad (05.04)$$

§05101.35 **Proof of Lemma §05101.34.** Given in the lecture. \square

§05101.36 **Lemma.** *For each $m \in \mathbb{N}$ denote $\text{bias}_m^2 := \|\mathbf{a} \mathbb{1}^{m\perp}\|_{\phi}^2 + (\mathbf{a}t^s)_{(m)}^2 \|\mathbf{t}^{-s} \mathbb{1}^m\|_{\phi}^2$. If $(\mathbf{a}\phi)_{\bullet} \in \ell_2$ and $(\mathbf{a}t^s)_{\bullet} \in (\mathbb{R}_{>0})_{j_0}^{\mathbb{N}}$ then it follows $\text{bias}_{\bullet}^2 \in (\mathbb{R}_{>0})_{j_0}^{\mathbb{N}}$.*

§05101.37 **Proof of Lemma §05101.36.** Given in the lecture. \square

§05|02 Generalised linear Galerkin approach

§05102.01 **Generalised linear Galerkin approach.** Given $T_{\bullet,\bullet} \in \mathbb{L}(\ell_2)$ and $g_{\bullet} \in \ell_2$ any element $\theta_{\bullet}^m \in \ell_2 \mathbb{1}^m$ satisfying $T_{\bullet,\bullet} \theta_{\bullet}^m = \mathbb{1}^m g_{\bullet}$, i.e., $[T_{\bullet,\bullet}]_m [\theta_{\bullet}^m]_m = [g_{\bullet}]_m$, is called a *generalised Galerkin solution*. \square

§05102.02 **Notation.** We denote by $\mathbb{L}^{\text{reg}}(\ell_2)$ the subset of all injective $A_{\bullet,\bullet} \in \mathbb{L}(\ell_2)$ such that $[A_{\bullet,\bullet}]_m \in \mathbb{R}^{(m,m)}$ is regular for all $m \in \mathbb{N}$. For each $m \in \mathbb{N}$ and $A_{\bullet,\bullet} \in \mathbb{L}(\ell_2)$, the inverse $[A_{\bullet,\bullet}]_m^{-1} \in \mathbb{R}^{(m,m)}$ of $[A_{\bullet,\bullet}]_m \in \mathbb{R}^{(m,m)}$ exists. Note that $\mathbb{L}^{\text{reg}}(\ell_2) \subseteq \mathbb{L}(\ell_2)$ (Lemma §05101.22). \square

§05102.03 **Assumption.** For $\mathbb{J} = \ell_2$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ and $V \in \mathbb{L}(\mathbb{G}, \ell_2)$ fixed and presumed to be *known* in advance, the operator $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ satisfies $T_{\bullet,\bullet} = VTU^* \in \mathbb{L}^{\text{reg}}(\ell_2) \subseteq \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$. Let $g_{\bullet} \in \text{dom}(T_{\bullet,\bullet}^{\dagger}) = \text{ran}(T_{\bullet,\bullet})$, and hence $\theta_{\bullet} = T_{\bullet,\bullet}^{\dagger} g_{\bullet} = T_{\bullet,\bullet}^{-1} g_{\bullet} \in \ell_2$. \square

§05102.04 **Remark.** We consider a generalised linear Galerkin approach under Assumption §05102.03, i.e. $[T_{\bullet,\bullet}]_m$ is assumed to be regular for each $m \in \mathbb{N}$, so that $[T_{\bullet,\bullet}]_m^{-1}$ always exists. We shall emphasise that it is a non-trivial problem to determine when such an assumption holds (cf. Efromovich and Koltchinskii [2001] and references therein). However, if $[T_{\bullet,\bullet}]_m$ is regular, then for each $g_{\bullet} \in \ell_2$ the *generalised Galerkin solution* $\theta_{\bullet}^m = T_{\bullet,\bullet}^{m\dagger} g_{\bullet} \in \ell_2 \mathbb{1}^m$ is by $[\theta_{\bullet}^m]_m = [T_{\bullet,\bullet}]_m^{-1} [g_{\bullet}]_m$ uniquely determined. \square

§05102.05 **Generalised link condition.** Given weights $\mathbf{t}_{\bullet} \in (\mathbb{R}_{>0})^{\mathbb{N}} \cap \ell_{\infty}$ an operator $T_{\bullet,\bullet} \in \mathbb{L}(\ell_2)$ satisfies a *generalised link condition* if there exist $D \in \mathbb{R}_{\geq 1}$ and $d \in [1, D]$ such that

$$T_{\bullet,\bullet} \in \mathbb{T}_{t,d} := \left\{ T_{\bullet,\bullet} \in \mathbb{T}_{t,d} \cap \mathbb{L}(\ell_2) : \|[M_{\mathbf{t}}]_m [T_{\bullet,\bullet}]_m^{-1}\|_{\text{spec}} = \|[T_{\bullet,\bullet}]_m^{-1} [M_{\mathbf{t}}]_m\|_{\text{spec}} \leq D \text{ for all } m \in \mathbb{N} \right\}. \quad \square$$

§05102.06 **Remark.** We shall emphasise that $\mathbb{T}_{t,d}$ contains the subset $\mathbb{M}_{t,d}$ of diagonal operator satisfying the link condition, i.e. $\mathbb{M}_{t,d} \subseteq \mathbb{T}_{t,d}$ (see Remark §05101.09). Indeed, any $M_{\mathbf{w}} \in \mathbb{M}_{t,d}$ satisfies $\|[M_{\mathbf{t}}]_m [M_{\mathbf{w}}]_m^{-1}\|_{\text{spec}} = \|\mathbf{t} \cdot \mathbf{w}^{-1} \mathbb{1}^m\|_{\ell_{\infty}} \leq d \leq D$. Moreover, we have $\mathbb{T}_{t,d}^{\geq} \subseteq \mathbb{T}_{t,d}$ whenever $D \geq 3d^2$ due to Lemma §05101.22 (ii). The link condition $T_{\bullet,\bullet} \in \mathbb{T}_{t,d}$ or in equal $(T_{\bullet,\bullet}^* T_{\bullet,\bullet})^{1/2} \in \mathbb{T}_{t,d}^{\geq}$ does not depend on an unitary V , i.e. $V^*V = \text{id}_{\mathbb{G}}$, (or more generally surjective partial isometry with $\text{ran}(T) \subseteq \overline{\text{ran}}(V^*)$ implying $T^*V^*VT = T^*T$) since for each $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ with $VTU^* = T_{\bullet,\bullet}$ we have $T_{\bullet,\bullet}^* T_{\bullet,\bullet} = UT^*V^*VTU^* = UT^*TU^*$. The general link condition Definition §05102.05 however involves both surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ and $V \in \mathbb{L}(\mathbb{G}, \ell_2)$. It is worth pointing out, that for each $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ and surjective partial isometry $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ satisfying $(UT^*TU^*)^{1/2} \in \mathbb{T}_{t,d}^{\geq}$ we can theoretically construct a surjective partial isometry $V \in \mathbb{L}(\mathbb{G}, \ell_2)$ such that $VTU^* = T_{\bullet,\bullet} \in \mathbb{L}(\ell_2)$ satisfies $T_{\bullet,\bullet}^* T_{\bullet,\bullet} = UT^*TU^*$ and $T_{\bullet,\bullet}^* = UT^*V^* \in \mathbb{T}_{t,d}^{\geq}$. Consequently, from Lemma §05101.22 (ii) it follows $\|[M_{\mathbf{t}}]_m [T_{\bullet,\bullet}]_m^{-1}\|_{\text{spec}} = \|[T_{\bullet,\bullet}]_m^{-1} [M_{\mathbf{t}}]_m\|_{\text{spec}} \leq 3d^2$ for each $m \in \mathbb{N}$, which implies $T_{\bullet,\bullet} \in \mathbb{T}_{t,d}$ for all $D \geq 3d^2$. The fundamental inequality of Heinz [1951] in

Property §05101.10 implies $\|(\mathbf{UT}^*\mathbf{TU}^*)^{-1/2}\mathbf{1}_{\cdot}^{[j]}\|_{\ell_2} \leq \mathbf{dt}_j^{-1} \in \mathbb{R}_{>0}$ for each $j \in \mathbb{N}$. Thereby, the sequence $(\mathbf{UT}^*\mathbf{TU}^*)^{-1/2}\mathbf{1}_{\cdot}^{[j]}$ is an element of ℓ_2 and, hence $\mathbf{v}_j := \mathbf{TU}^*(\mathbf{UT}^*\mathbf{TU}^*)^{-1/2}\mathbf{1}_{\cdot}^{[j]}$, $j \in \mathbb{N}$ belongs to \mathbb{G} . Then it is easily checked that $(\mathbf{v}_j)_{j \in \mathbb{N}}$ is an *orthonormal sequence* in \mathbb{G} which determines a surjective partial isometry $\mathbf{V} \in \mathbb{L}(\mathbb{G}, \ell_2)$ (**Notation** §01102.07). By construction we have $\mathbf{T}_{\cdot}^* = \mathbf{UT}^*\mathbf{V}^* = \mathbf{UT}^*\mathbf{TU}^*(\mathbf{UT}^*\mathbf{TU}^*)^{-1/2} = (\mathbf{UT}^*\mathbf{TU}^*)^{1/2} \in \mathbb{L}(\ell_2)$, hence $\mathbf{T}_{\cdot} = (\mathbf{UT}^*\mathbf{TU}^*)^{1/2} \in \mathbb{L}(\ell_2)$, and thus $\mathbf{T}_{\cdot} \in \mathbb{T}_{\mathbf{t}, \mathbf{d}}^{\geq}$ or in equal $(\mathbf{T}_{\cdot}, \mathbf{T}_{\cdot}^*)^{1/2} = (\mathbf{T}_{\cdot}^*, \mathbf{T}_{\cdot})^{1/2} = (\mathbf{UT}^*\mathbf{TU}^*)^{1/2} \in \mathbb{T}_{\mathbf{t}, \mathbf{d}}^{\geq}$ and $\mathbf{T}_{\cdot} \in \mathbb{T}_{\mathbf{t}, \mathbf{d}}$. \square

§05102.07 **Property.** If $\mathbf{T}_{\cdot}, \mathbf{T}_{\cdot}^* \in \mathbb{T}_{\mathbf{t}, \mathbf{d}}$ then also $\mathbf{T}_{\cdot} \in \mathbb{T}_{\mathbf{t}, \mathbf{d}, \mathbf{D}}$ for each $\mathbf{D} \geq 3\mathbf{d}^2$. (!) \square

(!) Wieso ??

§05|02|01 Global and maximal global \mathbf{v} -error

We shall measure first globally the accuracy of the Galerkin solution $\theta^m \in \ell_2 \mathbf{1}_{\cdot}^m$ of $\theta = \mathbf{T}_{\cdot}^{\dagger} g \in \ell_2$.

§05102.08 **Property (Global \mathbf{v} -error).** Consider $\mathbf{v}_{\cdot} \in (\mathbb{R}_{>0})^{\mathbb{N}}$, $\mathbf{T}_{\cdot} \in \mathbb{L}(\ell_2)$ and $g \in \text{dom}(\mathbf{T}_{\cdot}^{\dagger}) = \text{ran}(\mathbf{T}_{\cdot}) \subseteq \ell_2$ and hence $\theta = \mathbf{T}_{\cdot}^{\dagger} g = \mathbf{T}_{\cdot}^{-1} g \in \ell_2$. Given $m \in \mathbb{N}$ we have $\mathbf{v}_{\cdot}^2 \mathbf{1}_{\cdot}^m \in \ell_{\infty}$ and hence $\ell_2 \mathbf{1}_{\cdot}^m \subseteq \ell_2(\mathbf{v}_{\cdot}^2)$. Consequently, denoting by $\theta^m = \mathbf{T}_{\cdot}^{m \dagger} g \in \ell_2 \mathbf{1}_{\cdot}^m$ a generalised Galerkin solution we have $\theta^m \in \ell_2(\mathbf{v}_{\cdot}^2)$ with

$$\|\theta^m\|_{\ell_2(\mathbf{v}_{\cdot})} \leq \|[\mathbf{M}_{\mathbf{v}_{\cdot}}]_{\mathbf{T}_{\cdot}}^{-1}\|_{\text{spec}} \|g\| \in \mathbb{R}_{>0}.$$

If $\mathbf{C}_{\mathbf{T}} := \sup \{ \|\mathbf{M}_{\mathbf{v}_{\cdot}} \mathbf{T}_{\cdot}^{m \dagger} \mathbf{T}_{\cdot} \mathbf{M}_{\mathbf{1}_{\cdot}^{m \perp}}\|_{\mathbb{L}(\ell_2)} : m \in \mathbb{N} \} \in \mathbb{R}_{>0}$ then

$$\|\theta^m - \theta\|_{\mathbf{v}_{\cdot}} \leq (1 + \mathbf{C}_{\mathbf{T}}) \|\mathbf{1}_{\cdot}^{m \perp} \theta\|_{\ell_2}$$

which implies $\sup \{ \|\theta^j - \theta\|_{\mathbf{v}_{\cdot}} : j \in \mathbb{N}_{\geq m} \} = o(1)$ as $m \rightarrow \infty$. \square

§05102.09 **Lemma (Maximal global \mathbf{v} -error).** Under Assumption §05101.26 let $\mathbf{T}_{\cdot} \in \mathbb{T}_{\mathbf{t}, \mathbf{d}, \mathbf{D}}$, $g \in \text{dom}(\mathbf{T}_{\cdot}^{\dagger}) = \text{ran}(\mathbf{T}_{\cdot}) \subseteq \ell_2$ and $\theta = \mathbf{T}_{\cdot}^{\dagger} g = \mathbf{T}_{\cdot}^{-1} g \in \ell_2^{\mathbf{a}, \mathbf{r}}$. Given $m \in \mathbb{N}$ denoting by $\theta^m = \mathbf{T}_{\cdot}^{m \dagger} g \in \ell_2 \mathbf{1}_{\cdot}^m$ a generalised Galerkin solution for any $s \in [0, 1]$ we obtain

$$\begin{aligned} \|\theta - \theta^m\|_{\mathbf{v}_{\cdot}}^2 &\leq (\mathbf{D}^2 \mathbf{d}^2 \mathbf{C}_{(\mathbf{v}_{\cdot})}^{-2} + 1) (\mathbf{a}\mathbf{v})_{(m)}^2 \|\mathbf{1}_{\cdot}^{m \perp} \theta\|_{\mathbf{a}^{-1}}^2, \quad \|\theta^m\|_{\mathbf{a}^{-1}} \leq \mathbf{D}\mathbf{d} \|\theta\|_{\mathbf{a}^{-1}}, \quad \text{and} \\ &\|\mathbf{T}_{\cdot}^s(\theta - \theta^m)\|_{\ell_2} \leq (\mathbf{D}\mathbf{d} + 1) \mathbf{d}^s (\mathbf{a}\mathbf{t}^s)_{(m)} \|\mathbf{1}_{\cdot}^{m \perp} \theta\|_{\mathbf{a}^{-1}}. \end{aligned} \quad (05.05)$$

§05102.10 **Proof of Lemma** §05102.09. Given in the lecture. \square

§05|02|02 Global and maximal global ϕ -error

Secondly we measure locally the accuracy of the generalised Galerkin solution $\theta^m \in \ell_2 \mathbf{1}_{\cdot}^m$ of $\theta = \mathbf{T}_{\cdot}^{\dagger} g \in \ell_2$.

§05102.11 **Reminder.** Given $\phi \in (\mathbb{R}_{>0})^{\mathbb{N}}$ for $\text{dom}(\phi_{\mathbb{N}}) := \{h_{\cdot} \in \ell_2 : \phi h_{\cdot} \in \ell_1\}$ we consider as in **Notation** §04102.16 the linear functional $\phi_{\mathbb{N}} : \ell_2 \supseteq \text{dom}(\phi_{\mathbb{N}}) \rightarrow \mathbb{R}$ defined by

$$h_{\cdot} \mapsto \phi_{\mathbb{N}}(h_{\cdot}) := \nu_{\mathbb{N}}(\phi h_{\cdot}) = \sum_{j \in \mathbb{N}} \phi_j h_j.$$

For each $\theta \in \text{dom}(\phi_{\mathbb{N}})$ and $m \in \mathbb{N}$ by **Property** §04102.18 we have $\theta \mathbf{1}_{\cdot}^m \in \text{dom}(\phi_{\mathbb{N}})$ with

$$|\phi_{\mathbb{N}}(\theta - \theta \mathbf{1}_{\cdot}^m)| \leq |\phi| \nu_{\mathbb{N}}(|\theta| \mathbf{1}_{\cdot}^{m \perp}) \leq \nu_{\mathbb{N}}(|\phi \theta|) \in \mathbb{R}_{>0},$$

and $|\phi_{\mathbb{N}}(\theta - \theta \mathbf{1}_{\cdot}^m)| |\phi_{\mathbb{N}}(\theta) - \phi_{\mathbb{N}}(\theta^m)| = o(1)$ as $m \rightarrow \infty$ by dominated convergence. \square

§05102.12 **Property (Local ϕ -error).** Consider $\phi \in (\mathbb{R}_{>0})^{\mathbb{N}}$, $T_{\cdot, \cdot} \in \mathbb{L}^{\geq}(\ell_2)$ and $g \in \text{dom}(T_{\cdot, \cdot}^{\dagger}) = \text{ran}(T_{\cdot, \cdot}) \subseteq \ell_2$ and hence $\theta = T_{\cdot, \cdot}^{\dagger} g = T_{\cdot, \cdot}^{-1} g \in \ell_2$. Given $m \in \mathbb{N}$ we have $\phi^2 \mathbf{1}^m \in \ell_2$ and hence $\ell_2 \mathbf{1}^m \subseteq \text{dom}(\phi \nu_{\kappa})$. Consequently, denoting by $\theta^m = T_{\cdot, \cdot}^{m \dagger} g \in \ell_2 \mathbf{1}^m$ a Galerkin solution we have $\theta^m \in \text{dom}(\phi \nu_{\kappa})$ with

$$\|\phi \theta^m\|_{\ell_1} \leq \| [T_{\cdot, \cdot}]_m^{-1} [\phi]_m \| \| [g]_m \| \in \mathbb{R}_{>0}.$$

If $C_T := \sup \{ \| M_{\mathbb{T}^{m \perp}} T_{\cdot, \cdot}^* (T_{\cdot, \cdot}^{m \dagger})^* \phi \|_{\ell_2} : m \in \mathbb{N} \} \in \mathbb{R}_{>0}$ then

$$|\phi \nu_{\kappa}(\theta^m - \theta)| \leq (1 + C_T) \|\mathbf{1}^{m \perp} \theta\|_{\ell_2}$$

which implies $\sup \{ |\phi \nu_{\kappa}(\theta^j - \theta)| : j \in \mathbb{N}_{\geq m} \} = o(1)$ as $m \rightarrow \infty$. \square

§05102.13 **Reminder.** Under Assumption §05101.32 we have $\mathbb{J}^a = \ell_2^a = \text{dom}(M_a) = \ell_2 \mathbf{a} = \ell_2(\mathbf{a}^{-2})$ and the three measures $\nu_{\mathbb{N}}$, $\mathbf{a}^{-2} \nu_{\mathbb{N}}$ and $|\phi| \nu_{\mathbb{N}}$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04101.02). Consequently, since $(\mathbf{a}\phi)_{\cdot} \in \ell_2$ and (**Property** §04102.23)

$$\|\phi h\|_{\ell_1} = \nu_{\mathbb{N}}(|h \mathbf{a}^{\dagger}(\mathbf{a}\phi)_{\cdot}|) \leq \|(\mathbf{a}\phi)_{\cdot}\|_{\ell_2} \|h\|_{\mathbf{a}^{-1}} \in \mathbb{R}_{>0} \quad \text{for each } h \in \ell_2^a$$

we have $\ell_2^a \subseteq \text{dom}(\phi \nu_{\kappa})$. Moreover, from $(\mathbf{a}\phi)_{\cdot} \in \ell_2$ follows $\|\mathbf{a} \mathbf{1}^{m \perp}\|_{\phi} = \|(\mathbf{a}\phi)_{\cdot} \mathbf{1}^{m \perp}\|_{\ell_2} = o(1)$ as $m \rightarrow \infty$. For $s \in [0, 1]$ from $(\mathbf{a}^s)_{\cdot} = \mathbf{a} \mathbf{t}^s \in (\mathbb{R}_{>0})_{\times}^{\mathbb{N}}$ follows $(\mathbf{a}^s)_{\cdot} = ((\mathbf{a}^s)_{(m)}) := (\mathbf{a}^s)_{m+1} = \|(\mathbf{a}^s)_{\cdot} \mathbf{1}^{m \perp}\|_{\ell_{\infty}}\|_{m \in \mathbb{N}} \in (\mathbb{R}_{>0})_{\times}^{\mathbb{N}}$. \square

§05102.14 **Lemma (Maximal local ϕ -error).** Under Assumption §05101.32 let $T \in \mathbb{T}_{\mathbf{t}, \mathbf{d}, \mathbf{D}}$, $g \in \text{dom}(T_{\cdot, \cdot}^{\dagger}) = \text{ran}(T_{\cdot, \cdot}) \subseteq \ell_2$ and $\theta = T_{\cdot, \cdot}^{\dagger} g = T_{\cdot, \cdot}^{-1} g \in \ell_2^{\mathbf{a}, \mathbf{r}}$. Given $m \in \mathbb{N}$ denoting by $\theta^m = T_{\cdot, \cdot}^{m \dagger} g \in \ell_2 \mathbf{1}^m$ a generalised Galerkin solution for any $s \in [0, 1]$ we obtain

$$|\phi \nu_{\kappa}(\theta^m - \theta)|^2 \leq (1 + \text{Dd}) \text{Dd} \|\mathbf{1}^{m \perp} \theta\|_{\mathbf{a}^{-1}}^2 (\|\mathbf{a} \mathbf{1}^{m \perp}\|_{\phi}^2 + (\mathbf{a}^s)_{(m)}^2 \|\mathbf{t}^{-s} \mathbf{1}^m\|_{\phi}^2). \quad (05.06)$$

§05102.15 **Proof of Lemma §05102.14.** Given in the lecture. \square

§06 Spectral regularisation

§06100.01 **Notation.** Consider the measure space $(\mathcal{J}, \mathcal{J}, \nu)$ and the Hilbert space $\mathbb{J} = \mathbb{L}_2(\nu)$ as in **Notation** §01101.01. We suppose that $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ are surjective partial isometries, hence $VV^* = \text{id}_{\mathbb{J}} = UU^*$. As in **Definition** §03100.08 we denote for $A := VTU^* \in \mathbb{L}(\mathbb{J})$ its Moore-Penrose inverse by $A^{\dagger} : \mathbb{J} \supseteq \text{dom}(A^{\dagger}) \rightarrow \mathbb{J}$. \square

§06100.02 **Comment.** In case the operator $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ is fixed and presumed to be *known* in advance, a spectral regularisation is formally not restricted to the diagonal or non-diagonal case as considered in **Subsection** §01104 and **Subsection** §01105, respectively. Consequently, we use in this section the symbol $A := VTU^* \in \mathbb{L}(\mathbb{J})$. However, in case of a noisy operator we will restrict ourselves to the diagonal and non-diagonal case introduced in **Definition** §02104.03 and **Definition** §02102.03. \square

§06100.03 **Assumption.** For $\mathbb{J} = \mathbb{L}_2(\nu)$ let $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ be surjective partial isometries fixed and presumed to be *known* in advance, let $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$, hence $A = VTU^* \in \mathbb{L}(\mathbb{J})$ with Moore-Penrose inverse $A^{\dagger} : \mathbb{J} \supseteq \text{dom}(A^{\dagger}) \rightarrow \mathbb{J}$ and let $g \in \text{dom}(A^{\dagger})$, and hence $\theta = A^{\dagger} g \in \mathbb{J}$. \square

§06100.04 **Definition.** A collection $\{R_{\alpha} \in \mathbb{L}(\mathbb{J}) : \alpha \in (0, 1)\}$ of operators is called *regularisation* of A^{\dagger} if for any $g \in \text{dom}(A^{\dagger})$ holds $\|R_{\alpha} g - A^{\dagger} g\|_{\mathbb{J}} \rightarrow 0$ as $\alpha \rightarrow 0$. \square

§06100.05 **Remark.** If A^{\dagger} is not bounded, then we have $\|R_{\alpha}\|_{\mathbb{L}(\mathbb{J})} \rightarrow \infty$ as $\alpha \rightarrow 0$. However, for $g \in \text{dom}(A^{\dagger})$ if $(g^n)_{n \in \mathbb{N}}$ is a sequence in \mathbb{J} such that $\|g^n - g\|_{\mathbb{J}} \leq n^{-1}$ for all $n \in \mathbb{N}$, then there exists a sequence $(\alpha_n)_{n \in \mathbb{N}}$ in $(0, 1)$ such that $\|R_{\alpha_n} g^n - A^{\dagger} g\|_{\mathbb{J}} = o(1)$ as $n \rightarrow \infty$. \square

§06|01 (Generalised) Tikhonov regularisation

§06|01.01 **Definition.** The collection $\{A^{\alpha-1}A^* \in \mathbb{L}(\mathbb{J}): A^{\alpha-1} := (A^*A + \alpha \text{id}_{\mathbb{J}})^{-1} \in \mathbb{L}^{\tilde{z}}(\mathbb{J}), \alpha \in (0, 1)\}$ of operators is called *Tikhonov regularisation* of $A^\dagger: \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$. □

§06|01.02 **Remark.** Given $A \in \mathbb{L}(\mathbb{J})$ consider for each $\alpha \in (0, 1)$ the strictly positive definite operator $A^\alpha = A^*A + \alpha \text{id}_{\mathbb{J}} \in \mathbb{L}^{\tilde{z}}(\mathbb{J})$ where

$$(\|A\|_{\mathbb{L}(\mathbb{J})}^2 + \alpha)\|h\|_{\mathbb{J}}^2 \geq \|A^\alpha h\|_{\mathbb{J}}\|h\|_{\mathbb{J}} \geq \langle A^\alpha h, h \rangle_{\mathbb{J}} \geq \alpha\|h\|_{\mathbb{J}}^2 \in \mathbb{R}_{>0} \tag{06.01}$$

for any $h \in \mathbb{J}_0 = \mathbb{J} \setminus \{0\}$ by applying the Cauchy-Schwarz inequality and, hence

$$\inf \{ \|A^\alpha h\|_{\mathbb{J}}: \|h\|_{\mathbb{J}} = 1, h \in \mathbb{J} \} \geq \alpha \in \mathbb{R}_{>0}. \tag{06.02}$$

Using the last bound $A^\alpha \in \mathbb{L}^{\tilde{z}}(\mathbb{J})$ has a closed range $\text{ran}(A^\alpha)$. Indeed, if $(A^\alpha a_n)_{j \in \mathbb{N}}$ converges, say to $g \in \mathbb{J}$, then $(A^\alpha h_n^j)_{j \in \mathbb{N}}$ is a Cauchy sequence and also $(h_n^j)_{j \in \mathbb{N}}$ by (06.01). Since \mathbb{J} is complete, $(h_n^j)_{j \in \mathbb{N}}$ converges, say to $h \in \mathbb{J}$. Since A^α is continuous, $(A^\alpha h_n^j)_{j \in \mathbb{N}}$ converges to $A^\alpha h = g$. In other words the range is closed. Since $A^\alpha \in \mathbb{L}(\ell_2)$ is injective with closed range it follows $\text{ran}(A^\alpha) = \ker(A^\alpha)^\perp = \mathbb{J}$, which in turn implies A^α is invertible, and due to the open mapping theorem with inverse $A^{\alpha-1} = (A^\alpha)^{-1} \in \mathbb{L}(\mathbb{J})$. Moreover, exploiting $\text{ran}(A^\alpha) = \mathbb{J}$ and (06.02) we have $\|A^{\alpha-1}\|_{\mathbb{L}(\ell_2)} \leq \alpha^{-1}$ since

$$\begin{aligned} \|A^{\alpha-1}\|_{\mathbb{L}(\ell_2)} &= \sup \{ \|A^{\alpha-1}g\|_{\mathbb{J}}: g \in \mathbb{J}, \|g\|_{\mathbb{J}} = 1 \} = \sup \left\{ \frac{\|A^{\alpha-1}g\|_{\mathbb{J}}}{\|g\|_{\mathbb{J}}}: g \in \mathbb{J}_0 = \text{ran}(A^\alpha) \setminus \{0\} \right\} \\ &= \sup \left\{ \frac{\|h\|_{\mathbb{J}}}{\|A^\alpha h\|_{\mathbb{J}}}: h \in \mathbb{J}_0 \right\} = \sup \{ \|A^\alpha h\|_{\mathbb{J}}^{-1}: h \in \mathbb{J}, \|h\|_{\mathbb{J}} = 1 \} \leq \alpha^{-1}. \end{aligned}$$

Consequently, the collection $\{A^{\alpha-1}A^* = (A^*A + \alpha \text{id}_{\mathbb{J}})^{-1}A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ is well-defined. □

§06|01.03 **Lemma.** For each $h \in \ker(A)^\perp$ holds $\|\alpha(A^*A + \alpha \text{id}_{\mathbb{J}})^{-1}h\|_{\mathbb{J}} = o(1)$ as $\alpha \rightarrow 0$.

§06|01.04 **Proof** of Lemma §06|01.03. Given in the lecture. □

§06|01.05 **Remark.** Let $g \in \text{dom}(A^\dagger)$, $\theta = A^\dagger g \in \mathbb{J}$ and $\theta^\alpha := A^{\alpha-1}A^*g \in \mathbb{J}$ we have

$$A^\alpha(\theta - \theta^\alpha) = A^*AA^\dagger g + \alpha\theta - A^\alpha A^{\alpha-1}A^*g = A^*g + \alpha\theta - A^*g = \alpha\theta,$$

and rewriting the last identity $A^{\alpha-1}A^*g - A^\dagger g = -\alpha A^{\alpha-1}\theta$. Consequently, from Lemma §06|01.03 follows $\|A^{\alpha-1}A^*g - A^\dagger g\|_{\mathbb{J}} = o(1)$ as $\alpha \rightarrow 0$ since $\theta = A^\dagger g \in \mathbb{J}$. Thereby, the Tikhonov collection as in Definition §06|01.01 is indeed a regularisation in the sense of Definition §06|00.04. □

§06|01.06 **Lemma.** For each $C \in \mathbb{L}(\mathbb{J})$ the following statements are equivalent:

- (i) θ^α minimises the *generalised Tikhonov functional* $h \mapsto F_\alpha(h) := \frac{1}{2}\|g - Ah\|_{\mathbb{J}}^2 + \frac{\alpha}{2}\|Ch\|_{\mathbb{J}}^2$
- (ii) θ^α is solution of the normal equation: $A^*g = (A^*A + \alpha C^*C)\theta^\alpha$.

§06|01.07 **Proof** of Lemma §06|01.06. Given in the lecture. □

§06|01.08 **Remark.** Observe that $\ker(A) \cap \ker(C) = \ker(A^*A + \alpha C^*C)$ which in turn implies, that the solution of the generalised Tikhonov functional, if it exists, is unique if and only if $\ker(A) \cap \ker(C) = \{0\}$. Recall that there exists a solution, for example, if $(A^*A + \alpha C^*C)$ has a continuous inverse. □

§06101.09 **Corollary.** Given the Tikhonov regularisation $\{A^{\alpha|^{-1}}A^* = (A^*A + \alpha \text{id}_{\mathbb{J}})^{-1}A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ as in [Definition §06101.01](#) for each $g \in \mathbb{J}$, $\theta^\alpha := A^{\alpha|^{-1}}A^*g \in \mathbb{J}$ is the unique minimiser in \mathbb{J} of the Tikhonov functional $h_* \mapsto \frac{1}{2}\|g - Ah_*\|_{\mathbb{J}}^2 + \frac{\alpha}{2}\|h_*\|_{\mathbb{J}}^2$. \square

§06101.10 **Proof of Corollary §06101.09.** Given in the lecture. \square

§06101.11 **Definition.** Given an operator $C \in \mathbb{L}(\mathbb{J})$ satisfying [\(gTR1\)](#) $\text{ran}(C)$ is closed and [\(gTR2\)](#) there exists $c \in \mathbb{R}_{>0}$ such that for any $h_* \in \ker(C)$ it holds $\|Ah_*\|_{\mathbb{J}} \geq c\|h_*\|_{\mathbb{J}}$, the collection

$$\{\text{gTR}_\alpha := (A^*A + \alpha C^*C)^{-1}A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$$

is called *generalised Tikhonov regularisation* of A^\dagger . \square

§06101.12 **Remark.** Assumption [\(gTR1\)](#) and [\(gTR2\)](#) ensure together that the generalised Tikhonov regularisation is well-defined. More precisely, introduce inner products $\langle \cdot, \cdot \rangle_* := \langle A \cdot, A \cdot \rangle_{\mathbb{J}} + \langle C \cdot, C \cdot \rangle_{\mathbb{J}}$ and $\langle \cdot, \cdot \rangle_C := \langle \cdot, \cdot \rangle_{\mathbb{J}} + \langle C \cdot, C \cdot \rangle_{\mathbb{J}}$ on \mathbb{J} with associated norms $\|\cdot\|_*$ and $\|\cdot\|_C$. Since \mathbb{J} is complete with respect to both norms (due to [\(gTR1\)](#) and [\(gTR2\)](#)), it follows from ?? that $\|\cdot\|_*$ and $\|\cdot\|_C$ are equivalent (keeping in mind that $\|h_*\|_*^2 \leq \max(\|A\|_{\mathbb{L}(\mathbb{J})}^2, 1)\|h_*\|_C^2$). Consequently, there is $K > 0$ such that $\|h_*\|_* \geq K\|h_*\|_C$ and thus $\|Ah_*\|_{\mathbb{J}}^2 + \|Ch_*\|_{\mathbb{J}}^2 \geq K^2(\|h_*\|_{\mathbb{J}}^2 + \|Ch_*\|_{\mathbb{J}}^2)$. Exploiting the last inequality we obtain $\|A^*Ah_* + \alpha C^*Ch_*\|_{\mathbb{J}} \geq K^2 \min(1, \alpha)\|h_*\|_{\mathbb{J}}$ for any $h_* \in \mathbb{J}$. In analogy to the arguments exploiting [\(06.01\)](#) in [Remark §06101.02](#), $A^*A + \alpha C^*C$ is injective with closed range and, thus it has a continuous inverse, i.e., $(A^*A + \alpha C^*C)^{-1} \in \mathbb{L}(\mathbb{J})$. Consequently, the generalised Tikhonov regularisation $\{\text{gTR}_\alpha := (A^*A + \alpha C^*C)^{-1}A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ is well-defined. Moreover, keeping in mind [Lemma §06101.06](#) $\theta^\alpha := \text{gTR}_\alpha g \in \mathbb{J}$ is obviously a solution of the normal equation, and thus the unique minimiser of the generalised Tikhonov functional. \square

§06101.13 **Corollary.** Consider the generalised Tikhonov regularisation as in [Definition §06101.11](#). For each $g \in \mathbb{J}$, $\theta^\alpha := \text{gTR}_\alpha g = (A^*A + \alpha C^*C)^{-1}A^*g$ is the unique minimiser in \mathbb{J} of the *generalised Tikhonov functional* $h_* \mapsto \frac{1}{2}\|g - Ah_*\|_{\mathbb{J}}^2 + \frac{\alpha}{2}\|Ch_*\|_{\mathbb{J}}^2$. \square

§06101.14 **Proof of Corollary §06101.13.** Given in the lecture. \square

§06101.15 **Remark.** Introduce further the adjoint A_*^* and C_*^* of A and C , respectively, with respect to the inner product $\langle \cdot, \cdot \rangle_*$ introduced in [Remark §06101.12](#), i.e., $\langle Ah_*, g \rangle_{\mathbb{J}} = \langle h_*, A_*^*g \rangle_*$ and $\langle Ch_*, g \rangle_{\mathbb{J}} = \langle h_*, C_*^*g \rangle_*$ for all $h_*, g \in \mathbb{J}$. In particular, for each $g, h_* \in \mathbb{J}$ we have [\(a\)](#) $A_*^*g = (A^*A + C^*C)^{-1}A^*g$, [\(b\)](#) $C_*^*g = (A^*A + C^*C)^{-1}C^*g$ and [\(c\)](#) $(A_*^*A + C_*^*C)h_* = h_*$ (i.e., $A_*^*A + C_*^*C = \text{id}_{\mathbb{J}}$). We note that $\ker(A_*^*) = \ker(A^*)$ and $\overline{\text{ran}}(A_*^*) = \ker(A)^{\perp*}$ where $\ker(A)^{\perp*}$ denotes the orthogonal complement of $\ker(A)$ in $(\mathbb{J}, \langle \cdot, \cdot \rangle_*)$. \square

Consider the restriction of A as bijective map from $\ker(A)^{\perp*}$ to $\text{ran}(A)$ and denote its inverse by $A_*^{-1} : \text{ran}(A) \rightarrow \ker(A)^{\perp*}$. Given the orthogonal projection $\Pi_{\overline{\text{ran}}(A)}$ onto $\overline{\text{ran}}(A)$ its associated Moore-Penrose inverse A_*^\dagger (see [Definition §03100.08](#)) defined on $\text{dom}(A_*^\dagger) = \text{ran}(A) \oplus \text{ran}(A)^\perp = \text{dom}(A^\dagger)$ is given by $A_*^\dagger := A_*^{-1}\Pi_{\overline{\text{ran}}(A)}$.

§06101.16 **Proposition.** Consider the generalised Tikhonov regularisation $\{\text{gTR}_\alpha \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ as in [Definition §06101.11](#). Under the conditions [\(gTR1\)](#) and [\(gTR2\)](#) of [Definition §06101.11](#) for $g \in \mathbb{J}$ and $\theta^\alpha = \text{gTR}_\alpha g = (A^*A + \alpha C^*C)^{-1}A^*g \in \mathbb{J}$ the following statements are equivalent:

- (i) $g \in \text{dom}(A_*^\dagger) = \text{ran}(A) \oplus \text{ran}(A)^\perp = \text{dom}(A^\dagger)$;
- (ii) there is $\theta^* \in \mathbb{J}$ such that $\|\theta^\alpha - \theta^*\|_* = o(1)$ as $\alpha \rightarrow 0$.

Moreover, under the equivalent conditions we have $\theta^* = A_*^\dagger g$.

§06101.17 **Proof** of **Proposition** §06101.16. Given in the lecture. \square

§06101.18 **Remark.** Due to **Proposition** §06101.16 the generalised Tikhonov regularisation as in **Definition** §06101.11 is indeed a regularisation in the sense of **Definition** §06100.04. Moreover, we shall emphasise that $\|\theta^\alpha - \theta^*\|_* = o(1)$ if and only if $\|A\theta^\alpha - A\theta^*\|_{\mathbb{J}} = o(1)$ and $\|C\theta^\alpha - C\theta^*\|_{\mathbb{J}} = o(1)$, which in turn implies $\|\theta^\alpha - \theta^*\|_{\mathbb{J}} = o(1)$. Keep further in mind that $A_*^*g_* = A_*^*A\theta$ holds if and only if $A^*g = A^*A\theta$ is true, since $A^*A + C^*C$ is continuously invertible. Thereby, for each $g_* \in \text{dom}(A^\dagger)$ the set of least squares solution $A^{-1}(\Pi_{\text{ran}(A)}g_*)$ satisfies $A^{-1}(\Pi_{\text{ran}(A)}g_*) = \{h_* \in \mathbb{J}: A^*Ah_* = A^*g_*\} = \{h_* \in \mathbb{J}: A_*^*Ah_* = A_*^*g_*\} = \{\theta_*^*\} + \ker(A)$ with $\theta_*^* = A_*^\dagger g_*$. Each $\theta \in A^{-1}(\Pi_{\text{ran}(A)}g_*)$ can thus be written as $\theta = \theta_*^* + h_*$ for some $h_* \in \ker(A)$ with $\theta_*^* \in \ker(A)^{\perp*}$, and hence, $A\theta = A\theta_*^*$ and $\|\theta^*\|_*^2 \leq \|\theta_*^*\|_*^2 + \|h_*\|_*^2 = \|\theta\|_*^2$, which together implies $\|C\theta_*^*\|_{\mathbb{J}}^2 \leq \|C\theta\|_{\mathbb{J}}^2$ for any $\theta \in A^{-1}(\Pi_{\text{ran}(A)}g_*)$. In other words, θ_*^* is the unique least squares solution with minimal $\|C \cdot\|_{\mathbb{J}}$ -norm. \square

§06|02 Spectral regularisation

§06102.01 **Definition.** For $A \in \mathbb{L}(\mathbb{J})$ let $\{r_\alpha, \alpha \in (0, 1)\}$ be a collection of real-valued Borel-measurable functions defined on $[0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$. The collection $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ of operators is called *spectral regularisation* of $A^\dagger : \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ if

(sR1) for all $\alpha \in (0, 1)$ there exists $C_\alpha \in \mathbb{R}_{>0}$ such that $|r_\alpha(x)| \leq C_\alpha$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$,

(sR2) for all $x \in (0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$ holds $|1 - xr_\alpha(x)| = o(1)$ as $\alpha \rightarrow 0$, and

(sR3) there is $K \in \mathbb{R}_{>0}$ such that $|xr_\alpha(x)| \leq K$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$ and $\alpha \in (0, 1)$. \square

§06102.02 **Proposition.** For $A \in \mathbb{L}(\mathbb{J})$ a spectral regularisation $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ as in **Definition** §06102.01 is a *regularisation* in the sense of **Definition** §06100.04.

§06102.03 **Proof** of **Proposition** §06102.02. Given in the lecture. \square

§06102.04 **Remark.** We shall emphasise that under (sR3) for any $g_* \notin \text{dom}(A^\dagger)$ it can be shown that $\|R_\alpha g_*\|_{\mathbb{J}} = \|r_\alpha(A^*A)A^*g_*\|_{\mathbb{J}} \rightarrow \infty$ as $\alpha \rightarrow 0$ (Engl et al. [2000], Theorem 4.1, p. 72). \square

§06|02|01 Maximal global v-error

Given $A \in \mathbb{L}(\mathbb{J})$ and a spectral regularisation $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ of $A^\dagger : \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ as in **Definition** §06102.01 for $g_* \in \text{dom}(A^\dagger)$ and $\alpha \in (0, 1)$ we shall measure globally the accuracy of the approximation $\theta^\alpha := R_\alpha g_* = r_\alpha(A^*A)A^*g_* \in \mathbb{J}$ of $\theta := A^\dagger g_* \in \mathbb{J}$.

§06102.05 **Source condition.** Given $A \in \mathbb{L}(\mathbb{J})$ and $g_* \in \text{dom}(A^\dagger)$, the solution $\theta = A^\dagger g_* \in \mathbb{J}$ satisfies a *source condition*, if there is $s \in \mathbb{R}_{>0}$ such that $\theta \in \text{ran}((A^*A)^{s/2})$, i.e. $\theta = (A^*A)^{s/2}h_*$ for $h_* \in \mathbb{J}$. \square

§06102.06 **Proposition.** Given $A \in \mathbb{L}(\mathbb{J})$ let $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ be a spectral regularisation of $A^\dagger : \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ as in **Definition** §06102.01. Assume **Definition** §06102.01 (sR1), and (sR3), and in addition replace (sR2) by

(sR2a) there is $s_* \in \mathbb{R}_{>1}$ such that for all $s \in [0, s_*]$ there is a constant $C_s \in \mathbb{R}_{>0}$ satisfying

$$\sup \{x^s |1 - xr_\alpha(x)| : x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]\} \leq C_s \alpha^s \quad \forall \alpha \in (0, 1).$$

For $g_* \in \text{dom}(A^\dagger)$ and $\alpha \in (0, 1)$ consider $\theta^\alpha = R_\alpha g_* = r_\alpha(A^*A)A^*g_* \in \mathbb{J}$ and $\theta := A^\dagger g_* \in \mathbb{J}$. If there are $s \in [0, 2s_*]$ and $h_* \in \mathbb{J}$ such that $\theta_* = (A^*A)^{s/2}h_*$ (i.e. $\theta_* \in \text{ran}((A^*A)^{s/2})$) satisfies a source condition as in **Definition** §06102.05), then we have

$$\|\theta^\alpha - \theta\|_{\mathbb{J}} \leq C_{s/2} \alpha^{s/2} \|h_*\|_{\mathbb{J}} \quad \forall \alpha \in (0, 1). \tag{06.03}$$

§06102.07 **Proof** of **Proposition** §06102.06. Given in the lecture. □

§06102.08 **Link condition.** Given weights $\mathfrak{t}_\bullet \in \mathcal{M}_{>0}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ an operator $A \in \mathbb{L}(\mathbb{J})$ satisfies a *link condition* if there is $d \in \mathbb{R}_{>1}$ such that

$$A \in \mathbb{T}_{\mathfrak{t}_\bullet, d}^{\geq} := \left\{ T \in \mathbb{L}(\mathbb{J}) : d^{-1} \|a_\bullet\|_{\mathfrak{t}_\bullet} \leq \|T a_\bullet\|_{\mathbb{J}} \leq d \|a_\bullet\|_{\mathfrak{t}_\bullet} \text{ for all } a_\bullet \in \mathbb{J} \right\}$$

and we set $\mathbb{T}_{\mathfrak{t}_\bullet, d} := \left\{ T \in \mathbb{L}(\mathbb{J}) : (T^* T)^{1/2} \in \mathbb{T}_{\mathfrak{t}_\bullet, d}^{\geq} \right\}$. □

§06102.09 **Property.** If $A \in \mathbb{T}_{\mathfrak{t}_\bullet, d}^{\geq}$ with $\mathfrak{t}_\bullet \in \mathcal{M}_{>0}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ and $d \in \mathbb{R}_{>1}$ then for all $s \in [-1, 1]$ we have

$$\text{(inequality of Heinz [1951]) } d^{-|s|} \|a_\bullet\|_{\mathfrak{t}_\bullet} \leq \|A^s a_\bullet\|_{\mathbb{J}} \leq d^{|s|} \|a_\bullet\|_{\mathfrak{t}_\bullet} \quad \text{for all } a_\bullet \in \text{dom}(M_\tau). \quad \square$$

§06102.10 **Comment.** Given $A \in \mathbb{T}_{\mathfrak{t}_\bullet, d}^{\geq}$ we have $\ker(A) = \{0\}$ and on $\text{ran}(A)$ (which is dense in \mathbb{J}) we have $A^{-1} = A^\dagger$. Similarly, for each $s \in \mathbb{R}_{>0}$ on $\text{ran}(A^s)$ we have $A^{-s} = A^{s\dagger} = (A^s)^\dagger$. □

§06102.11 **Assumption.** Consider $\mathfrak{v}_\bullet \in \mathcal{M}_{>0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$, and for $t \in \mathbb{R}_{>0}$, $a \in (0, t]$ set $\mathfrak{t}_\bullet := \mathfrak{v}_\bullet^t$ and $\mathfrak{a}_\bullet := \mathfrak{v}_\bullet^a$ where $\mathfrak{t}_\bullet, \mathfrak{a}_\bullet \in \mathcal{M}_{>0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ and hence $\nu(\mathcal{N}_{\mathfrak{v}_\bullet}) = \nu(\mathcal{N}_{\mathfrak{t}_\bullet}) = \nu(\mathcal{N}_{\mathfrak{a}_\bullet}) = 0$. □

§06102.12 **Reminder.** Under Assumption §06102.11 we have $\mathbb{J}^a = \mathbb{L}_2(\nu) = \text{dom}(M_{\mathfrak{a}_\bullet}) = \mathbb{J} \mathfrak{a}_\bullet = \mathbb{L}_2(\mathfrak{a}_\bullet^{-2} \nu)$ and the measures ν , $\mathfrak{v}_\bullet^2 \nu$, $\mathfrak{t}_\bullet^2 \nu$ and $\mathfrak{a}_\bullet^{-2} \nu$ dominate mutually each other (see **Property** §04101.02). Consequently, $\mathbb{J}^a \subseteq \mathbb{J} = \mathbb{L}_2(\nu)$ and $\mathbb{J}^a \subseteq \mathbb{L}_2(\mathfrak{v}_\bullet^2 \nu)$ (**Property** §04102.11) since $(\mathfrak{a} \mathfrak{v})_\bullet = \mathfrak{v}_\bullet^{1+a} \in \mathbb{L}_\infty(\nu)$. We assume in the following that $\theta_\bullet \in \mathbb{J}$ satisfies an abstract smoothness condition (**Definition** §04102.12), i.e., there is $r \in \mathbb{R}_{>0}$ such that $\theta_\bullet \in \mathbb{J}^{a, r} = \{h_\bullet \in \mathbb{J}^a : \|h_\bullet\|_{\mathfrak{a}_\bullet} \leq r\} \subseteq \mathbb{J}^a \subseteq \mathbb{J}$. Under Assumption §06102.11 by **Corollary** §05101.14 (see **Comment** §05101.16) if $A \in \mathbb{T}_{\mathfrak{t}_\bullet, d}$ (or in equal $(A^* A)^{1/2} \in \mathbb{T}_{\mathfrak{t}_\bullet, d}^{\geq}$) then (i) for any $\theta_\bullet \in \mathbb{J}^a$ we have $\theta_\bullet = (A^* A)^{a/(2t)} h_\bullet$ with $\|h_\bullet\|_{\mathbb{J}} \leq d^{a/t} \|\theta_\bullet\|_{\mathfrak{a}_\bullet}$, and conversely (ii) for any $\theta_\bullet = (A^* A)^{a/(2t)} h_\bullet$ with $h_\bullet \in \mathbb{L}_2(\nu)$ we obtain $\theta_\bullet \in \mathbb{J}^a$ with $\|\theta_\bullet\|_{\mathfrak{a}_\bullet} \leq d^{a/t} \|h_\bullet\|_{\mathbb{J}}$. □

§06102.13 **Corollary.** Let Assumption §06102.11 with $(\mathfrak{t} \mathfrak{a})_\bullet = \mathfrak{v}_\bullet^{1+a} \in \mathcal{M}_{>0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ and $d, r \in \mathbb{R}_{>0}$ be satisfied. If $A \in \mathbb{T}_{\mathfrak{t}_\bullet, d}$ and $\theta_\bullet \in \mathbb{J}^{a, r}$, then we have $g_\bullet = A \theta_\bullet \in \mathbb{J}^{(1+a), dr}$.

§06102.14 **Proof** of **Corollary** §06102.13. Given in the lecture. □

§06102.15 **Proposition.** Given $A \in \mathbb{L}(\mathbb{J})$ let $\{R_\alpha := r_\alpha(A^* A) A^* \in \mathbb{L}(\mathbb{J}) : \alpha \in (0, 1)\}$ be a spectral regularisation of $A^\dagger : \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ as in **Definition** §06102.01. Assume (sR1), (sR2), and (sR3) (**Definition** §06102.01) and (sR2a) (**Proposition** §06102.06). For $g_\bullet \in \text{dom}(A^\dagger)$ and $\alpha \in (0, 1)$ consider $\theta_\bullet^\alpha = R_\alpha g_\bullet = r_\alpha(A^* A) A^* g_\bullet \in \mathbb{J}$ and $\theta_\bullet := A^\dagger g_\bullet \in \mathbb{J}$. Under Assumption §06102.11 if $T \in \mathbb{T}_{\mathfrak{t}_\bullet, d}$ (link condition as in **Definition** §06102.08) and $\theta_\bullet \in \mathbb{J}^{a, r}$ (abstract smoothness condition as in **Definition** §04102.12), then for any $q \in [-a, t]$ we have

$$\|\theta_\bullet^\alpha - \theta_\bullet\|_{\mathfrak{v}_\bullet^q} \leq C_{(q+a)/(2t)} d^{(a+|q|)/t} r \alpha^{(a+q)/(2t)}, \quad \forall \alpha \in (0, 1). \quad (06.04)$$

§06102.16 **Proof** of **Proof** §06102.16. Given in the lecture. □

§06102.17 **Remark.** Let us briefly comment on the Assumption §06102.11 imposed in **Proposition** §06102.15. We set $\theta_\bullet^0 := \theta_\bullet$ and write $\{\theta_\bullet^\alpha : \alpha \in [0, 1]\} = \{\theta_\bullet\} \cup \{\theta_\bullet^\alpha = R_\alpha A \theta_\bullet = r_\alpha(A^* A) A^* A \theta_\bullet : \alpha \in (0, 1)\}$, shortly. Note that, under $q \geq -a$ the *global \mathfrak{v}^q -error* is well-defined on \mathbb{J}^a since $\{\theta_\bullet^\alpha : \alpha \in [0, 1]\} \subseteq \mathbb{L}_2(\mathfrak{v}_\bullet^{2q} \nu)$ for all $\theta_\bullet \in \mathbb{J}^a$. Moreover, the additional condition $q \leq t$ together with $a \leq t$ allows us to apply the inequality of Heinz [1951] **Property** §06102.09. We can dismiss those upper bounds, if A and $M_{\mathfrak{v}_\bullet}$ commute. However, if A and $M_{\mathfrak{v}_\bullet}$ do not commute, then the smallest upper bound of the global approximation bias is up to a constant α since $(a+q)/(2t) \in [0, 1]$. □

§06102.18 **Example.** Let us discuss certain spectral regularisations satisfying (sR1), (sR2a) and (sR3).

- (a) *Tikhonov regularisation* as defined in §06101.01 is given by $x \mapsto r_\alpha(x) = (x + \alpha)^{-1}$ and satisfies (sR1) and (sR3) with $C_\alpha = \alpha^{-1}$ and $K = 1$, and (sR2a) with $s_0 = 1$ and $C_s = s^s(1 - s)^{1-s}$.
- (b) *Spectral cut-off* given by the piecewise continuous function $x \mapsto r_\alpha(x) = \frac{1}{x} \mathbb{1}_{[\alpha, \infty)}(x)$ satisfies (sR1) and (sR3) with $C_\alpha = \alpha^{-1}$ and $K = 1$, and (sR2a) with $s_0 = \infty$ and $C_s = 1$.
- (c) A special iterative regularisation is the *Landweber iteration*. This method is based on a transformation of the normal equation into an equivalent fixed point equation $\theta = \theta + \omega A^*(g - A\theta)$ with $\omega \in (0, \|A\|_{\mathbb{L}(\mathbb{J})}^{-2}]$. Then the corresponding fixed point operator $\text{id}_{\mathbb{J}} - \omega A^*A$ is non-expansive and θ may be approximated by θ^m determined by $\theta^j := \theta^{j-1} + \omega A^*(g - A\theta^{j-1})$, $j \in \llbracket m \rrbracket$, and $\theta^0 := 0$. Note, that without loss of generality, we can assume $\|A\|_{\mathbb{L}(\mathbb{J})} \leq 1$ and drop the parameter ω . By induction the iterate θ^m can be expressed non-recursively through $\theta^m = \sum_{j \in \llbracket m \rrbracket} (\text{id}_{\mathbb{J}} - A^*A)^{j-1} A^*g$ and thus $x \mapsto r_{1/m}(x) = \sum_{j \in \llbracket m \rrbracket} (1 - x)^{j-1}$ where $1 - xr_{1/m}(x) = (1 - x)^m$. Under the assumption $\|A\|_{\mathbb{L}(\mathbb{J})} \leq 1$, the Landweber iteration is thus a spectral regularisation with $\alpha = 1/m$ satisfying (sR1) and (sR3) with $C_\alpha = \alpha^{-1}$ and $K = 1$. Moreover, (sR2a) holds with $s_0 = \infty$ and $C_s = s^s e^{-s}$. \square

§06102.19 **Notation.** Given $A \in \mathbb{L}^{\geq}(\mathbb{J})$, i.e., A is positive definite, we eventually consider a spectral regularisation $\{R_\alpha := r_\alpha(A) \in \mathbb{L}(\mathbb{J}) : \alpha \in (0, 1)\}$ of A^\dagger for a given collection $\{r_\alpha : \alpha \in (0, 1)\}$ of real-valued Borel-measurable functions defined on $[0, \|A\|_{\mathbb{L}(\mathbb{J})}]$ satisfying

- (sR1') for all $\alpha \in (0, 1)$ there exists $C_\alpha \in \mathbb{R}_{>0}$ such that $|r_\alpha(x)| \leq C_\alpha$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}]$,
- (sR2'a) there are $s_0 \in [1, \infty)$ and $C_s \in \mathbb{R}_{>0}$ for all $s \in [0, s_0]$ such that $x^s |1 - xr_\alpha(x)| \leq C_s \alpha^s$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}]$ and $\alpha \in (0, 1)$,
- (sR3') there is $K \in \mathbb{R}_{>0}$ such that $|xr_\alpha(x)| \leq K$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}]$ and $\alpha \in (0, 1)$.

We shall measure in the sequel the accuracy of the approximation $\theta^\alpha = R_\alpha g = r_\alpha(A)g \in \mathbb{J}$ of $\theta := A^\dagger g \in \mathbb{J}$ for $g \in \text{dom}(A^\dagger)$, by its global approximation error. For convenient notation we eventually use the notation $\theta^0 := \theta$ and write and write $\{\theta^\alpha : \alpha \in [0, 1)\} = \{\theta\} \cup \{\theta^\alpha = R_\alpha A\theta = r_\alpha(A)A\theta : \alpha \in (0, 1)\}$. \square

§06102.20 **Proposition.** Given $A \in \mathbb{L}^{\geq}(\mathbb{J})$ let $\{R_\alpha = r_\alpha(A) \in \mathbb{L}^{\geq}(\mathbb{J}) : \alpha \in (0, 1)\}$ be a spectral regularisation of A^\dagger satisfying (sR1'), (sR2'a) and (sR3') in Notation §06102.19. For $g \in \text{dom}(A^\dagger)$ and $\alpha \in (0, 1)$ consider $\theta^\alpha = R_\alpha g = r_\alpha(A)g \in \mathbb{J}$ and $\theta := A^\dagger g \in \mathbb{J}$.

- (i) If there are $s \in [0, s_0]$ and $h_s \in \mathbb{J}$ such that $\theta = A^s h_s$ (i.e. $\theta \in \text{ran}(A^s)$) satisfies a source condition as in Definition §06102.05), then we have

$$\|\theta^\alpha - \theta\|_{\mathbb{J}} \leq C_s \alpha^s \|h_s\|_{\mathbb{J}} \quad \forall \alpha \in (0, 1). \tag{06.05}$$

- (ii) Under Assumption §06102.11 if $\mathbb{T} \in \mathbb{T}_{t,d}^{\geq}$ (link condition as in Definition §06102.08) and $\theta \in \mathbb{J}^{a,t}$ (abstract smoothness condition as in Definition §04102.12), then for any $q \in [-a, t \wedge (ts_0 - a)]$ we have

$$\|\theta^\alpha - \theta\|_{\mathbb{V}^q} \leq C_{(q+a)/t} d^{(a+|q|)/t} r \alpha^{(a+q)/t} \quad \forall \alpha \in (0, 1). \tag{06.06}$$

§06102.21 **Proof of Proposition §06102.20.** Given in the lecture. \square

§06102.22 **Remark.** If (sR2'a) is satisfied for some $s_0 \geq 2$ (excluding the Tikhonov regularisation as discussed in Example §06102.18 (a)) then (06.06) in Proposition §06102.20 (ii) holds for any $q \in [-a, t]$ as in Proposition §06102.15. \square

§06|02|02 Maximal local ϕ -error

Given $A \in \mathbb{L}(\mathbb{J})$ and a spectral regularisation $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ of $A^\dagger : \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ as in [Definition §06|02.01](#) for $g \in \text{dom}(A^\dagger)$ and $\alpha \in (0, 1)$ we shall measure locally the accuracy of the approximation $\theta^\alpha := R_\alpha g = r_\alpha(A^*A)A^*g \in \mathbb{J}$ of $\theta := A^\dagger g \in \mathbb{J}$.

§06|02.23 **Reminder.** For $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\text{dom}(\phi\nu) := \{h \in \mathbb{J} = \mathbb{L}_2(\nu): \phi h \in \mathbb{L}_1(\nu)\}$ we consider the linear functional $\phi\nu : \mathbb{J} \supseteq \text{dom}(\phi\nu) \rightarrow \mathbb{R}$ given by $h \mapsto \phi\nu(h) := \nu(\phi h)$ with a slight abuse of notations. Under [Assumption §06|02.11](#) we have $\mathbb{J}^a = \mathbb{L}_2^a(\nu) = \text{dom}(M_{\tau^{-1}}) = \mathbb{J}a = \mathbb{L}_2(a^{-2}\nu)$ and the measures $\nu, \mathfrak{v}^2\nu, \phi^2\nu, \mathfrak{t}^2\nu$ and $a^{-2}\nu$ dominate mutually each other (see [Property §04|01.02](#)). Consequently, $\mathbb{J}^a \subseteq \mathbb{J} = \mathbb{L}_2(\nu)$ and $\mathbb{J}^a \subseteq \mathbb{L}_2(\mathfrak{v}^2\nu)$ ([Property §04|02.11](#)) since $(a\mathfrak{v}) \cdot = \mathfrak{v}^{1+a} \in \mathbb{L}_\infty(\nu)$. We assume in the following that $\theta \in \mathbb{J}^{a,r}$ and $A \in \mathbb{T}_{t,d}$ satisfies, respectively, an abstract smoothness condition ([Definition §04|02.12](#)) and link condition ([Definition §06|02.08](#)). Under [Assumption §06|02.11](#) due to [Proposition §06|02.15](#) we have $\theta^\alpha - \theta \in \mathbb{L}_2(\mathfrak{v}^{2q}\nu)$, and thus if in addition $\mathfrak{v}^{-q} \in \mathbb{L}_2(\phi^2\nu)$ also $\theta^\alpha - \theta \in \text{dom}(\phi\nu)$. \square

§06|02.24 **Proposition.** Given $A \in \mathbb{L}(\mathbb{J})$ let $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ be a spectral regularisation of $A^\dagger : \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ as in [Definition §06|02.01](#). Assume (sR1), (sR3) ([Definition §06|02.01](#)) and (sR2a) ([Proposition §06|02.06](#)). For $g \in \text{dom}(A^\dagger)$ and $\alpha \in (0, 1)$ consider $\theta^\alpha = R_\alpha g = r_\alpha(A^*A)A^*g \in \mathbb{J}$ and $\theta := A^\dagger g \in \mathbb{J}$. Under [Assumption §06|02.11](#) if $A \in \mathbb{T}_{t,d}$ (link condition) and $\theta \in \mathbb{J}^{a,r}$ (abstract smoothness condition), then for any $q \in [-a, t]$ such that $\mathfrak{v}^{-q} \in \mathbb{L}_2(\phi^2\nu)$ with $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ we have

$$|\phi\nu(\theta^\alpha - \theta)| \leq C_{(q+a)/(2t)} d^{(a+|q|)/t} r \|\mathfrak{v}^{-q}\|_\phi \alpha^{(a+q)/(2t)}, \quad \forall \alpha \in (0, 1). \quad (06.07)$$

§06|02.25 **Proof of Proposition §06|02.24.** Given in the lecture. \square

§06|02.26 **Proposition.** Given $A \in \mathbb{L}(\mathbb{J})$ let $\{R_\alpha = r_\alpha(A) \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ be a spectral regularisation of A^\dagger satisfying (sR1'), (sR2'a) and (sR3') in [Notation §06|02.19](#). For $g \in \text{dom}(A^\dagger)$ and $\alpha \in (0, 1)$ consider $\theta^\alpha = R_\alpha g = r_\alpha(A)g \in \mathbb{J}$ and $\theta := A^\dagger g \in \mathbb{J}$. Under [Assumption §06|02.11](#) if $T \in \mathbb{T}_{t,d}^\geq$ (link condition) and $\theta \in \mathbb{J}^{a,r}$ (abstract smoothness condition), then for any $q \in [-a, t \wedge (ts - a)]$ such that $\mathfrak{v}^{-q} \in \mathbb{L}_2(\phi^2\nu)$ with $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ we have

$$|\phi\nu(\theta^\alpha - \theta)| \leq C_{(q+a)/t} d^{(a+|q|)/t} r \|\mathfrak{v}^{-q}\|_\phi \alpha^{(a+q)/t} \quad \forall \alpha \in (0, 1). \quad (06.08)$$

§06|02.27 **Proof of Proposition §06|02.26.** Given in the lecture. \square

§06|02.28 **Remark.** If (sR2'a) is satisfied for some $s_s \geq 2$ (excluding the Tikhonov regularisation as discussed in [Example §06|02.18 \(a\)](#)) then [Proposition §06|02.26](#) holds for any $q \in [-a, t]$ as in [Proposition §06|02.24](#). \square

Chapter 3

Regularised estimation

Making use of the regularisation approaches presented in [Chapter 2](#) we introduce estimators of the solution $\theta \in \mathbb{H}$ based on a noisy observation of the image $g = T\theta$ and eventually in addition of the operator T .

Overview

§07	Orthogonal projection estimator	61
§07 01	Diagonal statistical inverse problem	62
§07 01 01	Examples	62
§07 01 02	Global and maximal global \mathfrak{v} -risk	64
§07 01 03	Local and maximal local ϕ -risk	69
§07 02	Diagonal statistical inverse problem with noisy operator	75
§07 02 01	Examples	75
§07 02 02	Global and maximal global \mathfrak{v} -risk	76
§07 02 03	Local and maximal local ϕ -risk	82
§08	(Generalised) Galerkin estimator	88
§08 01	Non-diagonal statistical inverse problem	89
§08 01 01	Examples	89
§08 01 02	Global and maximal global \mathfrak{v} -risk	90
§08 01 03	Local and maximal local ϕ -risk	95
§08 02	Non-diagonal statistical inverse problem with noisy operator	100
§08 02 01	Examples	102
§08 02 02	Global and maximal global \mathfrak{v} -risk	104
§08 02 03	Local and maximal local ϕ -risk	111
§09	Spectral regularisation estimator	118
§09 01	Statistical inverse problem	119
§09 01 01	Global risk	119
§09 01 02	Maximal global \mathfrak{v} -risk	121

§07 Orthogonal projection estimator

§07|00.01 **Notation** (§04|00.01 continued). Consider the measure space $(\mathcal{J}, \mathcal{J}, \nu)$ and the Hilbert space $\mathbb{J} = \mathbb{L}_2(\nu)$ as in [Notation §01|01.01](#). For $w_\bullet \in \mathbb{R}^{\mathcal{J}}$ define the multiplication map $M_{w_\bullet} : \mathbb{R}^{\mathcal{J}} \rightarrow \mathbb{R}^{\mathcal{J}}$ with $a_\bullet \mapsto M_{w_\bullet} a_\bullet := w_\bullet a_\bullet := (w_j a_j)_{j \in \mathcal{J}}$. If $w_\bullet \in \mathcal{M}(\mathcal{J})$, i.e. w_\bullet is \mathcal{J} - \mathcal{B} -measurable, then we have $M_{w_\bullet} : \mathcal{M}(\mathcal{J}) \rightarrow \mathcal{M}(\mathcal{J})$ too. If in addition $w_\bullet \in \mathbb{L}_\infty(\nu)$ then we have also $M_{w_\bullet} \in \mathbb{L}(\mathbb{J})$ identifying again equivalence classes and representatives. We set $\mathbb{L}^{\mathfrak{w}}(\mathbb{J}) := \{M_{w_\bullet} : w_\bullet \in \mathbb{L}_\infty(\nu)\} \subseteq \mathbb{L}(\mathbb{J})$ noting that $\|M_{w_\bullet}\|_{\mathbb{L}(\mathbb{J})} = \sup \{\|w_\bullet a_\bullet\|_{\mathbb{J}} : \|a_\bullet\|_{\mathbb{J}} \leq 1\} \leq \|w_\bullet\|_{\mathbb{L}_\infty(\nu)}$ for each $M_{w_\bullet} \in \mathbb{L}^{\mathfrak{w}}(\mathbb{J})$ (see [Notation §01|04.01](#)). Finally, given surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ we define $\mathbb{L}^{U,V}(\mathbb{L}^{\mathfrak{w}}(\mathbb{J})) := V^*(\mathbb{L}^{\mathfrak{w}}(\mathbb{J}))U := \{V^* M_{w_\bullet} U \in \mathbb{L}(\mathbb{H}, \mathbb{G}) : M_{w_\bullet} \in \mathbb{L}^{\mathfrak{w}}(\mathbb{J})\}$. As a consequence, for each $T \in \mathbb{L}^{U,V}(\mathbb{L}^{\mathfrak{w}}(\mathbb{J}))$ we have $VTU^* = M_{w_\bullet} \in \mathbb{L}^{\mathfrak{w}}(\mathbb{J})$ for some $w_\bullet \in \mathbb{L}_\infty(\nu)$. \square

§07100.02 **Assumption.** The separable Hilbert space $\mathbb{J} = \mathbb{L}_2(\mathcal{J}, \mathcal{F}, \nu)$ with σ -algebra \mathcal{F} over \mathcal{J} containing all elementary events $\{j\}$, $j \in \mathcal{J}$, and all events $\llbracket m \rrbracket := [-m, m] \cap \mathcal{J}$, $m \in \mathbb{N}$, and with σ -finite measure $\nu \in \mathcal{M}_\sigma(\mathcal{J})$ such that $\nu(\llbracket m \rrbracket) \in \mathbb{R}_{>0}$, for all $m \in \mathbb{N}$, and the surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ are fixed and presumed to be *known* in advance. The operator satisfies $T \in \mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J})) \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$ and hence $VTU^* = M_s \in \mathbb{L}(\mathbb{J})$ for some $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ and the image fulfils $g \in \text{dom}(M_s)$, and hence $\mathfrak{s}^\dagger g \in \mathbb{J} = \mathbb{L}_2(\nu)$. \square

§07100.03 **Reminder.** Under Assumption §07100.02 we consider $T \in \mathbb{L}^{\mathbb{U}, \mathbb{V}}(\mathbb{L}(\mathbb{J})) \subseteq \mathbb{L}(\mathbb{H}, \mathbb{G})$, and hence $VTU^* = M_s \in \mathbb{L}(\mathbb{J})$ and $g = M_s \theta = \mathfrak{s} \theta \in \mathbb{J}$ for some $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$. Due to **Property** §04101.02 the Moore-Penrose inverse of $M_s \in \mathbb{L}(\mathbb{J})$ satisfies $M_s^\dagger = M_{\mathfrak{s}^\dagger} : \mathbb{J} \supseteq \text{dom}(M_s) \rightarrow \mathbb{J}$ with $\text{dom}(M_s) = \mathbb{J} \mathfrak{s} \oplus \mathbb{J} \mathbb{1}^{\mathcal{N}_s} = \mathbb{J}^s$. For each $m \in \mathbb{N}$, $M_{\mathbb{1}^m} \in \mathbb{L}(\mathbb{J})$ and $M_{\mathbb{1}^{m\perp}} \in \mathbb{L}(\mathbb{J})$ is the *orthogonal projection* onto the linear subspace $\mathbb{J} \mathbb{1}^m \subseteq \mathbb{J}$ and its orthogonal complement $\mathbb{J} \mathbb{1}^{m\perp} = (\mathbb{J} \mathbb{1}^m)^\perp \subseteq \mathbb{J}$, respectively, that is $\mathbb{J} = \mathbb{J} \mathbb{1}^m \oplus \mathbb{J} \mathbb{1}^{m\perp}$ (see **Property** §04102.02). Given $g \in \mathbb{J}$ we call $\theta \in \mathbb{J}$ satisfying $\|g - \mathfrak{s} \theta\|_{\mathbb{J}} = \inf \{ \|g - \mathfrak{s} h\|_{\mathbb{J}} : h \in \mathbb{J} \}$ a least squares solution, if it exists (see **Property** §03100.05). Writing $\mathfrak{s}^\dagger = \mathfrak{s}^{-1} \mathbb{1}^{\mathcal{N}_s^c}$ and $\mathcal{N}_s = \{j \in \mathbb{N} : \mathfrak{s}_j \in \mathbb{R}_{>0}\}$ for each $g \in \text{dom}(M_s) = \mathbb{J} \mathfrak{s} \oplus \mathbb{J} \mathbb{1}^{\mathcal{N}_s}$ is $\theta = M_s g = \mathfrak{s}^\dagger g$ the unique least square solution with minimal $\|\cdot\|_{\mathbb{J}}$ -norm in the set $\mathfrak{s}^\dagger g + \mathbb{J} \mathbb{1}^{\mathcal{N}_s}$ of all least square solutions (**Property** §04103.02). If in addition $\nu(\mathcal{N}_s) = 0$, i.e. M_s is injective, then $\theta = \mathfrak{s}^\dagger g$ is the *unique* least square solution. Given $m \in \mathbb{N}$ for each $g \in \text{dom}(M_s)$ we have $g \mathbb{1}^m \in \text{dom}(M_s)$ too. In particular, for $\theta = \mathfrak{s}^\dagger g$ follows $\theta \mathbb{1}^m = (\mathfrak{s}^\dagger g) \mathbb{1}^m = \mathfrak{s}^\dagger (g \mathbb{1}^m) \in \mathbb{J} \mathbb{1}^m$. \square

§07|01 Diagonal statistical inverse problem

§07101.01 **Assumption.** Consider a stochastic process $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ on a probability space $(\Omega, \mathcal{A}, \mathbb{P})$ satisfying Assumption §01101.04 (i.e. $\dot{\epsilon} \in \mathcal{M}(\mathcal{A} \otimes \mathcal{F})$) with *mean zero* (i.e. $\mathbb{P}(\dot{\epsilon}) = (\mathbb{P}(\dot{\epsilon}_j))_{j \in \mathcal{J}} = 0$), a sample size $n \in \mathbb{N}$ and let Assumption §07100.02 be satisfied where $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ is *known* in advance. For $\theta \in \mathbb{J}$ the observable noisy image with mean $g = \mathfrak{s} \theta \in \mathbb{J} = \mathbb{L}_2(\nu)$ takes the form $\widehat{g} = g + n^{-1/2} \dot{\epsilon}$. We denote by $\mathbb{P}_{\theta_s}^n$ the distribution of \widehat{g} . \square

§07101.02 **Definition.** Under Assumption §07101.01 for $\theta \in \mathbb{J}$ and $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ consider a noisy version $\widehat{g} \sim \mathbb{P}_{\theta_s}^n$ of $g = \mathfrak{s} \theta \in \text{dom}(M_s)$. For each $m \in \mathbb{N}$ we call $\widehat{g}^m := \widehat{g} \mathbb{1}^m$ and $\widehat{\theta}^m := \mathfrak{s}^\dagger \widehat{g}^m = \mathfrak{s}^\dagger \widehat{g} \mathbb{1}^m$ *orthogonal projection estimator (OPE)* of g and $\theta = \mathfrak{s}^\dagger g \in \mathbb{J}$, respectively. \square

§07|01|01 Examples

§07101.03 **GdiSM (§01104.09 continued).** Considering $\ell_2 = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ for $\mathbb{J} = \ell_2$ let Assumption §07100.02 be satisfied where $VTU^* = M_s \in \mathbb{L}(\ell_2)$ for some $\mathfrak{s} \in \ell_\infty = \mathbb{L}_\infty(\nu_{\mathbb{N}})$ is *known* in advance. We illustrate the OPE in a Gaussian diagonal inverse sequence model (GdiSM) as in §01104.09. Here the observable stochastic process $\widehat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta_s}^n$ is a noisy version of $g = \mathfrak{s} \theta \in \ell_2$ with $\theta = \mathfrak{s}^\dagger g \in \Theta \subseteq \ell_2$ and $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$. Consequently, \widehat{g} admits a $N_{\theta_s}^n$ -distribution belonging to the family $N_{\Theta \times \{\mathfrak{s}\}}^n := (N_{\theta_s}^n)_{\theta \in \Theta}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, N_{\Theta \times \{\mathfrak{s}\}}^n)$ where $\Theta \subseteq \ell_2$. \square

§07101.04 **Property** (GdiSM §07101.03 continued). *The error process $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ as in Model §07101.03 admits a covariance operator $\text{id}_{\ell_2} \in \mathbb{L}(\ell_2)$ which is evidently invertible with inverse $\text{id}_{\ell_2} \in \mathbb{L}(\ell_2)$ where $\|\text{id}_{\ell_2}\|_{\mathbb{L}(\ell_2)} = 1$ and $N_{(0,1)}^{\otimes \mathbb{N}}(\dot{B}^2) = \mathbf{1}$. For all $h \in \ell_2$ we have $\|h\|_{\ell_2}^2 = \|h\|_{\text{id}_{\ell_2}}^2 = \langle \text{id}_{\ell_2} h, h \rangle_{\ell_2}$. \square*

§07101.05 **Property.** For $\sigma^2 \in \mathbb{R}_{>0} \cap \ell_\infty$ and $P_{(0,\sigma^2)} \in \mathcal{W}_2(\mathcal{B})$, $j \in \mathbb{N}$, a stochastic process $Y \sim \otimes_{j \in \mathbb{N}} P_{(\mu_j, \sigma^2)}$ of independent random variables admits $M_{\sigma^2} \in \mathbb{L}(\ell_2) \cap \mathbb{L}(\ell_2)$ as covariance operator with $\|M_{\sigma^2}\|_{\mathbb{L}(\ell_2)} =$

$\|\sigma^2\|_{\ell_\infty}$, since

$$\langle M_{\sigma^2} a, b \rangle_{\ell_2} = \sum_{j \in \mathbb{N}} \sigma_j^2 a_j b_j = \sum_{j \in \mathbb{N}} a_j \sum_{j_0 \in \mathbb{N}} \text{Cov}(Y_j, Y_{j_0}) b_{j_0} \quad \forall a, b \in \ell_2.$$

If $\sigma^2, \sigma^{-2} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$ then $M_{\sigma^2} \in \mathbb{L}(\ell_2)$ is invertible with inverse $M_{\sigma^2}^{-1} = M_{\sigma^{-2}} \in \mathbb{L}(\ell_2)$ and $\|M_{\sigma^2}^{-1}\|_{\mathbb{L}(\ell_2)} = \|\sigma^{-2}\|_{\ell_\infty}$. \square

§07101.06 **diSM** (§01104.08 continued). For $\mathbb{J} = \ell_2$ let Assumption §07100.02 be satisfied where $\mathfrak{s} \in \ell_\infty = \mathbb{L}_\infty(\nu_{\mathbb{N}})$ is known in advance. We illustrate the OPE in a Diagonal inverse sequence model (diSM) as in §01104.08. Here the observable stochastic process $\widehat{g} = g + n^{-1/2}\dot{\epsilon}$ is a noisy version of $g = \mathfrak{s}\theta \in \ell_2$ with $\theta = \mathfrak{s}^\dagger g \in \Theta \subseteq \ell_2$ and $\dot{\epsilon} \sim \otimes_{j \in \mathbb{N}} \mathbb{P}^{\mathfrak{s}^j}$, where

(iSM1) for $\sigma \in \Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$ and $\mathbb{P}^{\mathfrak{s}^j} = \mathbb{P}_{(0, \sigma^j)} \in \mathcal{W}_2(\mathcal{B})$ for all $j \in \mathbb{N}$,

(iSM2) $\Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$ and for each $\sigma \in \Sigma$ we have $\sigma^{-1} \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$ too.

Under (iSM1) \widehat{g} admits a $\mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n$ -distribution belonging to the family $\mathbb{P}_{\Theta \times \{\mathfrak{s}\} \times \Sigma}^n := (\mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n)_{\theta \in \Theta, \sigma \in \Sigma}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, \mathbb{P}_{\Theta \times \{\mathfrak{s}\} \times \Sigma}^n)$ where $\Theta \subseteq \ell_2$ and $\Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$. \square

§07101.07 **Property** (diSM §07101.06 continued).

(i) Under (iSM1) the error process $\dot{\epsilon} \sim \otimes_{j \in \mathbb{N}} \mathbb{P}_{(0, \sigma^j)}$ admits a covariance operator $M_{\sigma^2} \in \mathbb{L}(\ell_2) \cap \mathbb{L}^{\geq}(\ell_2)$, i.e. $\dot{\epsilon} \sim \mathbb{P}_{(0, M_{\sigma^2})}$, satisfying $\|M_{\sigma^2}\|_{\mathbb{L}(\ell_2)} = \|\sigma^2\|_{\ell_\infty}$ (Property §07101.05) and $\mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n(\dot{\epsilon}^2) = \sigma^2$.

(ii) Under (iSM1) and (iSM2) the covariance operator $M_{\sigma^2} \in \mathbb{L}(\ell_2) \cap \mathbb{L}^{\geq}(\ell_2)$ is invertible with inverse $M_{\sigma^{-2}} \in \mathbb{L}(\ell_2) \cap \mathbb{L}(\ell_2)$ satisfying $\|M_{\sigma^{-2}}\|_{\mathbb{L}(\ell_2)} = \|\sigma^{-2}\|_{\ell_\infty}$.

Under (iSM1) and (iSM2) setting $\mathfrak{v}_\sigma := \max(\|\sigma^{-2}\|_{\ell_\infty}, \|\sigma^2\|_{\ell_\infty})$ we evidently have $\|M_{\sigma^2}\|_{\mathbb{L}(\ell_2)} \leq \mathfrak{v}_\sigma$ and $\|M_{\sigma^{-2}}\|_{\mathbb{L}(\ell_2)} \leq \mathfrak{v}_\sigma$. Consequently, from Lemma §01101.08 (01.03) we obtain

$$\mathfrak{v}_\sigma^{-1} \|h_\bullet\|_{\ell_2}^2 \leq \|h_\bullet\|_{M_{\sigma^2}}^2 = \langle M_{\sigma^2} h_\bullet, h_\bullet \rangle_{\ell_2} \leq \mathfrak{v}_\sigma \|h_\bullet\|_{\ell_2}^2 \quad \forall h_\bullet \in \ell_2. \quad \square$$

§07101.08 **dieMM** (§01104.07 continued). For $\mathbb{J} = \mathbb{L}_2(\nu)$ let Assumption §07100.02 be satisfied where $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ is known in advance. We illustrate the OPE in a Diagonal inverse empirical mean model (dieMM) as in §01104.07. Here the observable stochastic process $\widehat{g} = g + n^{-1/2}\dot{\epsilon}$ is a noisy version of $g = \mathfrak{s}\theta \in \mathbb{J}$ with $\theta = \mathfrak{s}^\dagger g \in \Theta \subseteq \mathbb{J}$, and error process $\dot{\epsilon} = n^{1/2}(\widehat{\mathbb{P}}_n(\psi) - \mathbb{P}_{\theta|\mathfrak{s}}(\psi)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J})$ satisfying Assumption §01101.04. More precisely, on a measurable space $(\mathcal{Z}, \mathcal{Z})$ for each $\theta \in \Theta \subseteq \mathbb{J}$ there is a probability measure $\mathbb{P}_{\theta|\mathfrak{s}} \in \mathcal{W}(\mathcal{Z})$. Consider a stochastic process $\psi = (\psi_j)_{j \in \mathcal{J}} \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ which in addition for $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ and for each $\theta \in \Theta \subseteq \mathbb{J}$ satisfies

(dieMM1) $\psi_j \in \mathcal{L}_1(\mathbb{P}_{\theta|\mathfrak{s}}) := \mathcal{L}_1(\mathcal{Z}, \mathcal{Z}, \mathbb{P}_{\theta|\mathfrak{s}})$ ν -a.e. $j \in \mathcal{J}$ and $\mathbb{P}_{\theta|\mathfrak{s}}(\psi) = \mathfrak{s}\theta = g$ ν -a.s.,

(dieMM2) $(\psi - \mathbb{P}_{\theta|\mathfrak{s}}(\psi))\mathbb{1}^m \in \mathbb{L}_\infty(\nu)$ $\mathbb{P}_{\theta|\mathfrak{s}}$ -a.s. for each $m \in \mathbb{N}$,

(dieMM3) there is $\mathfrak{v}_{\theta|\mathfrak{s}|\psi} \in \mathbb{R}_{\geq 1}$ such that $\|\mathbb{P}_{\theta|\mathfrak{s}}(\psi^2)\|_{\mathbb{L}_\infty(\nu)} \leq \mathfrak{v}_{\theta|\mathfrak{s}|\psi}$ and

$$\mathbb{P}_{\theta|\mathfrak{s}}(|\nu(h \cdot \psi)|^2) \leq \mathfrak{v}_{\theta|\mathfrak{s}|\psi} \|h_\bullet\|_{\mathbb{J}}^2, \quad \forall h_\bullet \in \mathbb{J},$$

(dieMM4) $\mathfrak{v}_{\theta|\mathfrak{s}} := \mathbb{P}_{\theta|\mathfrak{s}}(\psi^2) - |\mathbb{P}_{\theta|\mathfrak{s}}(\psi)|^2 \in \mathcal{M}_{>0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$, $\|(\mathfrak{v}_{\theta|\mathfrak{s}})^{-1}\|_{\mathbb{L}_\infty(\nu)} \leq \mathfrak{v}_{\theta|\mathfrak{s}|\psi}$ and

$$\mathbb{P}_{\theta|\mathfrak{s}}(|\nu(h \cdot \psi)|^2) \geq \mathbb{P}_{\theta|\mathfrak{s}}(|\nu(h \cdot \psi)|^2) - |\mathbb{P}_{\theta|\mathfrak{s}}(\nu(h \cdot \psi))|^2 \geq \mathfrak{v}_{\theta|\mathfrak{s}|\psi}^{-1} \|h_\bullet\|_{\mathbb{J}}^2, \quad \forall h_\bullet \in \mathbb{J}.$$

We consider a statistical product experiment $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, \mathbb{P}_{\Theta \times \{\mathfrak{s}\}}^{\otimes n} = (\mathbb{P}_{\theta|\mathfrak{s}}^{\otimes n})_{\theta \in \Theta})$ as in an Empirical mean function §01101.10 where $\Theta \subseteq \mathbb{J}$. \square

§0701.09 **Property** (dieMM §0701.08 continued).

- (i) Under (dieMM1)–(dieMM3) due to **Lemma** §0101.08 (i) the stochastic process $\psi \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ and hence the error process $\dot{\epsilon} = n^{1/2}(\hat{\mathbb{P}}_n - \mathbb{P}_{\theta|\mathfrak{s}})(\psi) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J})$ admits a covariance operator $\Gamma_{\theta|\mathfrak{s}} \in \mathbb{L}(\mathbb{J})$ satisfying $\|\Gamma_{\theta|\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} \leq \mathfrak{v}_{\theta|\mathfrak{s}|\psi}$.
- (ii) Under (dieMM1)–(dieMM4) due to **Lemma** §0101.08 (ii) the covariance operator $\Gamma_{\theta|\mathfrak{s}} \in \mathbb{L}(\mathbb{J})$ is invertible with inverse $\Gamma_{\theta|\mathfrak{s}}^{-1} \in \mathbb{L}(\mathbb{J})$ satisfying $\|\Gamma_{\theta|\mathfrak{s}}^{-1}\|_{\mathbb{L}(\mathbb{J})} \leq \mathfrak{v}_{\theta|\mathfrak{s}|\psi}$.

Consequently, from **Lemma** §0101.08 (01.03) we obtain

$$\mathfrak{v}_{\theta|\mathfrak{s}|\psi}^{-1} \|h_\bullet\|_{\mathbb{J}}^2 \leq \|h_\bullet\|_{\Gamma_{\theta|\mathfrak{s}}}^2 = \langle \Gamma_{\theta|\mathfrak{s}} h_\bullet, h_\bullet \rangle_{\mathbb{J}} \leq \mathfrak{v}_{\theta|\mathfrak{s}|\psi} \|h_\bullet\|_{\mathbb{J}}^2 \quad \forall h_\bullet \in \mathbb{J}. \quad \square$$

§07|01|02 Global and maximal global \mathfrak{v} -risk

We measure first the accuracy of the OPE $\hat{\theta}^m := \mathfrak{s}^\dagger \hat{g}^m$ of the projection $\theta^m = \mathfrak{s}^\dagger g^m \in \mathbb{J} \mathbb{1}^m$ with $g = \mathfrak{s} \theta \in \text{dom}(M_\mathfrak{s})$ by the mean of its global \mathfrak{v} -error introduced in §04|03|01, i.e. its \mathfrak{v} -risk.

§0701.10 **Reminder.** If $\mathfrak{v} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\theta \in \mathbb{L}_2(\mathfrak{v}^2 \nu)$ then we have $\theta^m \in \mathbb{L}_2(\mathfrak{v}^2 \nu)$ too and $\|\theta^m - \theta\|_{\mathfrak{v}}^2 = o(1)$ as $m \rightarrow \infty$ (**Property** §04|03.09). \square

§0701.11 **Assumption.** Consider a noisy version $\hat{g} = g + n^{-1/2} \dot{\epsilon} \sim \mathbb{P}_{\theta|\mathfrak{s}}^n$ satisfying Assumption §0701.01, (dSIPg1) $\mathfrak{v}_j^{\theta|\mathfrak{s}} := \mathbb{P}_{\theta|\mathfrak{s}}^n(\dot{\epsilon}_j^2) := (\mathfrak{v}_j^{\theta|\mathfrak{s}} := \mathbb{P}_{\theta|\mathfrak{s}}^n(|\dot{\epsilon}_j|^2))_{j \in \mathcal{J}} \in \mathbb{L}_\infty(\nu)$ and (dSIPg2) $\dot{\epsilon} \mathbb{1}^m \in \mathbb{L}_\infty(\nu) \mathbb{P}_{\theta|\mathfrak{s}}^n$ -a.s., for each $m \in \mathbb{N}$. \square

§0701.12 **Comment.** Under Assumption §0701.11 and $\mathfrak{v} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ set $(\mathfrak{s}^\dagger \mathfrak{v})_\bullet := \mathfrak{s}^\dagger \mathfrak{v} \in \mathcal{M}(\mathcal{J})$. If $\mathfrak{s}^\dagger \mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}^2 \nu)$ then we have $(\mathfrak{s}^\dagger \mathfrak{v})_\bullet \dot{\epsilon} \mathbb{1}^m \in \mathbb{J} \mathbb{P}_{\theta|\mathfrak{s}}^n$ -a.s.. If in addition $\theta \in \mathbb{L}_2(\mathfrak{v}^2 \nu)$, and hence $\theta^m \in \mathbb{L}_2(\mathfrak{v}^2 \nu)$ (**Property** §04|03.09), then it follows

$$\mathfrak{v} \hat{\theta}^m = (\mathfrak{s}^\dagger \mathfrak{v})_\bullet \hat{g} \mathbb{1}^m = n^{-1/2} (\mathfrak{s}^\dagger \mathfrak{v})_\bullet \dot{\epsilon} \mathbb{1}^m + \mathfrak{v} \theta^m \in \mathbb{J} = \mathbb{L}_2(\nu) \quad \mathbb{P}_{\theta|\mathfrak{s}}^n\text{-a.s.} \quad (07.01)$$

If $\mathcal{J} \subseteq \mathbb{Z}$ (at most countable) and ν_j is the counting measure over the index set \mathcal{J} then Assumption §0101.04 and (dSIPg1) $\mathfrak{v}_j^{\theta|\mathfrak{s}} = \mathbb{P}_{\theta|\mathfrak{s}}^n(\dot{\epsilon}_j^2) \in \mathbb{L}_\infty(\nu_j)$ implies the additional assumption (dSIPg2) $\dot{\epsilon} \mathbb{1}^m \in \mathbb{L}_\infty(\nu_j) \mathbb{P}_{\theta|\mathfrak{s}}^n$ -a.s.. However, the last implication does generally not hold, if $\mathcal{J} \in \{\mathbb{R}, \mathbb{R}_{\geq 0}\}$ for example. \square

§07|01|02|01 Global \mathfrak{v} -risk

§0701.13 **Assumption.** Let $\mathfrak{v} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$, $\theta \in \mathbb{L}_2(\mathfrak{v}^2 \nu)$, and $\mathfrak{s}^\dagger \mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}^2 \nu)$ for all $m \in \mathbb{N}$ be satisfied. \square

§0701.14 **Definition.** Under Assumptions §0701.11 and §0701.13 the *global \mathfrak{v} -risk* of an OPE $\hat{\theta}^m = \mathfrak{s}^\dagger \hat{g}^m = \mathfrak{s}^\dagger \hat{g} \mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}^2 \nu) \mathbb{P}_{\theta|\mathfrak{s}}^n$ -a.s. satisfies

$$\mathbb{E}_{\theta|\mathfrak{s}}^n(\|\hat{\theta}^m - \theta\|_{\mathfrak{v}}^2) = \mathbb{E}_{\theta|\mathfrak{s}}^n(\|\mathfrak{s}^\dagger(\hat{g} - g)\mathbb{1}^m\|_{\mathfrak{v}}^2) + \|\theta \mathbb{1}^{m \perp}\|_{\mathfrak{v}}^2 \quad (07.02)$$

with *variance* term $\mathbb{E}_{\theta|\mathfrak{s}}^n(\|\mathfrak{s}^\dagger(\hat{g} - g)\mathbb{1}^m\|_{\mathfrak{v}}^2) = n^{-1} \mathbb{E}_{\theta|\mathfrak{s}}^n(\|(\mathfrak{s}^\dagger \mathfrak{v})_\bullet \dot{\epsilon} \mathbb{1}^m\|_{\mathfrak{v}}^2)$ and *bias* term $\|\theta \mathbb{1}^{m \perp}\|_{\mathfrak{v}}$. \square

§0701.15 **Property.** Under Assumptions §0701.11 and §0701.13 we have

$$\mathbb{E}_{\theta|\mathfrak{s}}^n(\|(\mathfrak{s}^\dagger \mathfrak{v})_\bullet \dot{\epsilon} \mathbb{1}^m\|_{\mathfrak{v}}^2) = \int_{\mathcal{J}} \mathbb{E}_{\theta|\mathfrak{s}}^n(|\dot{\epsilon}_j|^2) (\mathfrak{s}^\dagger \mathfrak{v})_j^2 \mathbb{1}_j^m \nu(dj) = \nu(\mathfrak{v}_j^{\theta|\mathfrak{s}} (\mathfrak{s}^\dagger \mathfrak{v})_j^2 \mathbb{1}_j^m) \quad (07.03)$$

and consequently $\mathbb{E}_{\theta|\mathfrak{s}}^n(\|\mathfrak{s}^\dagger(\hat{g} - g)\mathbb{1}^m\|_{\mathfrak{v}}^2) \leq n^{-1} \|\mathfrak{v}_j^{\theta|\mathfrak{s}}\|_{\mathbb{L}_\infty(\nu)} \|\mathfrak{s}^\dagger \mathbb{1}^m\|_{\mathfrak{v}}^2 \in \mathbb{R}_{\geq 0}$. \square

§07101.16 **Notation.** For $a_\cdot \in \mathbb{R}^{\mathbb{N}}$ with minimal value in $B \subseteq \mathbb{N}$ we define

$$\arg \min \{a_m : m \in B\} := \min \{m \in B : a_m \leq a_j, \forall j \in B\}. \quad \square$$

§07101.17 **Proposition (Upper bound).** Under Assumptions §07101.11 and §07101.13 for all $n, m \in \mathbb{N}$ setting

$$\begin{aligned} R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) &:= \|\theta \cdot \mathbf{1}^{m \perp}\|_{\mathfrak{v}}^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\mathfrak{v}}^2, \quad m_n^\circ := \arg \min \{R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) : m \in \mathbb{N}\} \\ \text{and } R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) &:= R_n^{m_n^\circ}(\theta, \mathfrak{s}, \mathfrak{v}) = \min \{R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) : m \in \mathbb{N}\} \end{aligned} \quad (07.04)$$

we have $\mathbb{E}_{\theta|\mathfrak{s}}^n(\|\widehat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2) \leq (1 \vee \|\mathfrak{v}^{\theta|\mathfrak{s}}\|_{\mathbb{L}_\infty(\nu)}) R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$ for all $n \in \mathbb{N}$.

§07101.18 **Proof of Proposition §07101.17.** Given in the lecture. □

§07101.19 **Definition.** Let $\theta \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ and $\widehat{\theta}^m \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ $\mathbb{P}_{\theta|\mathfrak{s}}^n$ -a.s. for all $m \in \mathbb{N}$. If there exist $C \in \mathbb{R}_{>0}$ and for each $n \in \mathbb{N}$, $R_n^\circ \in \mathbb{R}_{>0}$ and $m_n^\circ \in \mathbb{N}$ satisfying

$$C^{-1} R_n^\circ \leq \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta|\mathfrak{s}}^n \|\widehat{\theta}^m - \theta\|_{\mathfrak{v}}^2 \leq \mathbb{E}_{\theta|\mathfrak{s}}^n \|\widehat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2 \leq C R_n^\circ \quad \forall n \in \mathbb{N},$$

then we call R_n° *oracle bound*, m_n° *oracle dimension* and $\widehat{\theta}^{m_n^\circ}$ *oracle optimal* (up to the constant C). As a consequence, up to the constant C^2 the statistic $\widehat{\theta}^{m_n^\circ}$ attains the lower global \mathfrak{v} -risk bound within the family of OPE's, that is, $\mathbb{E}_{\theta|\mathfrak{s}}^n \|\widehat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2 \leq C^2 \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta|\mathfrak{s}}^n \|\widehat{\theta}^m - \theta\|_{\mathfrak{v}}^2$. □

§07101.20 **Oracle inequality.** Under Assumptions §07101.11 and §07101.13 if in addition

$$1 \leq \max(\|\mathfrak{v}^{\theta|\mathfrak{s}}\|_{\mathbb{L}_\infty(\nu)}, \|(\mathfrak{v}^{\theta|\mathfrak{s}})^{-1}\|_{\mathbb{L}_\infty(\nu)}) \leq \mathfrak{v}_{\theta|\mathfrak{s}} \in \mathbb{R}_{\geq 1}$$

is satisfied then (07.04) implies

$$\begin{aligned} \mathfrak{v}_{\theta|\mathfrak{s}}^{-1} R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) &\leq \mathbb{E}_{\theta|\mathfrak{s}}^n(\|\widehat{\theta}^m - \theta\|_{\mathfrak{v}}^2) = n^{-1} \nu (\mathfrak{v}^{\theta|\mathfrak{s}}(\mathfrak{s}^\dagger \mathfrak{v})^2 \mathbf{1}^m) + \|\theta \cdot \mathbf{1}^{m \perp}\|_{\mathfrak{v}}^2 \\ &\leq \mathfrak{v}_{\theta|\mathfrak{s}} R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) \quad \forall m, n \in \mathbb{N}. \end{aligned}$$

As a consequence we immediately obtain the following *oracle inequality*

$$\begin{aligned} \mathfrak{v}_{\theta|\mathfrak{s}}^{-1} R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) &\leq \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta|\mathfrak{s}}^n(\|\widehat{\theta}^m - \theta\|_{\mathfrak{v}}^2) \leq \mathbb{E}_{\theta|\mathfrak{s}}^n(\|\widehat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2) \\ &\leq \mathfrak{v}_{\theta|\mathfrak{s}} R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) \leq \mathfrak{v}_{\theta|\mathfrak{s}}^2 \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta|\mathfrak{s}}^n(\|\widehat{\theta}^m - \theta\|_{\mathfrak{v}}^2) \quad \forall n \in \mathbb{N}, \end{aligned} \quad (07.05)$$

and, hence $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$, m_n° and the statistic $\widehat{\theta}^{m_n^\circ}$, respectively, is an *oracle bound*, an *oracle dimension* and *oracle optimal* (up to the constant $\mathfrak{v}_{\theta|\mathfrak{s}}^2$). □

§07101.21 **Remark.** For each fixed $m \in \mathbb{N}$ with $\|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\mathfrak{v}} \in \mathbb{R}_{\geq 0}$ we have $n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\mathfrak{v}} = o(1)$ as $n \rightarrow \infty$. As a consequence, if $\|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\mathfrak{v}} \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and $\|\theta \cdot \mathbf{1}^{m \perp}\|_{\mathfrak{v}} = o(1)$ as $m \rightarrow \infty$ then we obtain $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) = o(1)$ as $n \rightarrow \infty$, and thus, $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$ is also called an *oracle rate*. Indeed, for all $\delta \in \mathbb{R}_{>0}$ there exists $m_\delta \in \mathbb{N}$ and $n_\delta \in \mathbb{N}$ such that we have both $\|\theta \cdot \mathbf{1}^{m_\delta \perp}\|_{\mathfrak{v}}^2 \leq \delta/2$ and $n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^{m_\delta}\|_{\mathfrak{v}}^2 \leq \delta/2$ for all $n \in \mathbb{N}_{\geq n_\delta}$, and whence $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) \leq R_n^{m_\delta}(\theta, \mathfrak{s}, \mathfrak{v}) \leq \delta$. However, note that the oracle dimension $m_n^\circ = m_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$ as defined in **Proposition §07101.17** depends on the unknown parameter of interest θ , and thus also the oracle optimal statistic $\widehat{\theta}^{m_n^\circ}$. In other words $\widehat{\theta}^{m_n^\circ}$ is not a feasible estimator. □

§07101.22 **Corollary** (GdiSM §07101.03 continued). Consider $\widehat{g}_\bullet = g_\bullet + n^{-1/2}\dot{B}_\bullet \sim N_{\theta|\mathfrak{s}}^n$ as in Model §07101.03, where $\dot{B}_\bullet \sim N_{(0,1)}^{\otimes N}$, $\mathfrak{s}_\bullet \in \ell_\infty$, $\theta \in \ell_2$ and hence $g_\bullet = \mathfrak{s}_\bullet\theta \in \text{dom}(M_\bullet) \subseteq \ell_2$. Given $\mathfrak{v}_\bullet \in \mathbb{R}_{>0}^N$ and $\theta \in \ell_2(\mathfrak{v}_\bullet^2)$ the (infeasible) OPE $\widehat{\theta}_\bullet^{m_n^\circ} = \mathfrak{s}_\bullet^\dagger \widehat{g}_\bullet^{m_n^\circ} \in \ell_2(\mathfrak{v}_\bullet^2)$ with oracle dimension m_n° as in (07.04) satisfies

$$N_{\theta|\mathfrak{s}}^n (\|\widehat{\theta}_\bullet^{m_n^\circ} - \theta\|_{\mathfrak{v}_\bullet}^2) = R_n^\circ(\theta, \mathfrak{s}_\bullet, \mathfrak{v}_\bullet) = \inf_{m \in \mathbb{N}} N_{\theta|\mathfrak{s}}^n (\|\widehat{\theta}_\bullet^m - \theta\|_{\mathfrak{v}_\bullet}^2) \quad \forall n \in \mathbb{N},$$

and hence it is *oracle optimal* (with constant 1).

§07101.23 **Proof of Corollary** §07101.22. Given in the lecture. □

§07101.24 **Corollary** (diSM §07101.06 continued). Consider $\widehat{g}_\bullet = g_\bullet + n^{-1/2}\dot{\epsilon}_\bullet \sim P_{\theta|\mathfrak{s}\sigma}^n$ as in Model §07101.06, where $\dot{\epsilon}_\bullet \sim \otimes_{j \in \mathbb{N}} P_{(0,\sigma_j^2)}$ satisfies (iSM1) and (iSM2) with $\max(\|\sigma_\bullet^{-2}\|_{\ell_\infty}, \|\sigma_\bullet^2\|_{\ell_\infty}) =: \mathfrak{v}_\sigma \in \mathbb{R}_{>1}$, $\mathfrak{s}_\bullet \in \ell_\infty$, $\theta \in \ell_2$ and hence $g_\bullet = \mathfrak{s}_\bullet\theta \in \text{dom}(M_\bullet) \subseteq \ell_2$. Given $\mathfrak{v}_\bullet \in \mathbb{R}_{>0}^N$ and $\theta \in \ell_2(\mathfrak{v}_\bullet^2)$ the (infeasible) OPE $\widehat{\theta}_\bullet^{m_n^\circ} = \mathfrak{s}_\bullet^\dagger \widehat{g}_\bullet^{m_n^\circ} \in \ell_2(\mathfrak{v}_\bullet^2)$ with oracle dimension m_n° as in (07.04) satisfies

$$P_{\theta|\mathfrak{s}\sigma}^n (\|\widehat{\theta}_\bullet^{m_n^\circ} - \theta\|_{\mathfrak{v}_\bullet}^2) \leq \mathfrak{v}_\sigma R_n^\circ(\theta, \mathfrak{s}_\bullet, \mathfrak{v}_\bullet) \leq \mathfrak{v}_\sigma^2 \inf_{m \in \mathbb{N}} P_{\theta|\mathfrak{s}\sigma}^n (\|\widehat{\theta}_\bullet^m - \theta\|_{\mathfrak{v}_\bullet}^2) \quad \forall n \in \mathbb{N},$$

and hence it is *oracle optimal* (with constant \mathfrak{v}_σ).

§07101.25 **Proof of Corollary** §07101.24. Given in the lecture. □

§07101.26 **Corollary** (dieMM §07101.08 continued). Let $\widehat{g}_\bullet = g_\bullet + n^{-1/2}\dot{\epsilon}_\bullet$ be defined on $(\mathcal{Z}^n, \mathcal{L}^{\otimes n}, P_{\theta|\mathfrak{s}}^{\otimes n})$ as in Model §07101.08, where $\psi_\bullet \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ satisfies (dieMM1)–(dieMM4) for some $\mathfrak{v}_{\theta|\mathfrak{s}\psi} \in \mathbb{R}_{>1}$, $\mathfrak{s}_\bullet \in \mathbb{L}_\infty(\nu)$, $\theta \in \mathbb{J}$ and hence $g_\bullet = \mathfrak{s}_\bullet\theta \in \text{dom}(M_\bullet) \subseteq \mathbb{J}$. Under Assumption §07101.13 the (infeasible) OPE $\widehat{\theta}_\bullet^{m_n^\circ} = \mathfrak{s}_\bullet^\dagger \widehat{g}_\bullet^{m_n^\circ} \in \mathbb{L}_2(\mathfrak{v}_\bullet^2\nu)$ $P_{\theta|\mathfrak{s}}^{\otimes n}$ -a.s. with oracle dimension m_n° as in (07.04) satisfies

$$P_{\theta|\mathfrak{s}}^{\otimes n} (\|\widehat{\theta}_\bullet^{m_n^\circ} - \theta\|_{\mathfrak{v}_\bullet}^2) \leq \mathfrak{v}_{\theta|\mathfrak{s}\psi} R_n^\circ(\theta, \mathfrak{s}_\bullet, \mathfrak{v}_\bullet) \leq \mathfrak{v}_{\theta|\mathfrak{s}\psi}^2 \inf_{m \in \mathbb{N}} P_{\theta|\mathfrak{s}}^{\otimes n} (\|\widehat{\theta}_\bullet^m - \theta\|_{\mathfrak{v}_\bullet}^2) \quad \forall n \in \mathbb{N},$$

and hence it is *oracle optimal* (with constant $\mathfrak{v}_{\theta|\mathfrak{s}\psi}$).

§07101.27 **Proof of Corollary** §07101.26. Given in the lecture. □

§07101.28 **Illustration.** We illustrate the last results considering usual behaviour for $\theta, \mathfrak{s}_\bullet, \mathfrak{v}_\bullet \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$. We distinguish the following two cases

- (p) $\mathfrak{s}_\bullet^\dagger \in \mathbb{L}_2(\mathfrak{v}_\bullet^2\nu)$ or there is $m \in \mathbb{N}$ with $\|\theta_\bullet^m - \theta\|_{\mathfrak{v}_\bullet}^2 = 0$,
- (np) $\mathfrak{s}_\bullet^\dagger \notin \mathbb{L}_2(\mathfrak{v}_\bullet^2\nu)$ and for all $m \in \mathbb{N}$ holds $\|\theta_\bullet^m - \theta\|_{\mathfrak{v}_\bullet}^2 \in \mathbb{R}_{>0}$.

Interestingly, in case (p) the oracle bound is parametric, that is, $nR_n^\circ(\theta, \mathfrak{s}_\bullet, \mathfrak{v}_\bullet) = O(1)$, in case (np) the oracle bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^\circ(\theta, \mathfrak{s}_\bullet, \mathfrak{v}_\bullet) = \infty$. In case (np) consider the following three specifications:

Table 01 [§07]

Order of the oracle rate $R_n^\circ(\theta, \mathfrak{s}_\bullet, \mathfrak{v}_\bullet)$ as $n \rightarrow \infty$

$(j \in \mathcal{J})$ $\mathfrak{v}_j^2 = j^{2\nu}$	$(a \in \mathbb{R}_{>0})$ $(t \in \mathbb{R}_{>0})$	(squared bias)	(variance)	m_n°	$R_n^\circ(\theta, \mathfrak{s}_\bullet, \mathfrak{v}_\bullet)$
	θ_j^2 \mathfrak{s}_j^2	$\ \theta_\bullet \mathbf{1}_\bullet^{m \perp}\ _{\mathfrak{v}_\bullet}^2$	$\ \mathfrak{s}_\bullet^\dagger \mathbf{1}_\bullet^m\ _{\mathfrak{v}_\bullet}^2$		
(o-m) $\nu \in (-1/2 - t, a)$	j^{-2a-1} j^{-2t}	$m^{-2(a-\nu)}$	$m^{2\nu+2t+1}$	$n^{\frac{1}{2a+2t+1}}$	$n^{-\frac{2(a-\nu)}{2a+2t+1}}$
$\nu + t = -1/2$	j^{-2a-1} j^{-2t}	$m^{-2a-2t-1}$	$\log m$	$\left(\frac{n}{\log n}\right)^{\frac{1}{2a+2t+1}}$	$\frac{\log n}{n}$
(o-s) $a - \nu \in \mathbb{R}_{>0}$	j^{-2a-1} $e^{-j^{2t}}$	$m^{-2(a-\nu)}$	$m^{(1-2(t-\nu))+e^{m^{2t}}}$	$(\log n)^{\frac{1}{2t}}$	$(\log n)^{-\frac{a-\nu}{t}}$
(s-m) $\nu + t + 1/2 \in \mathbb{R}_{>0}$	$e^{-j^{2a}}$ j^{-2t}	$m^{(1-2(a-\nu))+e^{-m^{2a}}}$	$m^{2\nu+2t+1}$	$(\log n)^{\frac{1}{2a}}$	$\frac{(\log n)^{\frac{2t+2\nu+1}{2a}}}{n}$
$\nu + t = -1/2$	$e^{-j^{2a}}$ j^{-2t}	$e^{-m^{2a}}$	$\log m$	$(\log n)^{\frac{1}{2a}}$	$\frac{\log \log n}{n}$

We note that in case **(o-m)** and **(s-m)** for $v + t < -1/2$ the oracle rate $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$ is parametric. \square

§07|01|02|02 Maximal global \mathfrak{v} -risk

§07|01.29 **Notation (Reminder)**. For sequences $a_n, b_n \in (\mathbb{K})^{\mathbb{N}}$ taking its values in $\mathbb{K} \in \{\mathbb{R}, \mathbb{R}_{>0}, \mathbb{Q}, \mathbb{Z}, \dots\}$ we write $a_n \in (\mathbb{K})_{\nearrow}^{\mathbb{N}}$ and $b_n \in (\mathbb{K})_{\searrow}^{\mathbb{N}}$ if a_n and b_n , respectively, is monotonically *non-decreasing* and *non-increasing*. If in addition $a_n \rightarrow \infty$ and $b_n \rightarrow 0$ as $n \rightarrow \infty$, then we write $a_n \in (\mathbb{K})_{\nearrow}^{\mathbb{N}}$ and $b_n \in (\mathbb{K})_{\searrow}^{\mathbb{N}}$ for short. For $w_n \in \mathbb{L}_\infty(\nu)$ we set $w_{(0)} := \|w_n\|_{\mathbb{L}_\infty(\nu)}$ and $w_{(\bullet)} = (w_{(j)} := \|w_n \mathbb{1}_j^{\perp}\|_{\mathbb{L}_\infty(\nu)})_{j \in \mathbb{N}}$, where by construction $w_{(\bullet)} \in (\mathbb{R}_{\geq 0})_{\searrow}^{\mathbb{N}}$. \square

§07|01.30 **Assumption**. Consider weights $\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n \in \mathcal{M}_{>0, \nu}(\mathcal{J})$ (i.e. $\nu(\mathcal{N}_{\mathfrak{a}}) = \nu(\mathcal{N}_{\mathfrak{t}}) = 0 = \nu(\mathcal{N}_{\mathfrak{v}})$), such that $\mathfrak{a}_n, \mathfrak{t}_n \in \mathbb{L}_\infty(\nu)$, $(\mathfrak{a}\mathfrak{v})_n = \mathfrak{a}_n \mathfrak{v}_n \in \mathbb{L}_\infty(\nu)$, $(\mathfrak{a}\mathfrak{v})_{(j)} \in (\mathbb{R}_{\geq 0})_{\searrow}^{\mathbb{N}}$, and $\mathfrak{t}_n^\dagger \mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ for all $m \in \mathbb{N}$. \square

§07|01.31 **Reminder**. Under Assumption §07|01.30 we have $\mathbb{J}^{\mathfrak{a}} = \mathbb{L}_2(\nu) = \text{dom}(M_{\mathfrak{a}}) = \mathfrak{J}\mathfrak{a} \subseteq \mathbb{J}$ and the three measures ν , $\mathfrak{a}^{2\dagger} \nu$ and $\mathfrak{v}^2 \nu$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04|01.02). We consider $\mathbb{J}^{\mathfrak{a}}$ endowed with $\|\cdot\|_{\mathfrak{a}^\dagger} = \|M_{\mathfrak{a}} \cdot\|_{\mathbb{J}}$ and given a constant $r \in \mathbb{R}_{>0}$ the ellipsoid $\mathbb{J}^{\mathfrak{a}, r} := \{h_n \in \mathbb{J}^{\mathfrak{a}} : \|h_n\|_{\mathfrak{a}^\dagger} \leq r\} \subseteq \mathbb{J}^{\mathfrak{a}}$. Since $(\mathfrak{a}\mathfrak{v})_n \in \mathbb{L}_\infty(\nu)$, and hence $(\mathfrak{a}\mathfrak{v})_{(m)} := \|(\mathfrak{a}\mathfrak{v})_n \mathbb{1}_m^{\perp}\|_{\mathbb{L}_\infty(\nu)} \in \mathbb{R}_{\geq 0}$ for each $m \in \mathbb{N}$ we have $\mathbb{J}^{\mathfrak{a}} \subseteq \mathbb{L}_2(\mathfrak{v}^2 \nu)$ (**Property** §04|02.11), and $\|\theta_n \mathbb{1}_m^{\perp}\|_{\mathfrak{v}} \leq r (\mathfrak{a}\mathfrak{v})_{(m)}$ for all $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$ (**Lemma** §04|02.13). Consequently, if Assumption §07|01.30, $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$ and $\mathfrak{s}_n^\dagger \mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ for all $m \in \mathbb{N}$ are satisfied, then Assumption §07|01.13 is fulfilled. Moreover, under Assumption §07|01.30 for each $M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, d}$ we have $\|\mathfrak{s}_n^\dagger \mathbb{1}^m\|_{\mathfrak{v}} \leq d \|\mathfrak{t}_n^\dagger \mathbb{1}^m\|_{\mathfrak{v}} \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ (**Definition** §04|03.05). Therefore, if Assumption §07|01.30, $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$ and $M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, d}$ are satisfied, then Assumption §07|01.13 is also fulfilled. \square

§07|01.32 **Proposition (Upper bound)**. Under Assumptions §07|01.11 and §07|01.30 let $\mathfrak{s}_n^\dagger \mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ for all $m \in \mathbb{N}$. Setting for $n, m \in \mathbb{N}$

$$R_n^m(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) := [(\mathfrak{a}\mathfrak{v})_{(m)}^2 \vee n^{-1} \|\mathfrak{s}_n^\dagger \mathbb{1}^m\|_{\mathfrak{v}}^2], \quad m_n^* := \arg \min \{R_n^m(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) : m \in \mathbb{N}\}$$

$$\text{and } R_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) := R_n^{m_n^*}(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) = \min \{R_n^m(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) : m \in \mathbb{N}\} \quad (07.06)$$

and $\|\mathfrak{v}_{\theta_n^{\mathfrak{s}}}\|_{\mathbb{L}_\infty(\nu)} =: \mathfrak{v}_{\theta_n^{\mathfrak{s}}} \in \mathbb{R}_{\geq 0}$, for all $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$, hence $g_n = \mathfrak{s}_n \theta_n \in \text{dom}(M_{\mathfrak{a}}) \subseteq \mathbb{J}$, we have

$$\mathbb{E}_{\theta_n^{\mathfrak{s}}}^n (\|\widehat{\theta}_n^{m_n^*} - \theta_n\|_{\mathfrak{v}}^2) \leq (\mathfrak{v}_{\theta_n^{\mathfrak{s}}} + r^2) R_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) \quad \forall n \in \mathbb{N}.$$

§07|01.33 **Proof of Proposition** §07|01.32. Given in the lecture. \square

§07|01.34 **Remark**. Under the assumptions of **Proposition** §07|01.32 if there exists in addition $\mathfrak{v}_n \in \mathbb{R}_{\geq 0}$ satisfying $\|\mathfrak{v}_{\theta_n^{\mathfrak{s}}}\|_{\mathbb{L}_\infty(\nu)} \leq \mathfrak{v}_n$ for all $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$ then

$$\sup \{ \mathbb{E}_{\theta_n^{\mathfrak{s}}}^n (\|\widehat{\theta}_n^{m_n^*} - \theta_n\|_{\mathfrak{v}}^2) : \theta_n \in \mathbb{J}^{\mathfrak{a}, r} \} \leq (\mathfrak{v}_n + r^2) R_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) \quad \forall n \in \mathbb{N}.$$

Arguing similarly as in **Remark** §07|01.21 we note that $R_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) = o(1)$ as $n \rightarrow \infty$, whenever $\|\mathfrak{s}_n^\dagger \mathbb{1}^m\|_{\mathfrak{v}} \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ (note that $(\mathfrak{a}\mathfrak{v})_{(m)} = o(1)$ as $m \rightarrow \infty$ by Assumption §07|01.30 which is satisfied, for example, if $(\mathfrak{a}\mathfrak{v})_n = \mathfrak{a}_n \mathfrak{v}_n \in \mathbb{J}$ or in equal $\mathfrak{a}_n \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$). Note that the dimension $m_n^* := m_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v})$ as defined in (07.06) does not depend on the unknown parameter of interest θ_n but on the class $\mathbb{J}^{\mathfrak{a}, r}$ only, and thus also the statistic $\widehat{\theta}_n^{m_n^*}$. In other words, if the regularity of θ_n is known in advance, then the OPE $\widehat{\theta}_n^{m_n^*}$ is a feasible estimator. \square

§07|01.35 **Corollary (Upper bound)**. Under Assumptions §07|01.11 and §07|01.30 setting for $n, m \in \mathbb{N}$

$$R_n^m(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) := [(\mathfrak{a}\mathfrak{v})_{(m)}^2 \vee n^{-1} \|\mathfrak{t}_n^\dagger \mathbb{1}^m\|_{\mathfrak{v}}^2], \quad m_n^* := \arg \min \{R_n^m(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) : m \in \mathbb{N}\}$$

$$\text{and } R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) := R_n^{m_n^*}(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) = \min \{R_n^m(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) : m \in \mathbb{N}\} \quad (07.07)$$

and $\|\mathbb{V}_{\theta|\mathfrak{s}}^{\theta|\mathfrak{s}}\|_{\mathbb{L}_{\infty}(\nu)} =: \mathbb{V}_{\theta|\mathfrak{s}} \in \mathbb{R}_{\geq 0}$, for each $M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t},d}$ known in advance, for all $\theta \in \mathbb{J}^{\mathfrak{a},r}$, hence $g = \mathfrak{s}\theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \mathbb{J}$, we have

$$\mathbb{E}_{\theta|\mathfrak{s}}^n (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbb{V}}^2) \leq (d^2 \mathbb{V}_{\theta|\mathfrak{s}} + r^2) R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) \quad \forall n \in \mathbb{N}.$$

§07I01.36 **Proof of Corollary §07I01.35.** Given in the lecture. \square

§07I01.37 **Remark.** Under the assumptions of **Corollary §07I01.35** if there exists in addition $\mathfrak{v} \in \mathbb{R}_{\geq 0}$ satisfying $\|\mathbb{V}_{\theta|\mathfrak{s}}^{\theta|\mathfrak{s}}\|_{\mathbb{L}_{\infty}(\nu)} \leq \mathfrak{v}$ for all $\theta \in \mathbb{J}^{\mathfrak{a},r}$ and $M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t},d}$ then

$$\sup \{ \mathbb{E}_{\theta|\mathfrak{s}}^n (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbb{V}}^2) : \theta \in \mathbb{J}^{\mathfrak{a},r}, M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t},d} \} \leq (\mathfrak{v}d^2 + r^2) R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) \quad \forall n \in \mathbb{N}.$$

Arguing similarly as in **Remark §07I01.21** we note that $R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) = o(1)$ as $n \rightarrow \infty$ since $\|\mathfrak{t}^\dagger \mathbb{1}^m\|_{\mathbb{V}} \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and $(\mathfrak{a}\mathfrak{v})_{(m)} = o(1)$ as $m \rightarrow \infty$ by Assumption §07I01.30. Note that the dimension $m_n^* := m_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ as defined in (07.07) does neither depend on the unknown parameter of interest θ nor on the known multiplication operator $M_{\mathfrak{s}}$ but on the classes $\mathbb{J}^{\mathfrak{a},r}$ and $\mathbb{M}_{\mathfrak{t},d}$ only, and thus also the statistic $\widehat{\theta}^{m_n^*}$. In other words, if the regularity of θ is known in advance, then the OPE $\widehat{\theta}^{m_n^*}$ is a feasible estimator. \square

§07I01.38 **Corollary** (GdiSM §07I01.03 continued). Consider $\widehat{g}_n = g_n + n^{-1/2} \dot{B}_n \sim N_{\theta|\mathfrak{s}}^n$ as in Model §07I01.03, where $\dot{B}_n \sim N_{(0,1)}^{\otimes \mathbb{N}}$, $\mathfrak{s}_n \in \ell_{\infty}$, $\theta \in \ell_2$ and hence $g_n = \mathfrak{s}_n \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \ell_2$. Under Assumption §07I01.30 the OPE $\widehat{\theta}_n^{m_n^*} = \mathfrak{s}_n^\dagger \widehat{g}_n \mathbb{1}^{m_n^*} \in \ell_2(\mathfrak{v}^2)$ satisfies

(i) with dimension $m_n^* = m_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v})$ as in (07.06) and constant $C = 1 + r^2$

$$\sup \{ N_{\theta|\mathfrak{s}}^n (\|\widehat{\theta}_n^{m_n^*} - \theta\|_{\mathbb{V}}^2) : \theta \in \ell_2^{\mathfrak{a},r} \} \leq C R_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) \quad \forall n \in \mathbb{N}; \quad (07.08)$$

(ii) with dimension $m_n^* = m_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ as in (07.07) and constant $C = d^2 + r^2$

$$\sup \{ N_{\theta|\mathfrak{s}}^n (\|\widehat{\theta}_n^{m_n^*} - \theta\|_{\mathbb{V}}^2) : \theta \in \ell_2^{\mathfrak{a},r}, M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t},d} \} \leq C R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) \quad \forall n \in \mathbb{N}. \quad (07.09)$$

§07I01.39 **Proof of Corollary §07I01.38.** Given in the lecture. \square

§07I01.40 **Corollary** (diSM §07I01.06 continued). Consider $\widehat{g}_n = g_n + n^{-1/2} \dot{\mathfrak{e}}_n \sim P_{\theta|\mathfrak{s}|\sigma}^n$ as in Model §07I01.06, where $\dot{\mathfrak{e}}_n \sim \otimes_{j \in \mathbb{N}} P_{(0,\sigma_j^2)}$ satisfies (iSM1) with $\|\sigma^2\|_{\ell_{\infty}} =: \mathfrak{v}_{\sigma} \in \mathbb{R}_{>0}$, $\mathfrak{s}_n \in \ell_{\infty}$, $\theta \in \ell_2$ and hence $g_n = \mathfrak{s}_n \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \ell_2$. Under Assumption §07I01.30 the OPE $\widehat{\theta}_n^{m_n^*} = \mathfrak{s}_n^\dagger \widehat{g}_n \mathbb{1}^{m_n^*} \in \ell_2(\mathfrak{v}^2)$ satisfies

(i) with dimension $m_n^* = m_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v})$ as in (07.06) and constant $C = \mathfrak{v}_{\sigma} + r^2$

$$\sup \{ P_{\theta|\mathfrak{s}|\sigma}^n (\|\widehat{\theta}_n^{m_n^*} - \theta\|_{\mathbb{V}}^2) : \theta \in \ell_2^{\mathfrak{a},r} \} \leq C R_n^*(\mathfrak{a}, \mathfrak{s}, \mathfrak{v}) \quad \forall n \in \mathbb{N}; \quad (07.10)$$

(ii) with dimension $m_n^* = m_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ as in (07.07) and constant $C = \mathfrak{v}_{\sigma} d^2 + r^2$

$$\sup \{ P_{\theta|\mathfrak{s}|\sigma}^n (\|\widehat{\theta}_n^{m_n^*} - \theta\|_{\mathbb{V}}^2) : \theta \in \ell_2^{\mathfrak{a},r}, M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t},d} \} \leq C R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) \quad \forall n \in \mathbb{N}. \quad (07.11)$$

§07I01.41 **Proof of Corollary §07I01.40.** Given in the lecture. \square

§07I01.42 **Corollary** (dieMM §07I01.08 continued). Let $\widehat{g}_n = g_n + n^{-1/2} \dot{\mathfrak{e}}_n$ be defined on $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, P_{\theta|\mathfrak{s}}^{\otimes n})$ as in Model §07I01.08, where $\psi \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ satisfies (dieMM1)–(dieMM3) for some $\mathfrak{v}_{\theta|\mathfrak{s}|\psi} \in \mathbb{R}_{\geq 1}$, $\mathfrak{s}_n \in \mathbb{L}_{\infty}(\nu)$, $\theta \in \mathbb{J}$ and hence $g_n = \mathfrak{s}_n \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \mathbb{J}$. Under Assumption §07I01.30 the OPE $\widehat{\theta}_n^{m_n^*} = \mathfrak{s}_n^\dagger \widehat{g}_n \mathbb{1}^{m_n^*} \in \mathbb{L}_2(\mathfrak{v}^2 \nu) P_{\theta|\mathfrak{s}}^{\otimes n}$ -a.s. satisfies

(i) with constant

$$C_{a,r,s} := \sup \{ \mathbb{V}_{\theta|s|\psi} : \theta \in \mathbb{J}^{a,r} \} + r^2$$

and dimension $m_n^* = m_n^*(\mathbf{a}, \mathbf{s}, \mathbf{v})$ as in (07.06)

$$\sup \{ \mathbb{P}_{\theta|s}^{\otimes n} (\| \widehat{\theta}^{m_n^*} - \theta \|_{\mathbf{v}}^2) : \theta \in \mathbb{J}^{a,r} \} \leq C_{a,r,s} R_n^*(\mathbf{a}, \mathbf{s}, \mathbf{v}) \quad \forall n \in \mathbb{N} \quad (07.12)$$

provided $\mathfrak{s}_j^\dagger \mathbb{1}_*^m \in \mathbb{L}_2(\mathbf{v}^2 \nu)$ for all $m \in \mathbb{N}$;

(ii) with constant

$$C_{a,r,t,d} := d^2 \sup \{ \mathbb{V}_{\theta|s|\psi} : \theta \in \mathbb{J}^{a,r}, M_s \in \mathbb{M}_{t,d} \} + r^2$$

and dimension $m_n^* = m_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ as in (07.07)

$$\sup \{ \mathbb{P}_{\theta|s}^{\otimes n} (\| \widehat{\theta}^{m_n^*} - \theta \|_{\mathbf{v}}^2) : \theta \in \mathbb{J}^{a,r}, M_s \in \mathbb{M}_{t,d} \} \leq C_{a,r,t,d} R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \quad \forall n \in \mathbb{N}. \quad (07.13)$$

§07101.43 **Proof** of **Corollary** §07101.42. Given in the lecture. □

§07101.44 **Illustration**. We illustrate the last results considering usual behaviour for $\mathbf{a}, \mathbf{s}, \mathbf{t}, \mathbf{v} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\mathbf{w} \in \{ \mathbf{s}, \mathbf{t} \}$. We distinguish similar to **Illustration** §07101.28 the following two cases **(p)** $\mathbf{w}^\dagger \in \mathbb{L}_2(\mathbf{v}^2 \nu)$, and **(np)** $\mathbf{w}^\dagger \notin \mathbb{L}_2(\mathbf{v}^2 \nu)$. Interestingly, in case **(p)** the bounds in **Proposition** §07101.32 and **Corollary** §07101.35 are parametric, that is, $n R_n^*(\mathbf{a}, \mathbf{w}, \mathbf{v}) = O(1)$, in case **(np)** the bounds are nonparametric, i.e. $\lim_{n \rightarrow \infty} n R_n^*(\mathbf{a}, \mathbf{w}, \mathbf{v}) = \infty$. In case **(np)** consider the following three specifications:

Table 02 [§07]

Order of the oracle rate $R_n^*(\mathbf{a}, \mathbf{w}, \mathbf{v})$ as $n \rightarrow \infty$

$(j \in \mathcal{J})$ $\mathbf{v}_j^2 = j^{2v}$	$(\mathbf{a} \in \mathbb{R}_{>0})$ \mathbf{a}_j^2	$(\mathbf{t} \in \mathbb{R}_{>0})$ \mathbf{w}_j^2	(squared bias) $(\mathbf{a}\mathbf{v})_{(m)}^2$	(variance) $\ \mathbf{w}^\dagger \mathbb{1}_*^m \ _{\mathbf{v}}^2$	m_n^*	$R_n^*(\mathbf{a}, \mathbf{w}, \mathbf{v})$
(o-m) $v \in (-1/2 - t, a)$	j^{-2a}	j^{-2t}	$m^{-2(a-v)}$	$m^{2v+2t+1}$	$n^{\frac{1}{2a+2t+1}}$	$n^{-\frac{2(a-v)}{2a+2t+1}}$
$v + t = -1/2$	j^{-2a}	j^{-2t}	$m^{-2a-2t-1}$	$\log m$	$\left(\frac{n}{\log n} \right)^{\frac{1}{2a+2t+1}}$	$\frac{\log n}{n}$
(o-s) $a - v \in \mathbb{R}_{>0}$	j^{-2a}	$e^{-j^{2t}}$	$m^{-2(a-v)}$	$m^{(1-2(t-v))+} e^{m^{2t}}$	$(\log n)^{\frac{1}{2t}}$	$(\log n)^{-\frac{a-v}{t}}$
(s-m) $v + t + 1/2 \in \mathbb{R}_{>0}$	$e^{-j^{2a}}$	j^{-2t}	$m^{2v} e^{-m^{2a}}$	$m^{2v+2t+1}$	$(\log n)^{\frac{1}{2a}}$	$\frac{(\log n)^{\frac{2t+2v+1}{2a}}}{n}$
$v + t = -1/2$	$e^{-j^{2a}}$	j^{-2t}	$m^{2v} e^{-m^{2a}}$	$\log m$	$(\log n)^{\frac{1}{2a}}$	$\frac{\log \log n}{n}$

We note that in case **(o-m)** and **(s-m)** for $v + t < -1/2$ the rate $R_n^*(\mathbf{a}, \mathbf{w}, \mathbf{v})$ is parametric. □

§07|01|03 Local and maximal local ϕ -risk

Secondly, we measure the accuracy of the OPE $\widehat{\theta}^m := \mathfrak{s}_j^\dagger \widehat{g}^m$ of $\theta^m = \mathfrak{s}_j^\dagger g^m \in \mathbb{J} \mathbb{1}_*^m$ with $g = \mathfrak{s}_j \theta \in \text{dom}(M_{\mathfrak{s}_j})$ by the mean of its local ϕ -error introduced in §04|03|02, i.e. its ϕ -risk.

§07101.45 **Reminder**. If $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\theta \in \text{dom}(\phi \nu)$, then for each $m \in \mathbb{N}$ we have $\theta^m \in \text{dom}(\phi \nu)$ too and $|\phi \nu(\theta) - \phi \nu(\theta^m)| = o(1)$ as $m \rightarrow \infty$ (**Property** §04|03|13). □

§07101.46 **Assumption**. Consider a noisy version $\widehat{g} = g + n^{-1/2} \dot{\epsilon} \sim \mathbb{P}_{\theta|s}^n$ satisfying Assumption §07101.01. In addition

(dSIPI1) $\dot{\epsilon}$ admits a covariance operator, say $\Gamma_{\theta|s} \in \mathbb{L}(\mathbb{J})$, i.e. $\dot{\epsilon} \sim P_{(0, \Gamma_{\theta|s})}$, and

(dSIPI2) $\dot{\varepsilon} \cdot \mathbf{1}^m \in \mathbb{J} \mathbb{P}_{\theta_{\mathfrak{s}}}^n$ -a.s., for each $m \in \mathbb{N}$. \square

§07101.47 **Comment.** Under Assumption §07101.46 and $\phi \in \mathcal{M}_{z_{0,\nu}}(\mathcal{J})$ set $(\mathfrak{s}^\dagger \phi)_\bullet := \mathfrak{s}^\dagger \phi \in \mathcal{M}(\mathcal{J})$. If $\mathfrak{s}^\dagger \mathbf{1}^m \in \mathbb{L}_2(\phi^2 \nu)$ then we have $\mathfrak{s}^\dagger \dot{\varepsilon} \cdot \mathbf{1}^m \in \text{dom}(\phi \nu) \mathbb{P}_{\theta_{\mathfrak{s}}}^n$ -a.s. since $\nu(|(\mathfrak{s}^\dagger \phi)_\bullet \cdot \dot{\varepsilon} \cdot \mathbf{1}^m|) \leq \|\mathfrak{s}^\dagger \mathbf{1}^m\|_\phi \|\dot{\varepsilon} \cdot \mathbf{1}^m\|_{\mathbb{J}} \in \mathbb{R}_{\geq 0} \mathbb{P}_{\theta_{\mathfrak{s}}}^n$ -a.s.. If in addition $\theta \in \text{dom}(\phi \nu)$, and hence $\theta^m \in \text{dom}(\phi \nu)$ (Property §04103.13), then it follows

$$\hat{\theta}^m = \mathfrak{s}^\dagger \hat{g} \cdot \mathbf{1}^m = n^{-1/2} \mathfrak{s}^\dagger \dot{\varepsilon} \cdot \mathbf{1}^m + \theta^m \in \text{dom}(\phi \nu) \quad \mathbb{P}_{\theta_{\mathfrak{s}}}^n\text{-a.s.} \quad (07.14)$$

If $\mathcal{J} \subseteq \mathbb{Z}$ (at most countable) and $\nu_{\mathcal{J}}$ is the counting measure over the index set \mathcal{J} then Assumption §01101.04 and (dSIPI1) $\dot{\varepsilon} \cdot \sim \mathbb{P}_{(0, \Gamma_{\mathfrak{s}})}$ implies $\nu_{\mathfrak{s}}^{\theta_{\mathfrak{s}}} = \mathbb{P}_{\theta_{\mathfrak{s}}}^n(|\dot{\varepsilon} \cdot|^2) \in \mathbb{L}_\infty(\nu_{\mathcal{J}})$ and hence the additional assumption (dSIPI2) $\dot{\varepsilon} \cdot \mathbf{1}^m \in \mathbb{J} = \mathbb{L}_2(\nu_{\mathcal{J}}) \mathbb{P}_{\theta_{\mathfrak{s}}}^n$ -a.s.. However, the last implication does generally not hold, if $\mathcal{J} \in \{\mathbb{R}, \mathbb{R}_{\geq 0}\}$ for example. \square

§07|01|03|01 Local ϕ -risk

§07101.48 **Assumption.** Let $\phi \in \mathcal{M}_{z_{0,\nu}}(\mathcal{J})$, $\theta \in \text{dom}(\phi \nu)$, and $\mathfrak{s}^\dagger \mathbf{1}^m \in \mathbb{L}_2(\phi^2 \nu)$ for all $m \in \mathbb{N}$ be satisfied. \square

§07101.49 **Definition.** Under Assumptions §07101.46 and §07101.48 the *local ϕ -risk* of an OPE $\hat{\theta}^m = \mathfrak{s}^\dagger \hat{g} \cdot \mathbf{1}^m = \mathfrak{s}^\dagger \hat{g} \cdot \mathbf{1}^m \in \text{dom}(\phi \nu) \mathbb{P}_{\theta_{\mathfrak{s}}}^n$ -a.s. satisfies

$$\mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\hat{\theta}^m - \theta)|^2) = \mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\mathfrak{s}^\dagger(\hat{g} - g) \cdot \mathbf{1}^m)|^2) + |\phi \nu(\theta \cdot \mathbf{1}^{m \perp})|^2. \quad (07.15)$$

with *variance* $\mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\mathfrak{s}^\dagger(\hat{g} - g) \cdot \mathbf{1}^m)|^2) = n^{-1} \mathbb{P}_{(0, \Gamma_{\mathfrak{s}})}(|\phi \nu(\mathfrak{s}^\dagger \dot{\varepsilon} \cdot \mathbf{1}^m)|^2)$ and *bias* $|\phi \nu(\theta \cdot \mathbf{1}^{m \perp})|$. \square

§07101.50 **Property.** Under Assumptions §07101.46 and §07101.48 we have

$$\begin{aligned} \mathbb{P}_{(0, \Gamma_{\mathfrak{s}})}(|\phi \nu(\mathfrak{s}^\dagger \dot{\varepsilon} \cdot \mathbf{1}^m)|^2) &= \mathbb{P}_{(0, \Gamma_{\mathfrak{s}})}(|\nu(\dot{\varepsilon} \cdot (\mathfrak{s}^\dagger \phi)_\bullet \cdot \mathbf{1}^m)|^2) \\ &= \langle \Gamma_{\theta_{\mathfrak{s}}}((\mathfrak{s}^\dagger \phi)_\bullet \cdot \mathbf{1}^m), (\mathfrak{s}^\dagger \phi)_\bullet \cdot \mathbf{1}^m \rangle_{\mathbb{J}} =: \|(\mathfrak{s}^\dagger \phi)_\bullet \cdot \mathbf{1}^m\|_{\Gamma_{\theta_{\mathfrak{s}}}}^2 \end{aligned} \quad (07.16)$$

and consequently $\mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\mathfrak{s}^\dagger(\hat{g} - g) \cdot \mathbf{1}^m)|^2) \leq n^{-1} \|\Gamma_{\theta_{\mathfrak{s}}}\|_{\mathbb{L}(\mathbb{J})} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_\phi^2 \in \mathbb{R}_{\geq 0}$. \square

§07101.51 **Proposition (Upper bound).** Under Assumptions §07101.46 and §07101.48 for all $m, n \in \mathbb{N}$ setting

$$\begin{aligned} R_n^m(\theta, \mathfrak{s}, \phi) &:= |\phi \nu(\theta \cdot \mathbf{1}^{m \perp})|^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_\phi^2, \quad m_n^\circ := \arg \min \{R_n^m(\theta, \mathfrak{s}, \phi) : m \in \mathbb{N}\} \\ \text{and } R_n^\circ(\theta, \mathfrak{s}, \phi) &:= R_n^{m_n^\circ}(\theta, \mathfrak{s}, \phi) := \min \{R_n^m(\theta, \mathfrak{s}, \phi) : m \in \mathbb{N}\} \end{aligned} \quad (07.17)$$

we have $\mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\hat{\theta}^{m_n^\circ} - \theta)|^2) \leq (1 \vee \|\Gamma_{\theta_{\mathfrak{s}}}\|_{\mathbb{L}(\mathbb{J})}) R_n^\circ(\theta, \mathfrak{s}, \phi)$ for all $n \in \mathbb{N}$.

§07101.52 **Proof of Proposition §07101.51.** Given in the lecture. \square

§07101.53 **Definition.** Let $\theta \in \text{dom}(\phi \nu)$ and $\hat{\theta}^m \in \text{dom}(\phi \nu) \mathbb{P}_{\theta_{\mathfrak{s}}}^n$ -a.s. for all $m \in \mathbb{N}$. If there exist $C \in \mathbb{R}_{>0}$ and for each $n \in \mathbb{N}$, $R_n^\circ \in \mathbb{R}_{>0}$ and $m_n^\circ \in \mathbb{N}$ satisfying

$$C^{-1} R_n^\circ \leq \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\hat{\theta}^m - \theta)|^2) \leq \mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\hat{\theta}^{m_n^\circ} - \theta)|^2) \leq C R_n^\circ \quad \forall n \in \mathbb{N},$$

then we call R_n° *oracle bound*, m_n° *oracle dimension* and $\hat{\theta}^{m_n^\circ}$ *oracle optimal* (up to the constant C). As a consequence, up to the constant C^2 the statistic $\hat{\theta}^{m_n^\circ}$ attains the lower local ϕ -risk bound within the family of OPE's, that is, $\mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\hat{\theta}^{m_n^\circ} - \theta)|^2) \leq C^2 \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta_{\mathfrak{s}}}^n(|\phi \nu(\hat{\theta}^m - \theta)|^2)$. \square

§07101.54 **Comment.** If $\Gamma_{\theta|s} \in \mathbb{L}(\mathbb{J})$ is invertible with inverse $\Gamma_{\theta|s}^{-1} \in \mathbb{L}(\mathbb{J})$, i.e. $\Gamma_{\theta|s} \Gamma_{\theta|s}^{-1} = \text{id}_{\mathbb{J}} = \Gamma_{\theta|s}^{-1} \Gamma_{\theta|s}$, then we write shortly $1 \leq \max(\|\Gamma_{\theta|s}\|_{\mathbb{L}(\mathbb{J})}, \|\Gamma_{\theta|s}^{-1}\|_{\mathbb{L}(\mathbb{J})}) \leq \mathfrak{v}_{\theta|s} \in \mathbb{R}_{\geq 1}$. In this situation for all $h_{\bullet} \in \mathbb{J}$ we have $\mathfrak{v}_{\theta|s}^{-1} \|h_{\bullet}\|_{\mathbb{J}}^2 \leq \|h_{\bullet}\|_{\Gamma_{\theta|s}}^2 = \langle \Gamma_{\theta|s} h_{\bullet}, h_{\bullet} \rangle_{\mathbb{J}} \leq \mathfrak{v}_{\theta|s} \|h_{\bullet}\|_{\mathbb{J}}^2$. \square

§07101.55 **Oracle inequality.** Under Assumptions §07101.46 and §07101.48 if in addition

$$1 \leq \max(\|\Gamma_{\theta|s}\|_{\mathbb{L}(\mathbb{J})}, \|\Gamma_{\theta|s}^{-1}\|_{\mathbb{L}(\mathbb{J})}) \leq \mathfrak{v}_{\theta|s} \in \mathbb{R}_{\geq 1}$$

is satisfied then (07.17) (and *Comment* §07101.54) implies

$$\begin{aligned} \mathfrak{v}_{\theta|s}^{-1} R_n^m(\theta, \mathfrak{s}, \phi) &\leq \mathbb{E}_{\theta|s}^n (|\phi\nu(\widehat{\theta}^m - \theta)|^2) = n^{-1} \|(\mathfrak{s}^\dagger \phi) \cdot \mathbf{1}^m\|_{\Gamma_{\theta|s}}^2 + |\phi\nu(\theta \mathbf{1}^{m\perp})|^2 \\ &\leq \mathfrak{v}_{\theta|s} R_n^m(\theta, \mathfrak{s}, \phi) \quad \forall m, n \in \mathbb{N}. \end{aligned}$$

As a consequence we immediately obtain the following *oracle inequality*

$$\begin{aligned} \mathfrak{v}_{\theta|s}^{-1} R_n^\circ(\theta, \mathfrak{s}, \phi) &\leq \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta|s}^n (|\phi\nu(\widehat{\theta}^m - \theta)|^2) \leq \mathbb{E}_{\theta|s}^n (|\phi\nu(\widehat{\theta}^{m_n^\circ} - \theta)|^2) \\ &\leq \mathfrak{v}_{\theta|s} R_n^\circ(\theta, \mathfrak{s}, \phi) \leq \mathfrak{v}_{\theta|s}^2 \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta|s}^n (|\phi\nu(\widehat{\theta}^m - \theta)|^2) \quad \forall n \in \mathbb{N}, \quad (07.18) \end{aligned}$$

and hence $R_n^\circ(\theta, \mathfrak{s}, \phi)$, m_n° and the statistic $\widehat{\theta}^{m_n^\circ}$, respectively, is an *oracle bound*, an *oracle dimension* and *oracle optimal* (up to the constant $\mathfrak{v}_{\theta|s}^2$). \square

§07101.56 **Remark.** Arguing similarly as in *Remark* §07101.21 we note that $R_n^\circ(\theta, \mathfrak{s}, \phi) = o(1)$ as $n \rightarrow \infty$, whenever $\|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\phi}^2 \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and $|\phi\nu(\theta \mathbf{1}^{m\perp})| = o(1)$ as $m \rightarrow \infty$. The latter is satisfied, for example, if $\theta = \mathfrak{s}^\dagger g \in \text{dom}(\phi\nu)$. The oracle dimension $m_n^\circ = m_n^\circ(\theta, \mathfrak{s}, \phi)$ as defined in (§07101.51) depends again on the unknown parameter of interest θ , and thus also the oracle optimal statistic $\widehat{\theta}^{m_n^\circ}$. In other words $\widehat{\theta}^{m_n^\circ}$ is not a feasible estimator. \square

§07101.57 **Corollary** (GdiSM §07101.03 continued). Consider $\widehat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta|s}^n$ as in *Model* §07101.03, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$, $\mathfrak{s} \in \ell_\infty$, $\theta \in \ell_2$ and hence $g = \mathfrak{s} \cdot \theta \in \text{dom}(M_s) \subseteq \ell_2$. Given $\phi \in \mathbb{R}_{\lambda_0}^{\mathbb{N}}$ and $\theta \in \text{dom}(\phi\nu_N)$ the (infeasible) OPE $\widehat{\theta}^{m_n^\circ} = \mathfrak{s}^\dagger \widehat{g} \cdot \mathbf{1}^{m_n^\circ} \in \text{dom}(\phi\nu_N)$ with oracle dimension m_n° as in (07.17) satisfies

$$N_{\theta|s}^n (|\phi\nu_N(\widehat{\theta}^{m_n^\circ} - \theta)|^2) = R_n^\circ(\theta, \mathfrak{s}, \phi) = \inf_{m \in \mathbb{N}} N_{\theta|s}^n (|\phi\nu_N(\widehat{\theta}^m - \theta)|^2),$$

and hence it is *oracle optimal* (with constant 1).

§07101.58 **Proof of Corollary** §07101.57. Given in the lecture. \square

§07101.59 **Corollary** (diSM §07101.06 continued). Consider $\widehat{g} = g + n^{-1/2} \dot{\varepsilon} \sim P_{\theta|s|\sigma}^n$ as in *Model* §07101.06, where $\dot{\varepsilon} \sim \otimes_{j \in \mathbb{N}} P_{(0,\sigma^2)}$ satisfies (iSM1) and (iSM2) with $\max(\|\sigma^{-2}\|_{\ell_\infty}, \|\sigma^2\|_{\ell_\infty}) =: \mathfrak{v}_\sigma \in \mathbb{R}_{\geq 1}$, $\mathfrak{s} \in \ell_\infty$, $\theta \in \ell_2$ and hence $g = \mathfrak{s} \cdot \theta \in \text{dom}(M_s) \subseteq \ell_2$. Given $\phi \in \mathbb{R}_{\lambda_0}^{\mathbb{N}}$ and $\theta \in \text{dom}(\phi\nu_N)$ the (infeasible) OPE $\widehat{\theta}^{m_n^\circ} = \mathfrak{s}^\dagger \widehat{g} \cdot \mathbf{1}^{m_n^\circ} \in \text{dom}(\phi\nu_N)$ with oracle dimension m_n° as in (07.17) satisfies

$$P_{\theta|s|\sigma}^n (|\phi\nu_N(\widehat{\theta}^{m_n^\circ} - \theta)|^2) \leq \mathfrak{v}_\sigma R_n^\circ(\theta, \mathfrak{s}, \phi) \leq \mathfrak{v}_\sigma^2 \inf_{m \in \mathbb{N}} P_{\theta|s|\sigma}^n (|\phi\nu_N(\widehat{\theta}^m - \theta)|^2),$$

and hence it is *oracle optimal* (with constant \mathfrak{v}_σ).

§07101.60 **Proof of Corollary** §07101.59. Given in the lecture. \square

§07101.61 **Corollary** (dieMM §07101.08 continued). Let $\widehat{g}_\bullet = g_\bullet + n^{-1/2}\widehat{\epsilon}_\bullet$ be defined on $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, \mathbb{P}_{\theta|\mathfrak{s}}^{\otimes n})$ as in Model §07101.08, where $\psi_\bullet \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ satisfies (dieMM1)–(dieMM4) for some $\nu_{\theta|\mathfrak{s}|\psi} \in \mathbb{R}_{>1}$, $\mathfrak{s}_\bullet \in \mathbb{L}_\infty(\nu)$, $\theta \in \mathbb{J}$ and hence $g_\bullet = \mathfrak{s}_\bullet\theta \in \text{dom}(M_\nu) \subseteq \mathbb{J}$. Under Assumption §07101.48 the (infeasible) OPE $\widehat{\theta}_\bullet^{m_n} = \mathfrak{s}_\bullet^\dagger \widehat{g}_\bullet \mathbb{1}_n^{m_n} \in \text{dom}(\phi\nu)$ with oracle dimension m_n^o as in (07.17) satisfies

$$\mathbb{E}_{\theta|\mathfrak{s}}^n(|\phi\nu(\widehat{\theta}_\bullet^{m_n} - \theta)|^2) \leq \nu_{\theta|\mathfrak{s}|\psi} R_n^o(\theta, \mathfrak{s}_\bullet, \phi) \leq \nu_{\theta|\mathfrak{s}|\psi}^2 \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta|\mathfrak{s}}^n(|\phi\nu(\widehat{\theta}_\bullet^m - \theta)|^2),$$

and hence it is *oracle optimal* (with constant $\nu_{\theta|\mathfrak{s}|\psi}$).

§07101.62 **Proof of Corollary** §07101.61. Given in the lecture. \square

§07101.63 **Illustration**. We illustrate the last results considering usual behaviour for $\theta, \mathfrak{s}_\bullet, \phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$. Similar to the two cases (p) and (np) in Illustration §07101.28 we distinguish here the following two cases

(p) $\mathfrak{s}_\bullet^\dagger \in \mathbb{L}_2(\phi^2\nu)$ or there is $K \in \mathbb{N}$ with $\sup\{|\phi\nu(\theta \mathbb{1}_n^{m\perp})|^2 : m \in \mathbb{N}_{\geq K}\} = 0$,

(np) $\mathfrak{s}_\bullet^\dagger \notin \mathbb{L}_2(\phi^2\nu)$ and for all $m \in \mathbb{N}$ holds $\sup\{|\phi\nu(\theta \mathbb{1}_n^{m\perp})|^2 : m \in \mathbb{N}_{\geq K}\} \in \mathbb{R}_{>0}$.

In case (p) the oracle bound is again parametric, i.e. $nR_n^o(\theta, \mathfrak{s}_\bullet, \phi) = O(1)$, while in case (np) the oracle bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^o(\theta, \mathfrak{s}_\bullet, \phi) = \infty$. In case (np) consider the following three specifications:

Table 03 [§07]

Order of the oracle rate $R_n^o(\theta, \mathfrak{s}_\bullet, \phi)$ as $n \rightarrow \infty$

$(j \in \mathcal{J})$	$(a \in \mathbb{R}_{>0})$	$(t \in \mathbb{R}_{>0})$	(squared bias)	(variance)	m_n^o	$R_n^o(\theta, \mathfrak{s}_\bullet, \phi)$	
$\phi_j = j^{\nu-1/2}$	θ_j	\mathfrak{s}_j^2	$ \phi\nu(\theta \mathbb{1}_n^{m\perp}) ^2$	$\ \mathfrak{s}_\bullet^\dagger \mathbb{1}_n^m\ _\phi^2$			
(o-m)	$\nu \in (-t, a)$	$j^{-a-1/2}$	j^{-2t}	$m^{-2(a-\nu)}$	$m^{2\nu+2t}$	$n^{\frac{1}{2a+2t}}$	$n^{-\frac{a-\nu}{a+t}}$
	$\nu = -t$	$j^{-a-1/2}$	j^{-2t}	$m^{-2(a+t)}$	$\log m$	$\left(\frac{n}{\log n}\right)^{\frac{1}{2(a+t)}}$	$\frac{\log n}{n}$
(o-s)	$a - \nu \in \mathbb{R}_{>0}$	$j^{-a-1/2}$	$e^{-j^{2a}}$	$m^{-2(a-\nu)}$	$m^{2(\nu-t)+e^{m^{2a}}}$	$(\log n)^{\frac{1}{2t}}$	$(\log n)^{-\frac{a-\nu}{t}}$
(s-m)	$\nu + t \in \mathbb{R}_{>0}$	$e^{-j^{2a}}$	j^{-2t}	$m^{(1-4a+2\nu)+e^{-2m^{2a}}}$	$m^{2\nu+2t}$	$(\log n)^{\frac{1}{2a}}$	$\frac{(\log n)^{\frac{t+\nu}{a}}}{n}$
	$\nu = -t$	$e^{-j^{2a}}$	j^{-2t}	$m^{(1-4a-2t)+e^{-2m^{2a}}}$	$\log m$	$(\log n)^{\frac{1}{2a}}$	$\frac{\log \log n}{n}$

We note that in case (o-m) and (s-m) for $\nu < -t$ the oracle rate $R_n^o(\theta, \mathfrak{s}_\bullet, \phi)$ is parametric. \square

§07|01|03|02 Maximal local ϕ -risk

§07101.64 **Assumption**. Consider weights $\mathfrak{a}_\bullet, \mathfrak{t}_\bullet \in \mathcal{M}_{>0, \nu}(\mathcal{J})$ and $\phi_\bullet \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ (i.e. $\nu(\mathcal{N}_\mathfrak{a}) = \nu(\mathcal{N}_\mathfrak{t}) = 0 = \nu(\mathcal{N}_\phi)$), such that $\mathfrak{a}_\bullet, \mathfrak{t}_\bullet \in \mathbb{L}_\infty(\nu)$, $\mathfrak{a}_\bullet \in \mathbb{L}_2(\phi^2\nu)$, and $\mathfrak{t}_\bullet^\dagger \mathbb{1}_n^m \in \mathbb{L}_2(\phi^2\nu)$ for all $m \in \mathbb{N}$. \square

§07101.65 **Reminder**. Under Assumption §07101.64 we have $\mathbb{J}^\mathfrak{a} = \mathbb{L}_2(\nu) = \text{dom}(M_\nu) = \mathfrak{J}\mathfrak{a}_\bullet \subseteq \mathbb{J}$ and the three measures ν , $\mathfrak{a}_\bullet^{2\dagger}\nu$ and $|\phi_\bullet|\nu$ dominate mutually each other, i.e. they share the same null sets (see Property §04101.02). We consider $\mathbb{J}^\mathfrak{a}$ endowed with $\|\cdot\|_{\mathfrak{a}_\bullet^\dagger} = \|M_{\mathfrak{a}_\bullet^\dagger}\cdot\|_{\mathbb{J}}$ and given a constant $r \in \mathbb{R}_{>0}$ the ellipsoid $\mathbb{J}^{\mathfrak{a}_\bullet, r} := \{h_\bullet \in \mathbb{J}^\mathfrak{a} : \|h_\bullet\|_{\mathfrak{a}_\bullet^\dagger} \leq r\} \subseteq \mathbb{J}^\mathfrak{a}$. Since $\mathfrak{a}_\bullet \in \mathbb{L}_2(\phi^2\nu)$, and hence $\|\mathfrak{a}_\bullet \mathbb{1}_n^{m\perp}\|_\phi = \|(\mathfrak{a}_\bullet\phi)_\bullet \mathbb{1}_n^{m\perp}\|_{\mathbb{J}} \in \mathbb{R}_{>0}$ for each $m \in \mathbb{N}$ ($\|\mathfrak{a}_\bullet \mathbb{1}_n^{m\perp}\|_\phi = o(1)$ as $m \rightarrow \infty$ by dominated convergence) we have $\mathbb{J}^\mathfrak{a} \subseteq \text{dom}(\phi\nu)$ (Property §04102.23), and $|\phi\nu(\theta \mathbb{1}_n^{m\perp})| \leq r \|\mathfrak{a}_\bullet \mathbb{1}_n^{m\perp}\|_\phi$ for all $\theta \in \mathbb{J}^{\mathfrak{a}_\bullet, r}$ (Lemma §04102.25). Consequently, if Assumption §07101.64, $\theta_\bullet \in \mathbb{J}^{\mathfrak{a}_\bullet, r}$ and $\mathfrak{s}_\bullet^\dagger \mathbb{1}_n^m \in \mathbb{L}_2(\phi^2\nu)$ for all $m \in \mathbb{N}$ are satisfied, then Assumption §07101.48 is fulfilled. Moreover, under Assumption §07101.64 for each $M_\mathfrak{s} \in \mathbb{M}_{\mathfrak{t}_\bullet, \mathfrak{d}}$ we have $\|\mathfrak{s}_\bullet^\dagger \mathbb{1}_n^m\|_\phi \leq d \|\mathfrak{t}_\bullet^\dagger \mathbb{1}_n^m\|_\phi \in \mathbb{R}_{>0}$ for all $m \in \mathbb{N}$ (Definition §04103.05). Therefore, if Assumption §07101.64, $\theta_\bullet \in \mathbb{J}^{\mathfrak{a}_\bullet, r}$ and $M_\mathfrak{s} \in \mathbb{M}_{\mathfrak{t}_\bullet, \mathfrak{d}}$ are satisfied, then Assumption §07101.48 is also fulfilled. \square

§07101.66 **Proposition (Upper bound).** Under Assumptions §07101.46 and §07101.64 let $\mathfrak{s}^\dagger \mathbf{1}_*^m \in \mathbb{L}_2(\phi^2 \nu)$ for all $m \in \mathbb{N}$. Setting for $n, m \in \mathbb{N}$

$$\begin{aligned} R_n^m(\mathfrak{a}, \mathfrak{s}, \phi) &:= \|\mathfrak{a} \mathbf{1}_*^{m\perp}\|_\phi^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}_*^m\|_\phi^2, \quad m_n^* := \arg \min \{R_n^m(\mathfrak{a}, \mathfrak{s}, \phi) : m \in \mathbb{N}\} \\ \text{and } R_n^*(\mathfrak{a}, \mathfrak{s}, \phi) &:= R_n^{m_n^*}(\mathfrak{a}, \mathfrak{s}, \phi) = \min \{R_n^m(\mathfrak{a}, \mathfrak{s}, \phi) : m \in \mathbb{N}\} \end{aligned} \quad (07.19)$$

and $\|\Gamma_{\theta|\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} =: \mathfrak{v}_{\theta|\mathfrak{s}} \in \mathbb{R}_{\geq 0}$, for all $\theta = \mathfrak{s}^\dagger g \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}$ we have

$$\mathbb{E}_{\theta|\mathfrak{s}}^n (|\phi \nu(\widehat{\theta}^{m_n^*} - \theta)|^2) \leq (\mathfrak{v}_{\theta|\mathfrak{s}} \vee r^2) R_n^*(\mathfrak{a}, \mathfrak{s}, \phi) \quad \forall n \in \mathbb{N}.$$

§07101.67 **Proof of Proposition §07101.66.** Given in the lecture. \square

§07101.68 **Remark.** Under the assumptions of Proposition §07101.66 if there exists in addition $\mathfrak{v}_\mathfrak{s} \in \mathbb{R}_{>0}$ satisfying $\|\Gamma_{\theta|\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} \leq \mathfrak{v}_\mathfrak{s}$ for all $\theta \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}$ then

$$\sup \{ \mathbb{E}_{\theta|\mathfrak{s}}^n (|\phi \nu(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}} \} \leq (\mathfrak{v}_\mathfrak{s} \vee r^2) R_n^*(\mathfrak{a}, \mathfrak{s}, \phi) \quad \forall n \in \mathbb{N}.$$

Arguing similarly as in Remark §07101.21 we note that $R_n^*(\mathfrak{a}, \mathfrak{s}, \phi) = o(1)$ as $n \rightarrow \infty$, whenever $\|\mathfrak{s}^\dagger \mathbf{1}_*^m\|_\phi^2 \in \mathbb{R}_{>0}$ for all $m \in \mathbb{N}$ and $\|\mathfrak{a} \mathbf{1}_*^{m\perp}\|_\phi = o(1)$ as $m \rightarrow \infty$. The latter is satisfied since $\mathfrak{a} \in \mathbb{L}_2(\phi^2 \nu)$ by Assumption §07101.64. Note that the dimension $m_n^* := m_n^*(\mathfrak{a}, \mathfrak{s}, \phi)$ as defined in (07.19) does not depend on the unknown parameter of interest θ but on the class $\mathbb{J}^{\mathfrak{a},\mathfrak{r}}$ only, and thus also the statistic $\widehat{\theta}^{m_n^*}$. In other words, if the regularity of θ is known in advance, then the OPE $\widehat{\theta}^{m_n^*}$ is a feasible estimator. \square

§07101.69 **Corollary (Upper bound).** Under Assumptions §07101.46 and §07101.64 setting for $n, m \in \mathbb{N}$

$$\begin{aligned} R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) &:= \|\mathfrak{a} \mathbf{1}_*^{m\perp}\|_\phi^2 + n^{-1} \|\mathfrak{t}^\dagger \mathbf{1}_*^m\|_\phi^2, \quad m_n^* := \arg \min \{R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) : m \in \mathbb{N}\} \\ \text{and } R_n^*(\mathfrak{a}, \mathfrak{t}, \phi) &:= R_n^{m_n^*}(\mathfrak{a}, \mathfrak{t}, \phi) = \min \{R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) : m \in \mathbb{N}\} \end{aligned} \quad (07.20)$$

and $\|\Gamma_{\theta|\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} =: \mathfrak{v}_{\theta|\mathfrak{s}} \in \mathbb{R}_{\geq 0}$, for each $M_\mathfrak{s} \in \mathbb{M}_{\mathfrak{t},\mathfrak{d}}$ known in advance, for all $\theta \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}$, hence $g = \mathfrak{s} \theta \in \text{dom}(M_\mathfrak{s}) \subseteq \mathbb{J}$, we have

$$\mathbb{E}_{\theta|\mathfrak{s}}^n (|\phi \nu(\widehat{\theta}^{m_n^*} - \theta)|^2) \leq (d^2 \mathfrak{v}_{\theta|\mathfrak{s}} \vee r^2) R_n^*(\mathfrak{a}, \mathfrak{t}, \phi) \quad \forall n \in \mathbb{N}.$$

§07101.70 **Proof of Corollary §07101.69.** Given in the lecture. \square

§07101.71 **Remark.** Under the assumptions of Corollary §07101.69 if there exists in addition $\mathfrak{v} \in \mathbb{R}_{>0}$ satisfying $\|\Gamma_{\theta|\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} \leq \mathfrak{v}$ for all $\theta \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}$ and $M_\mathfrak{s} \in \mathbb{M}_{\mathfrak{t},\mathfrak{d}}$ then

$$\sup \{ \mathbb{E}_{\theta|\mathfrak{s}}^n (|\phi \nu(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}, M_\mathfrak{s} \in \mathbb{M}_{\mathfrak{t},\mathfrak{d}} \} \leq (\mathfrak{v} d^2 \vee r^2) R_n^*(\mathfrak{a}, \mathfrak{t}, \phi) \quad \forall n \in \mathbb{N}.$$

Arguing similarly as in Remark §07101.21 we note that $R_n^*(\mathfrak{a}, \mathfrak{t}, \phi) = o(1)$ as $n \rightarrow \infty$ since $\|\mathfrak{t}^\dagger \mathbf{1}_*^m\|_\phi \in \mathbb{R}_{>0}$ for all $m \in \mathbb{N}$ and $\|\mathfrak{a} \mathbf{1}_*^{m\perp}\|_\phi = o(1)$ as $m \rightarrow \infty$ by Assumption §07101.64. Note that the dimension $m_n^* := m_n^*(\mathfrak{a}, \mathfrak{t}, \phi)$ as defined in (07.20) does neither depend on the unknown parameter of interest θ nor on the known multiplication operator $M_\mathfrak{s}$ but on the classes $\mathbb{J}^{\mathfrak{a},\mathfrak{r}}$ and $\mathbb{M}_{\mathfrak{t},\mathfrak{d}}$ only, and thus also the statistic $\widehat{\theta}^{m_n^*}$. In other words, if the regularity of θ is known in advance, then the OPE $\widehat{\theta}^{m_n^*}$ is a feasible estimator. \square

§07101.72 **Corollary (GdiSM §07101.03 continued).** Consider $\widehat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta|\mathfrak{s}}^n$ as in Model §07101.03, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$, $\mathfrak{s} \in \ell_\infty$, $\theta \in \ell_2$ and hence $g = \mathfrak{s} \theta \in \text{dom}(M_\mathfrak{s}) \subseteq \ell_2$. Under Assumption §07101.64 the OPE $\widehat{\theta}^{m_n^*} = \mathfrak{s}^\dagger \widehat{g} \mathbf{1}_*^{m_n^*} \in \text{dom}(\phi \nu_n)$ satisfies

(i) with dimension $m_n^* = m_n^*(\mathbf{a}, \mathfrak{s}, \phi)$ as in (07.19) and constant $C_r = 1 \vee r^2$

$$\sup \{N_{\theta|\mathfrak{s}}^n (|\phi_{\mathcal{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{\mathfrak{a},r}\} \leq C_r R_n^*(\mathbf{a}, \mathfrak{s}, \phi) \quad \forall n \in \mathbb{N} \quad (07.21)$$

(ii) with dimension $m_n^* = m_n^*(\mathbf{a}, \mathfrak{t}, \phi)$ as in (07.20) and constant $C_{r,d} = d^2 \vee r^2$

$$\sup \{N_{\theta|\mathfrak{s}}^n (|\phi_{\mathcal{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{\mathfrak{a},r}, M_s \in \mathbb{M}_{t,d}\} \leq C_{r,d} R_n^*(\mathbf{a}, \mathfrak{t}, \phi) \quad \forall n \in \mathbb{N}. \quad (07.22)$$

§07101.73 **Proof of Corollary** §07101.72. Given in the lecture. \square

§07101.74 **Corollary** (diSM §07101.06 continued). Consider $\widehat{g} = g + n^{-1/2}\dot{\epsilon} \sim P_{\theta|\mathfrak{s}\sigma}^n$ as in Model §07101.06, where $\dot{\epsilon} \sim P_{(q, M_s)}$ satisfies (iSM1) with $\|\alpha^2\|_{\ell_\infty} =: \nu_\sigma \in \mathbb{R}_{>0}$, $\mathfrak{s} \in \ell_\infty$, $\theta \in \ell_2$ and hence $g = \mathfrak{s}\theta \in \text{dom}(M_s) \subseteq \ell_2$. Under Assumption §07101.64 the OPE $\widehat{\theta}^{m_n^*} = \mathfrak{s}^\dagger \widehat{g} \mathbb{1}^{m_n^*} \in \text{dom}(\phi_{\mathcal{N}})$ satisfies

(i) with dimension $m_n^* = m_n^*(\mathbf{a}, \mathfrak{s}, \phi)$ as in (07.19) and constant $C_{r,\sigma} = \nu_\sigma \vee r^2$

$$\sup \{P_{\theta|\mathfrak{s}\sigma}^n (|\phi_{\mathcal{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{\mathfrak{a},r}\} \leq C_{r,\sigma} R_n^*(\mathbf{a}, \mathfrak{s}, \phi) \quad \forall n \in \mathbb{N} \quad (07.23)$$

(ii) with dimension $m_n^* = m_n^*(\mathbf{a}, \mathfrak{t}, \phi)$ as in (07.20) and constant $C_{r,d,\sigma} = \nu_\sigma d^2 \vee r^2$

$$\sup \{P_{\theta|\mathfrak{s}\sigma}^n (|\phi_{\mathcal{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{\mathfrak{a},r}, M_s \in \mathbb{M}_{t,d}\} \leq C R_n^*(\mathbf{a}, \mathfrak{t}, \phi) \quad \forall n \in \mathbb{N}. \quad (07.24)$$

§07101.75 **Proof of Corollary** §07101.74. Given in the lecture. \square

§07101.76 **Corollary** (dieMM §07101.08 continued). Let $\widehat{g} = g + n^{-1/2}\dot{\epsilon}$ be defined on $(\mathbb{Z}^n, \mathcal{Z}^{\otimes n}, P_{\theta|\mathfrak{s}}^{\otimes n})$ as in Model §07101.08, where $\psi \in \mathcal{M}(\mathcal{X} \otimes \mathcal{Y})$ satisfies (dieMM1)–(dieMM3) for some $\nu_{\theta|\mathfrak{s}\psi} \in \mathbb{R}_{\geq 1}$, $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$, $\theta \in \mathbb{J}$ and hence $g = \mathfrak{s}\theta \in \text{dom}(M_s) \subseteq \mathbb{J}$. Under Assumption §07101.64 the OPE $\widehat{\theta}^{m_n^*} = \mathfrak{s}^\dagger \widehat{g} \mathbb{1}^{m_n^*} \in \text{dom}(\phi_\nu) P_{\theta|\mathfrak{s}}^{\otimes n}$ -a.s. satisfies

(i) with constant

$$C_{a,r,s} := \sup \{\nu_{\theta|\mathfrak{s}\psi} : \theta \in \mathbb{J}^{\mathfrak{a},r}\} \vee r^2$$

and with dimension $m_n^* = m_n^*(\mathbf{a}, \mathfrak{s}, \phi)$ as in (07.19)

$$\sup \{P_{\theta|\mathfrak{s}}^{\otimes n} (|\phi_{\mathcal{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \mathbb{J}^{\mathfrak{a},r}\} \leq C_{a,r,s} R_n^*(\mathbf{a}, \mathfrak{s}, \phi) \quad \forall n \in \mathbb{N} \quad (07.25)$$

provided $\mathfrak{s}^\dagger \mathbb{1}^m \in \mathbb{L}_2(\phi^2 \nu)$ for all $m \in \mathbb{N}$;

(ii) with constant

$$C_{a,r,t,d} := d^2 \sup \{\nu_{\theta|\mathfrak{s}\psi} : \theta \in \mathbb{J}^{\mathfrak{a},r}, M_s \in \mathbb{M}_{t,d}\} \vee r^2$$

and dimension $m_n^* = m_n^*(\mathbf{a}, \mathfrak{t}, \phi)$ as in (07.20)

$$\sup \{P_{\theta|\mathfrak{s}}^{\otimes n} (|\phi_{\mathcal{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \mathbb{J}^{\mathfrak{a},r}, M_s \in \mathbb{M}_{t,d}\} \leq C_{a,r,t,d} R_n^*(\mathbf{a}, \mathfrak{t}, \phi) \quad \forall n \in \mathbb{N}. \quad (07.26)$$

§07101.77 **Proof of Corollary** §07101.76. Given in the lecture. \square

§07101.78 **Illustration.** We illustrate the last results considering usual behaviour for $\mathbf{a}, \mathfrak{s}, \mathfrak{t}, \phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{S})$ and $w \in \{\mathfrak{s}, \mathfrak{t}\}$. We distinguish the following two cases **(p)** $w^\dagger \in \mathbb{L}_2(\phi^2 \nu)$, and **(np)** $w^\dagger \notin \mathbb{L}_2(\phi^2 \nu)$. Interestingly, in case **(p)** the bound in Proposition §07101.66 is parametric, that is, $nR_n^*(\mathbf{a}, w, \phi) = O(1)$, in case **(np)** the bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^*(\mathbf{a}, w, \phi) = \infty$. In case **(np)** consider the following three specifications:

Table 04 [§07]

Order of the rate $R_n^*(\mathbf{a}, \mathbf{w}, \phi)$ as $n \rightarrow \infty$

	$(j \in \mathcal{J})$	$(\mathbf{a} \in \mathbb{R}_{>0})$	$(t \in \mathbb{R}_{>0})$	(squared bias)	(variance)	m_n^*	$R_n^*(\mathbf{a}, \mathbf{w}, \phi)$
	$\phi_j^2 = j^{2v-1}$	\mathbf{a}_j^2	w_j^2	$\ \mathbf{a} \cdot \mathbb{1}^{m \cdot}\ _\phi^2$	$\ \mathbf{w} \cdot \mathbb{1}^m\ _\phi^2$		
(o-m)	$v \in (-t, \mathbf{a})$	$j^{-2\mathbf{a}}$	j^{-2t}	$m^{-2(\mathbf{a}-v)}$	m^{2v+2t}	$n^{\frac{1}{2\mathbf{a}+2t}}$	$n^{-\frac{\mathbf{a}-v}{\mathbf{a}+t}}$
	$v = -t$	$j^{-2\mathbf{a}}$	j^{-2t}	$m^{-2(\mathbf{a}+t)}$	$\log m$	$\left(\frac{n}{\log n}\right)^{\frac{1}{2(\mathbf{a}+t)}}$	$\frac{\log n}{n}$
(o-s)	$\mathbf{a} - v \in \mathbb{R}_{>0}$	$j^{-2\mathbf{a}}$	$e^{-j^{2t}}$	$m^{-2(\mathbf{a}-v)}$	$m^{2(v-t)+} e^{m^{2t}}$	$(\log n)^{\frac{1}{2t}}$	$(\log n)^{-\frac{\mathbf{a}-v}{t}}$
(s-m)	$v + t \in \mathbb{R}_{>0}$	$e^{-j^{2\mathbf{a}}}$	j^{-2t}	$e^{-m^{2\mathbf{a}}}$	m^{2v+2t}	$(\log n)^{\frac{1}{2\mathbf{a}}}$	$\frac{(\log n)^{\frac{t+v}{\mathbf{a}}}}{n}$
	$v = -t$	$e^{-j^{2\mathbf{a}}}$	j^{-2t}	$e^{-m^{2\mathbf{a}}}$	$\log m$	$(\log n)^{\frac{1}{2\mathbf{a}}}$	$\frac{\log \log n}{n}$

We note that in case **(o-m)** and **(s-m)** for $v < -t$ the rate $R_n^*(\mathbf{a}, \mathbf{w}, \phi)$ is parametric. \square

§07|02 Diagonal statistical inverse problem with noisy operator

§07|02.01 **Assumption.** Consider stochastic processes $\dot{\epsilon}_j = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ and $\dot{\eta}_j = (\dot{\eta}_j)_{j \in \mathcal{J}}$ on a probability space $(\Omega, \mathcal{A}, \mathbb{P})$ satisfying Assumption §01|01.04 (i.e. $\dot{\epsilon}_j, \dot{\eta}_j \in \mathcal{M}(\mathcal{A} \otimes \mathcal{J})$) with *mean zero* (i.e. $\mathbb{P}(\dot{\epsilon}_j) = 0_j = \mathbb{P}(\dot{\eta}_j)$), sample sizes $n, k \in \mathbb{N}$ and let Assumption §07|00.02 and in addition $\mathfrak{s}_j \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ be satisfied where $\mathfrak{s}_j \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ is *not known* anymore. The observable noisy image and operator, respectively, has mean $g_j = \mathfrak{s}_j \theta_j \in \mathbb{J} = \mathbb{L}_2(\nu)$ and mean-function $\mathfrak{s}_j \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$, and takes the form $\widehat{g}_j = g_j + n^{-1/2} \dot{\epsilon}_j$ and $\widehat{\mathfrak{s}}_j = \mathfrak{s}_j + k^{-1/2} \dot{\eta}_j$. We denote by $\mathbb{P}_{\theta_j \mathfrak{s}_j}^{n, k}$ the joint distribution of $(\widehat{g}_j, \widehat{\mathfrak{s}}_j)$. Denoting by $\mathbb{P}_{\theta_j}^n$ and $\mathbb{P}_{\mathfrak{s}_j}^k$ the marginal distribution of \widehat{g}_j and $\widehat{\mathfrak{s}}_j$, respectively, if $\dot{\epsilon}_j$ and $\dot{\eta}_j$ are *independent* then we write $\mathbb{P}_{\theta_j \mathfrak{s}_j}^{n, k} = \mathbb{P}_{\theta_j}^n \otimes \mathbb{P}_{\mathfrak{s}_j}^k$ for the joint product distribution of $(\widehat{g}_j, \widehat{\mathfrak{s}}_j)$. \square

§07|02.02 **Comment.** We restrict ourselves in this section to the case $\mathfrak{s}_j \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ only, which ensure identification of the solution θ_j of the equation $g_j = \mathfrak{s}_j \theta_j$. \square

§07|02.03 **Notation.** Introduce the random index set $\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\} := \{j \in \mathcal{J} : k \widehat{\mathfrak{s}}_j^2 \geq 1\} \in \mathcal{J}$, for each $j \in \mathcal{J}$ the elementary random variable $\mathbb{1}_j^{\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\}}$ taking the value one on the event $\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\}$ and zero otherwise, and the stochastic process $\mathbb{1}_j^{\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\}} := (\mathbb{1}_j^{\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\}})_{j \in \mathcal{J}} \in \mathcal{M}(\mathcal{A} \otimes \mathcal{J})$ satisfying hence Assumption §01|01.04. Furthermore, we define $\widehat{\mathfrak{s}}_j^{(k)} := \widehat{\mathfrak{s}}_j \mathbb{1}_j^{\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\}}$ and denote its Moore-Penrose inverse by $\widehat{\mathfrak{s}}_j^{(k)\dagger} := \widehat{\mathfrak{s}}_j^{-1} \mathbb{1}_j^{\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\}}$. We eventually use the elementary identity $\widehat{\mathfrak{s}}_j^{(k)} \widehat{\mathfrak{s}}_j^{(k)\dagger} = \mathbb{1}_j^{\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\}} = \widehat{\mathfrak{s}}_j^{(k)\dagger} \widehat{\mathfrak{s}}_j^{(k)}$ and the upper bound $\|\widehat{\mathfrak{s}}_j^{(k)\dagger}\|_{\mathbb{L}_\infty(\nu)} \leq k^{1/2}$. \square

§07|02.04 **Definition.** Under Assumption §07|02.01 for $\theta_j \in \mathbb{J}$ let $(\widehat{g}_j, \widehat{\mathfrak{s}}_j) \sim \mathbb{P}_{\theta_j \mathfrak{s}_j}^{n, k}$ be noisy versions of $g_j = \mathfrak{s}_j \theta_j \in \text{dom}(M_{\mathfrak{s}_j})$ and $\mathfrak{s}_j \in \mathbb{L}_\infty(\nu)$. For each $m \in \mathbb{N}$ we call $\widehat{\theta}_j^m := \widehat{\mathfrak{s}}_j^{(k)\dagger} \widehat{g}_j^m = \widehat{\mathfrak{s}}_j^{\dagger} \mathbb{1}_j^{\{\widehat{\mathfrak{s}}_j^2 \geq k^{-1}\}} \widehat{g}_j \mathbb{1}_j^m$ *thresholded orthogonal projection estimator (tOPE)* of $\theta_j = \mathfrak{s}_j^\dagger g_j \in \mathbb{J}$ where $\widehat{g}_j^m = \widehat{g}_j \mathbb{1}_j^m$ is an orthogonal projection estimator (OPE) of g_j . \square

§07|02|01 Examples

§07|02.05 **GdiSM with noisy operator (§02|04.06 continued).** Considering $\mathbb{J} = \ell_2 = \mathbb{L}_2(\mathbb{N}, \nu_{\mathbb{N}})$ let Assumption §07|00.02 be satisfied where $\mathfrak{s}_j \in \mathcal{S} \subseteq \mathbb{R}_{\setminus 0}^{\mathbb{N}} \cap \ell_\infty$ is *not known* anymore. We illustrate the tOPE in a Gaussian diagonal inverse sequence model (GdiSM) with noisy operator as in §02|04.06. Here the observable process $\widehat{\mathfrak{s}}_j = \mathfrak{s}_j + k^{-1/2} \dot{W}_j \sim N_{\mathfrak{s}_j}^k$ and $\widehat{g}_j = g_j + n^{-1/2} \dot{B}_j \sim N_{\theta_j \mathfrak{s}_j}^n$ is a noisy version of $\mathfrak{s}_j \in \mathcal{S} \subseteq \mathbb{R}_{\setminus 0}^{\mathbb{N}} \cap \ell_\infty$ and $g_j = \mathfrak{s}_j \theta_j \in \text{dom}(M_{\mathfrak{s}_j}) \subseteq \ell_2$ with $\theta_j \in \Theta \subseteq \ell_2$, respec-

tively, where $\dot{B}_\cdot \sim N_{(0,1)}^{\otimes N}$ and $\dot{W}_\cdot \sim N_{(0,1)}^{\otimes N}$ are *independent*. Consequently, $(\widehat{g}_\cdot, \widehat{\mathfrak{s}}_\cdot)$ admits a joint $N_{\theta|\mathfrak{s}}^{n \otimes k} = N_{\theta|\mathfrak{s}}^n \otimes N_{\mathfrak{s}}^k$ distribution belonging to the family $N_{\Theta \times \mathfrak{S}}^{n \otimes k} := (N_{\theta|\mathfrak{s}}^n \otimes N_{\mathfrak{s}}^k)_{\theta \in \Theta, \mathfrak{s} \in \mathfrak{S}}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{N^2}, \mathcal{B}^{\otimes N^2}, N_{\Theta \times \mathfrak{S}}^{n \otimes k})$ where $\Theta \subseteq \ell_2$ and $\mathfrak{S} \subseteq \mathbb{R}_{\geq 0}^N \cap \ell_\infty$. \square

§07102.06 **Property** (GdiSM with noisy operator §07102.05 continued). *For $\dot{W}_\cdot := (\dot{W}_j)_{j \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ we have $N_{(0,1)} \in \mathcal{W}_4(\mathcal{B})$ with $\mathbf{31}_\cdot = N_{(0,1)}^{\otimes \mathbb{N}}(\dot{W}_\cdot^4)$, $\mathbf{1}_\cdot = N_{(0,1)}^{\otimes \mathbb{N}}(\dot{W}_\cdot^2)$, and $\mathbf{0}_\cdot = N_{(0,1)}^{\otimes \mathbb{N}}(\dot{W}_\cdot)$.* \square

§07102.07 **diSM with noisy operator** (§02104.05 continued). For $\mathbb{J} = \ell_2$ let Assumption §07100.02 be satisfied where $\mathfrak{s}_\cdot \in \mathfrak{S} \subseteq \mathbb{R}_{\geq 0}^N \cap \ell_\infty$ is *not known* anymore. We illustrate the tOPE in a Diagonal inverse sequence model (diSM) with noisy operator as in §02104.05. Here the observable stochastic process $\widehat{\mathfrak{s}}_\cdot = \mathfrak{s}_\cdot + k^{-1/2} \dot{\eta}_\cdot$ and $\widehat{g}_\cdot = g_\cdot + n^{-1/2} \dot{\varepsilon}_\cdot$ is a noisy version of $\mathfrak{s}_\cdot \in \mathfrak{S} \subseteq \mathbb{R}_{\geq 0}^N \cap \ell_\infty$ and $g_\cdot = \mathfrak{s}_\cdot \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \ell_2$ with $\theta \in \Theta \subseteq \ell_2$, respectively, where $\dot{\varepsilon}_\cdot \sim \otimes_{j \in \mathbb{N}} P_{\mathfrak{s}_j}^{\varepsilon_j}$ and $\dot{\eta}_\cdot \sim \otimes_{j \in \mathbb{N}} P_{\mathfrak{s}_j}^{\eta_j}$ are *independent*. In addition, let $\dot{\varepsilon}_\cdot$ satisfy (iSM1) of Model §07101.06 for $\sigma \in \Sigma \subseteq \mathbb{R}_{> 0}^N \cap \ell_\infty$ and (diSMnO1) for $\xi_\cdot \in \Xi \subseteq \mathbb{R}_{> 0}^N \cap \ell_\infty$ we have $P^{\dot{\eta}} \in \mathcal{W}_4(\mathcal{B})$ with $\xi_j^4 = P(\dot{\eta}_j^4)$ and $0 = P(\dot{\eta}_j)$ for all $j \in \mathbb{N}$.

Under (iSM1) \widehat{g}_\cdot admits a $P_{\theta|\mathfrak{s}|\sigma}^n$ -distribution belonging to the family $P_{\Theta \times \mathfrak{S} \times \Sigma}^n := (P_{\theta|\mathfrak{s}|\sigma}^n)_{\theta \in \Theta, \mathfrak{s} \in \mathfrak{S}, \sigma \in \Sigma}$ and under (diSMnO1) $\widehat{\mathfrak{s}}_\cdot$ admits a $P_{\mathfrak{s}|\xi}^k$ -distribution belonging to the family $P_{\mathfrak{S} \times \Xi}^k := (P_{\mathfrak{s}|\xi}^k)_{\mathfrak{s} \in \mathfrak{S}, \xi \in \Xi}$. Consequently, $(\widehat{g}_\cdot, \widehat{\mathfrak{s}}_\cdot)$ admits a joint $P_{\theta|\mathfrak{s}|\sigma|\xi}^{n \otimes k} = P_{\theta|\mathfrak{s}|\sigma}^n \otimes P_{\mathfrak{s}|\xi}^k$ distribution belonging to the family $P_{\Theta \times \mathfrak{S} \times \Sigma \times \Xi}^{n \otimes k} := (P_{\theta|\mathfrak{s}|\sigma}^n \otimes P_{\mathfrak{s}|\xi}^k)_{\theta \in \Theta, \mathfrak{s} \in \mathfrak{S}, \sigma \in \Sigma, \xi \in \Xi}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{N^2}, \mathcal{B}^{\otimes N^2}, P_{\Theta \times \mathfrak{S} \times \Sigma \times \Xi}^{n \otimes k})$ where $\Sigma, \Xi \subseteq \mathbb{R}_{> 0}^N \cap \ell_\infty$, $\mathfrak{S} \subseteq \mathbb{R}_{\geq 0}^N \cap \ell_\infty$ and $\Theta \subseteq \ell_2$. \square

§07102.08 **Property** (diSM with noisy operator §07102.07 continued). *Under (diSMnO1) the process $\dot{\eta}_\cdot \sim \otimes_{j \in \mathbb{N}} P_{\mathfrak{s}_j}^{\eta_j}$ satisfies $P^{\dot{\eta}} \in \mathcal{W}_4(\mathcal{B})$ with $\xi_j^4 = P^{\dot{\eta}}(\dot{\eta}_j^4)$, $\xi_j^2 \geq P^{\dot{\eta}}(\dot{\eta}_j^2)$, and $0 = P^{\dot{\eta}}(\dot{\eta}_j)$ for all $j \in \mathbb{N}$.* \square

§07102.09 **dieMM with noisy operator** (§02104.04 continued). For $\mathbb{J} = \mathbb{L}_2(\nu)$ let Assumption §07100.02 be satisfied where $\mathfrak{s}_\cdot \in \mathfrak{S} \subseteq \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ is *not known* anymore. We illustrate the tOPE in a Diagonal inverse empirical mean model (dieMM) with noisy operator as in §02104.04. Here the observable stochastic processes $\widehat{\mathfrak{s}}_\cdot = \mathfrak{s}_\cdot + k^{-1/2} \dot{\eta}_\cdot$ and $\widehat{g}_\cdot = g_\cdot + n^{-1/2} \dot{\varepsilon}_\cdot$ are noisy version of $\mathfrak{s}_\cdot \in \mathfrak{S}$ and $g_\cdot = \mathfrak{s}_\cdot \theta \in \mathbb{J}$ with $\theta \in \Theta \subseteq \mathbb{J}$, respectively, and *independent* error processes $\dot{\varepsilon}_\cdot = n^{1/2}(\widehat{P}_n(\psi_\cdot) - P_{\mathfrak{s}_\cdot}(\psi_\cdot)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes \mathcal{J})$ and $\dot{\eta}_\cdot = k^{1/2}(\widehat{P}_k(\varphi_\cdot) - P_{\mathfrak{s}_\cdot}(\varphi_\cdot)) \in \mathcal{M}(\mathcal{Z}^{\otimes k} \otimes \mathcal{J})$ satisfying Assumption §01101.04. More precisely, on a measurable space $(\mathcal{Z}, \mathcal{Z})$ for each $\theta \in \Theta$ and $\mathfrak{s}_\cdot \in \mathfrak{S}$ there are probability measures $P_{\theta|\mathfrak{s}}, P_{\mathfrak{s}} \in \mathcal{W}(\mathcal{Z})$. Similar to Model §02104.04 consider stochastic processes $\psi_\cdot, \varphi_\cdot \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$. In addition for all $\theta \in \Theta$ and $\mathfrak{s}_\cdot \in \mathfrak{S}$ the process $\psi_\cdot \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ satisfies (dieMM1)-(dieMM3) of Model §07101.08 for $v_{\theta|\mathfrak{s}|\psi} \in \mathbb{R}_{\geq 1}$ and the process $\varphi_\cdot \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ fulfils

(dieMMnO1) $\varphi_j \in \mathcal{L}_1(P_{\mathfrak{s}}) := \mathcal{L}_1(\mathcal{Z}, \mathcal{Z}, P_{\mathfrak{s}})$ ν -a.e. $j \in \mathcal{J}$ and $P_{\mathfrak{s}}(\varphi_\cdot) = \mathfrak{s}_\cdot$ ν -a.s.,

(dieMMnO2) there is $v_{\mathfrak{s}|\varphi}^2 \in \mathbb{R}_{\geq 1}$ such that $\|P_{\mathfrak{s}}(\varphi_\cdot^4)\|_{\mathbb{L}_\infty(\nu)} \leq v_{\mathfrak{s}|\varphi}^2$ and hence $\|P_{\mathfrak{s}}(\varphi_\cdot^2)\|_{\mathbb{L}_\infty(\nu)} \leq v_{\mathfrak{s}|\varphi}$.

We consider a statistical product experiment $(\mathcal{Z}^{n+k}, \mathcal{Z}^{\otimes(n+k)}, P_{\Theta \times \mathfrak{S}}^{n \otimes k} = (P_{\theta|\mathfrak{s}}^{\otimes n} \otimes P_{\mathfrak{s}}^{\otimes k})_{\theta \in \Theta, \mathfrak{s} \in \mathfrak{S}})$ as in an Empirical mean function §01101.10 where $\mathfrak{S} \subseteq \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ and $\Theta \subseteq \mathbb{J}$. \square

§07102.10 **Property** (dieMM with noisy operator §07102.09 continued). *Under (dieMMnO1) and (dieMMnO2) the process $\dot{\eta}_\cdot = k^{1/2}(\widehat{P}_k - P_{\mathfrak{s}})(\varphi_\cdot) \in \mathcal{M}(\mathcal{Z}^{\otimes k} \otimes \mathcal{J})$ satisfies $v_{\mathfrak{s}|\varphi}^2 \geq \|P_{\mathfrak{s}}^{\otimes k}(\dot{\eta}_\cdot^4)\|_{\mathbb{L}_\infty(\nu)}$, $v_{\mathfrak{s}|\varphi} \geq \|P_{\mathfrak{s}}^{\otimes k}(\dot{\eta}_\cdot^2)\|_{\mathbb{L}_\infty(\nu)}$, and $0 = P_{\mathfrak{s}}^{\otimes k}(\dot{\eta}_j)$ for ν -a.e. $j \in \mathcal{J}$.* \square

§07|02|02 Global and maximal global ν -risk

We measure first the accuracy of the tOPE $\widehat{\theta}_\cdot^m := \widehat{\mathfrak{s}}_\cdot^{(k)\dagger} \widehat{g}_\cdot^m$ of the projection $\theta_\cdot^m = \mathfrak{s}_\cdot^\dagger g_\cdot^m \in \mathbb{J} \mathbb{1}_\cdot^m$ with

$\underline{g} = \underline{s}\theta \in \text{dom}(\underline{M}_s)$ and $\underline{s} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ by the mean of its global \mathfrak{v} -error introduced in §04103101, i.e. its \mathfrak{v} -risk.

§07102.11 **Reminder.** If $\mathfrak{v} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\theta \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ then for each $m \in \mathbb{N}$ we have $\theta^m \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ too and $\|\theta^m - \theta\|_{\mathfrak{v}}^2 = o(1)$ as $m \rightarrow \infty$ (**Property** §04103.09). \square

§07102.12 **Assumption.** Let $(\widehat{\underline{g}}, \widehat{\underline{s}}) = (\underline{g} + n^{-1/2}\underline{\dot{\epsilon}}, \underline{s} + k^{-1/2}\underline{\dot{\eta}}) \sim \mathbb{P}_{\theta_s}^{n \otimes k} := \mathbb{P}_{\theta_s}^n \otimes \mathbb{P}_s^k$ be independent noisy versions satisfying Assumption §07102.01. In addition

(dSIPg1) $\underline{\mathfrak{v}}^{\theta_s} := \mathbb{P}_{\theta_s}^n(\underline{\dot{\epsilon}}^2) := (\underline{\mathfrak{v}}_j^{\theta_s} := \mathbb{P}_s^n(\dot{\epsilon}_j^2))_{j \in \mathcal{J}} \in \mathbb{L}_\infty(\nu)$, $\mathbb{K}_{\theta_s}^2 := 1 \vee \|\underline{\mathfrak{v}}^{\theta_s}\|_{\mathbb{L}_\infty(\nu)}$,

(dSIPg2) $\underline{\dot{\epsilon}} \cdot \mathbb{1}^m \in \mathbb{L}_\infty(\nu)$ $\mathbb{P}_{\theta_s}^n$ -a.s. for each $m \in \mathbb{N}$, and

(dSIPnO) $\underline{\mathfrak{v}}^{s(2)} := \mathbb{P}_s^k(\underline{\dot{\eta}}^4) := (\underline{\mathfrak{v}}_j^{s(2)} := \mathbb{P}_s^k(\dot{\eta}_j^4))_{j \in \mathcal{J}} \in \mathbb{L}_\infty(\nu)$, $\mathbb{K}_s^4 := 1 \vee \|\underline{\mathfrak{v}}^{s(2)}\|_{\mathbb{L}_\infty(\nu)}$.

Moreover, from (dSIPnO) (i.e. $\underline{\mathfrak{v}}^{s(2)} = \mathbb{P}_s^k(\underline{\dot{\eta}}^4) \in \mathbb{L}_\infty(\nu)$) follows $\mathbb{P}_s^k(\underline{\dot{\eta}}_j^2) =: \underline{\mathfrak{v}}_j^s \leq (\underline{\mathfrak{v}}_j^{s(2)})^{1/2}$ for ν -a.e. $j \in \mathcal{J}$, and hence $\|\underline{\mathfrak{v}}^s \vee \mathbb{1}\|_{\mathbb{L}_\infty(\nu)} \leq \mathbb{K}_s^2$. \square

§07102.13 **Notation.** Since $\|\widehat{\underline{s}}^{(k)\dagger}\|_{\mathbb{L}_\infty(\nu)} \leq k^{1/2}$ (**Notation** §07102.03), $\underline{s} \in \mathbb{L}_\infty(\nu)$ and $\mathbb{1}^m \in \mathbb{L}_\infty(\nu)$ for all $m \in \mathbb{N}$, for $(\widehat{\underline{s}}^{(k)\dagger}\underline{s}) := \widehat{\underline{s}}^{(k)\dagger}\underline{s} \in \mathcal{M}(\mathcal{A} \otimes \mathcal{J})$ we have $(\widehat{\underline{s}}^{(k)\dagger}\underline{s}) \cdot \mathbb{1}^m \in \mathbb{L}_\infty(\nu)$ for all $m \in \mathbb{N}$ too. If in addition $\mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ for all $m \in \mathbb{N}$ then for $(\widehat{\underline{s}}^{(k)\dagger}\mathfrak{v}) := \widehat{\underline{s}}^{(k)\dagger}\mathfrak{v} \in \mathcal{M}(\mathcal{A} \otimes \mathcal{J})$ we also have $(\widehat{\underline{s}}^{(k)\dagger}\mathfrak{v}) \cdot \mathbb{1}^m \in \mathbb{J}$ for all $m \in \mathbb{N}$. \square

§07102.14 **Comment.** Under Assumption §07102.12 and $\mathfrak{v} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ if $\mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ for all $m \in \mathbb{N}$ then we have $(\widehat{\underline{s}}^{(k)\dagger}\mathfrak{v}) \cdot \underline{\dot{\epsilon}} \cdot \mathbb{1}^m \in \mathbb{J}$ $\mathbb{P}_{\theta_s}^{n \otimes k}$ -a.s.. If in addition $\theta \in \mathbb{L}_2(\mathfrak{v}^2\nu)$, and hence $\theta^m \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ (**Property** §04103.09), then it follows

$$\mathfrak{v} \cdot \widehat{\theta}^m = (\widehat{\underline{s}}^{(k)\dagger}\mathfrak{v}) \cdot \widehat{\underline{g}} \cdot \mathbb{1}^m = n^{-1/2}(\widehat{\underline{s}}^{(k)\dagger}\mathfrak{v}) \cdot \underline{\dot{\epsilon}} \cdot \mathbb{1}^m + (\widehat{\underline{s}}^{(k)\dagger}\underline{s}) \cdot \mathfrak{v} \cdot \theta^m \in \mathbb{J} \quad \mathbb{P}_{\theta_s}^{n \otimes k}\text{-a.s..} \quad (07.27)$$

If $\mathcal{J} \subseteq \mathbb{Z}$ (at most countable) and ν_j is the counting measure over the index set \mathcal{J} then Assumption §01101.04 and (dSIPg1) (i.e. $\underline{\mathfrak{v}}^{\theta_s} = \mathbb{P}_{\theta_s}^n(\underline{\dot{\epsilon}}^2) \in \mathbb{L}_\infty(\nu_j)$) imply the additional assumption (dSIPg2) $\underline{\dot{\epsilon}} \cdot \mathbb{1}^m \in \mathbb{L}_\infty(\nu_j)$ $\mathbb{P}_{\theta_s}^n$ -a.s.. However, the last implication does generally not hold, if $\mathcal{J} \in \{\mathbb{R}, \mathbb{R}_{>0}\}$ for example. \square

§07102102101 Global \mathfrak{v} -risk

§07102.15 **Assumption.** Let $\mathfrak{v} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$, $\theta \in \mathbb{L}_2(\mathfrak{v}^2\nu)$, and $\widehat{\underline{s}}^\dagger \mathbb{1}^m, \mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ for $m \in \mathbb{N}$ be satisfied. \square

§07102.16 **Definition.** Under Assumptions §07102.12 and §07102.15 for $m \in \mathbb{N}$ the *global \mathfrak{v} -risk* of a thresholded OPE $\widehat{\theta}^m = \widehat{\underline{s}}^{(k)\dagger}\widehat{\underline{g}}^m = \widehat{\underline{s}}^\dagger \mathbb{1}^{\{\widehat{\epsilon}^2 > k^{-1}\}} \widehat{\underline{g}} \cdot \mathbb{1}^m \in \mathbb{L}_2(\mathfrak{v}^2\nu)$ $\mathbb{P}_{\theta_s}^{n \otimes k}$ -a.s. satisfies

$$\mathbb{P}_{\theta_s}^{n \otimes k}(\|\widehat{\theta}^m - \theta\|_{\mathfrak{v}}^2) = \mathbb{P}_{\theta_s}^{n \otimes k}(\|\widehat{\underline{s}}^{(k)\dagger}(\widehat{\underline{g}} - \widehat{\underline{s}}\theta) \cdot \mathbb{1}^m\|_{\mathfrak{v}}^2) + \mathbb{P}_s^k(\|\mathbb{1}^{\{\widehat{\epsilon}^2 < k^{-1}\}} \theta \cdot \mathbb{1}^m\|_{\mathfrak{v}}^2) + \|\theta \cdot \mathbb{1}^{m \perp}\|_{\mathfrak{v}}^2 \quad (07.28)$$

with *variance* terms $\mathbb{P}_{\theta_s}^{n \otimes k}(\|\widehat{\underline{s}}^{(k)\dagger}(\widehat{\underline{g}} - \widehat{\underline{s}}\theta) \cdot \mathbb{1}^m\|_{\mathfrak{v}}^2)$, $\mathbb{P}_s^k(\|\mathbb{1}^{\{\widehat{\epsilon}^2 < k^{-1}\}} \theta \cdot \mathbb{1}^m\|_{\mathfrak{v}}^2)$ and *bias* term $\|\theta \cdot \mathbb{1}^{m \perp}\|_{\mathfrak{v}}^2$.

§07102.17 **Property.** Under Assumptions §07102.12 and §07102.15 for each $m \in \mathbb{N}$ we have

$$\begin{aligned} \mathbb{P}_{\theta_s}^{n \otimes k} \|\widehat{\underline{s}}^{(k)\dagger}(\widehat{\underline{g}} - \widehat{\underline{s}}\theta) \cdot \mathbb{1}^m\|_{\mathfrak{v}}^2 &= \mathbb{P}_{\theta_s}^n \otimes \mathbb{P}_s^k \|\widehat{\underline{s}}^{(k)\dagger}(n^{-1/2}\underline{\dot{\epsilon}} + (\underline{s} - \widehat{\underline{s}})\theta) \cdot \mathbb{1}^m\|_{\mathfrak{v}}^2 \\ &= n^{-1}\nu \left(\mathbb{P}_s^k((\widehat{\underline{s}}^{(k)\dagger}\underline{s})^2) \underline{\mathfrak{v}}^{\theta_s} (\underline{s}^\dagger \mathfrak{v})^2 \cdot \mathbb{1}^m \right) + \nu \left(\mathbb{P}_s^k(|\widehat{\underline{s}}^{(k)\dagger}|^2 |\underline{s} - \widehat{\underline{s}}|^2) \mathfrak{v}^2 \theta^2 \cdot \mathbb{1}^m \right) \end{aligned}$$

($\underline{s} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ by Assumption §07102.01) and $\mathbb{P}_s^k \|\mathbb{1}^{\{\widehat{\epsilon}^2 < k^{-1}\}} \theta \cdot \mathbb{1}^m\|_{\mathfrak{v}}^2 = \nu \left(\mathbb{P}_s^k(\widehat{\epsilon}^2 < k^{-1}) \mathfrak{v}^2 \theta^2 \cdot \mathbb{1}^m \right)$. \square

§07102.18 **Lemma.** Under Assumption §07102.12 (dSIPnO) for all $j \in \mathcal{J}$ we have

$$(i) \quad \mathbb{P}_s^k((\widehat{\underline{s}}^{(k)\dagger}\underline{s})_j^2) \leq 2(\underline{\mathfrak{v}}_j^s + 1) \leq 4(1 \vee \underline{\mathfrak{v}}_j^{s(2)})^{1/2},$$

- (ii) $\mathbb{P}_s^k(\widehat{\mathfrak{s}}_j^2 < k^{-1}) \leq 4(1 \vee \mathfrak{v}_j^s)(1 \vee k\mathfrak{s}_j^2)^{-1} \leq 4(1 \vee \mathfrak{v}_j^{s(2)})^{1/2}(1 \vee k\mathfrak{s}_j^2)^{-1}$, and
 (iii) $\mathbb{P}_s^k(|\mathfrak{s}_j - \widehat{\mathfrak{s}}_j|^2 |\widehat{\mathfrak{s}}_j^{(k)\dagger}|^2) \leq 2(\mathfrak{v}_j^{s(2)} + \mathfrak{v}_j^s)(1 \vee k\mathfrak{s}_j^2)^{-1} \leq 4(1 \vee \mathfrak{v}_j^{s(2)})(1 \vee k\mathfrak{s}_j^2)^{-1}$.

§07102.19 **Proof** of Lemma §07102.18. Given in the lecture. \square

§07102.20 **Reminder**. If Assumptions §07101.11 and §07101.13 are satisfied, then for all $n, m \in \mathbb{N}$ setting

$$\begin{aligned} R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) &:= \|\theta \mathbb{1}^{m \perp}\|_{\mathfrak{v}}^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbb{1}^m\|_{\mathfrak{v}}^2, \quad m_n^\circ := \arg \min \{R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) : m \in \mathbb{N}\} \\ \text{and } R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) &:= R_n^{m_n^\circ}(\theta, \mathfrak{s}, \mathfrak{v}) = \min \{R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) : m \in \mathbb{N}\} \end{aligned} \quad (07.29)$$

the OPE $\widehat{\theta}^{m_n^\circ} := \mathfrak{s}^\dagger \widehat{g}^{m_n^\circ}$ with known $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ fulfils $\mathbb{E}_{\theta|\mathfrak{s}}^{n \otimes k}(\|\widehat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2) \leq (1 \vee \|\mathfrak{v}^{\theta|\mathfrak{s}}\|_{\mathbb{L}_\infty(\nu)}) R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$ due to Proposition §07101.17. Keep in mind that Assumption §07101.11 is part of Assumption §07102.12 and that Assumption §07101.13 is part of Assumption §07102.15. \square

§07102.21 **Proposition (Upper bound)**. Let Assumptions §07102.12 and §07102.15 be satisfied. The thresholded OPE $\widehat{\theta}^m = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^m \in \mathbb{L}_2(\mathfrak{v}^2 \nu)$ $\mathbb{E}_{\theta|\mathfrak{s}}^{n \otimes k}$ -a.s. for all $n, k, m \in \mathbb{N}$ fulfils

$$\begin{aligned} \mathbb{E}_{\theta|\mathfrak{s}}^{n \otimes k}(\|\widehat{\theta}^m - \theta\|_{\mathfrak{v}}^2) &\leq 4\|\mathfrak{v}^s \vee \mathbb{1}\|_{\mathbb{L}_\infty(\nu)} \|\mathfrak{v}^{\theta|\mathfrak{s}} \vee \mathbb{1}\|_{\mathbb{L}_\infty(\nu)} R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) \\ &\quad + 2(\|\mathfrak{v}^{s(2)}\|_{\mathbb{L}_\infty(\nu)} + 3\|\mathfrak{v}^s \vee \mathbb{1}\|_{\mathbb{L}_\infty(\nu)}) \|(1 \vee k\mathfrak{s}^2)^{-1/2} \theta \mathbb{1}^m\|_{\mathfrak{v}}^2 \end{aligned} \quad (07.30)$$

$$\leq 4K_s^2 K_{\theta|\mathfrak{s}}^2 R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) + 8K_s^4 \|(1 \vee k\mathfrak{s}^2)^{-1/2} \theta \mathbb{1}^m\|_{\mathfrak{v}}^2. \quad (07.31)$$

§07102.22 **Proof** of Proposition §07102.21. Given in the lecture. \square

§07102.23 **Comment**. For each $m \in \mathbb{N}$ we have

$$\begin{aligned} \|(1 \vee k\mathfrak{s}^2)^{-1/2} \theta \mathbb{1}^m\|_{\mathfrak{v}}^2 &\leq \|(1 \vee k\mathfrak{s}^2)^{-1/2} \theta\|_{\mathfrak{v}}^2 = \|(1 \vee k\mathfrak{s}^2)^{-1/2} \theta \mathbb{1}^m\|_{\mathfrak{v}}^2 + \|(1 \vee k\mathfrak{s}^2)^{-1/2} \theta \mathbb{1}^{m \perp}\|_{\mathfrak{v}}^2 \\ &\leq \|(1 \vee k\mathfrak{s}^2)^{-1/2} \theta \mathbb{1}^m\|_{\mathfrak{v}}^2 + \|\theta \mathbb{1}^{m \perp}\|_{\mathfrak{v}}^2. \end{aligned} \quad (07.32)$$

Consequently, under the assumptions of Proposition §07102.21 from (??) (Proof §07102.22) follows

$$\begin{aligned} \mathbb{E}_{\theta|\mathfrak{s}}^{n \otimes k}(\|\widehat{\theta}^m - \theta\|_{\mathfrak{v}}^2) &\leq 2K_s^2 K_{\theta|\mathfrak{s}}^2 R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) + 8K_s^4 \|\theta(1 \vee k\mathfrak{s}^2)^{-1/2} \mathbb{1}^m\|_{\mathfrak{v}}^2 \\ &\leq 2K_s^2 K_{\theta|\mathfrak{s}}^2 R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) + 8K_s^4 \|\theta(1 \vee k\mathfrak{s}^2)^{-1/2}\|_{\mathfrak{v}}^2 \\ &\leq 10K_s^4 K_{\theta|\mathfrak{s}}^2 R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) + 8K_s^4 \|\theta(1 \vee k\mathfrak{s}^2)^{-1/2} \mathbb{1}^m\|_{\mathfrak{v}}^2. \end{aligned}$$

Selecting $m_n^\circ := \arg \min \{R_n^m(\theta, \mathfrak{s}, \mathfrak{v}) : m \in \mathbb{N}\}$ and $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) = R_n^{m_n^\circ}(\theta, \mathfrak{s}, \mathfrak{v})$ as in (07.29) (Reminder §07102.20) we obtain

$$\mathbb{E}_{\theta|\mathfrak{s}}^{n \otimes k}(\|\widehat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2) \leq 2K_s^2 K_{\theta|\mathfrak{s}}^2 R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) + 8K_s^4 \|\theta(1 \vee k\mathfrak{s}^2)^{-1/2}\|_{\mathfrak{v}}^2. \quad (07.33)$$

We shall emphasise, that the upper bound consists (up to the constants) of the sum of the two terms $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$ and $\|\theta(1 \vee k\mathfrak{s}^2)^{-1/2}\|_{\mathfrak{v}}^2$ depending each on one of the sample sizes n and k only. Moreover, $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$ is the oracle rate (Property §07101.20) in case of an in advanced known $M_s \in \mathbb{L}^{\mathfrak{M}(\mathfrak{J})}$. \square

§07102.24 **Corollary** (GdiSM with noisy operator §07102.05 continued). Consider independent noisy versions $(\widehat{g}, \widehat{\mathfrak{s}}) = (g + n^{-1/2} \dot{B}, \mathfrak{s} + k^{-1/2} \dot{W}) \sim N_{\theta|\mathfrak{s}}^{n \otimes k} = N_{\theta|\mathfrak{s}}^n \otimes N_s^k$ as in Model §07102.05, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{W} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ are independent, $\mathfrak{s} \in \mathbb{R}_{\nu_0}^{\mathbb{N}} \cap \ell_\infty$ and $\theta \in \ell_2$, and hence $g = \mathfrak{s} \cdot \theta \in \text{dom}(M_s) \subseteq \ell_2$. Given $\mathfrak{v}_s \in \mathbb{R}_{\nu_0}^{\mathbb{N}}$ and $\theta \in \ell_2(\mathfrak{v}^2)$ the (infeasible) thresholded OPE $\widehat{\theta}^{m_n^\circ} = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^{m_n^\circ} \in \ell_2(\mathfrak{v}^2)$ with oracle dimension m_n° as in (07.29) satisfies

$$N_{\theta|\mathfrak{s}}^{n \otimes k}(\|\widehat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2) \leq 4R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v}) + 12\|(1 \vee k\mathfrak{s}^2)^{-1/2} \theta\|_{\mathfrak{v}}^2 \quad \forall n, k \in \mathbb{N} \quad (07.34)$$

where $R_n^\circ(\theta, \mathfrak{s}, \mathfrak{v})$ is the oracle rate in a GdiSM §07101.03 (see Corollary §07101.22).

§07102.25 **Proof of Corollary §07102.24.** Given in the lecture. \square

§07102.26 **Corollary** (diSM with noisy operator §07102.07 continued). *Consider independent noisy versions $(\widehat{g}_n, \widehat{\mathfrak{s}}_n) = (g_n + n^{-1/2}\widehat{\epsilon}_n, \mathfrak{s}_n + k^{-1/2}\widehat{\eta}_n) \sim \mathbb{P}_{\theta|\mathfrak{s}|\xi}^{n \otimes k} = \mathbb{P}_{\theta|\mathfrak{s}}^n \otimes \mathbb{P}_{\mathfrak{s}|\xi}^k$ as in Model §07102.07, where $\widehat{\epsilon}_n$ and $\widehat{\eta}_n$ satisfy (iSM1) and (diSMnO1) with $K_\sigma := \|\sigma_n\|_{\ell_\infty} \vee 1 \in \mathbb{R}_{>1}$ and $K_\xi := \|\xi_n\|_{\ell_\infty} \vee 1 \in \mathbb{R}_{>1}$, respectively, $\mathfrak{s}_n \in \mathbb{R}_{\neq 0}^N \cap \ell_\infty$ and $\theta_n \in \ell_2$, and hence $g_n = \mathfrak{s}_n \theta_n \in \text{dom}(M_n) \subseteq \ell_2$. Given $\mathfrak{v}_n \in \mathbb{R}_{\neq 0}^N$ and $\theta_n \in \ell_2(\mathfrak{v}_n^2)$ the (infeasible) thresholded OPE $\widehat{\theta}_n^{m_n^\circ} = \widehat{\mathfrak{s}}_n^{(k)\dagger} \widehat{g}_n^{m_n^\circ} \in \ell_2(\mathfrak{v}_n^2)$ with oracle dimension m_n° as in (07.29) satisfies*

$$\mathbb{P}_{\theta|\mathfrak{s}|\xi}^{n \otimes k} (\|\widehat{\theta}_n^{m_n^\circ} - \theta_n\|_{\mathfrak{v}_n}^2) \leq 4K_\sigma^2 K_\xi^2 R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n) + 8K_\xi^4 \|(1 \vee k \mathfrak{s}_n^2)^{-1/2} \theta_n\|_{\mathfrak{v}_n}^2 \quad \forall n, k \in \mathbb{N} \quad (07.35)$$

where $R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n)$ is the oracle rate in a diSM §07101.06 (see **Corollary §07101.24**).

§07102.27 **Proof of Corollary §07102.26.** Given in the lecture. \square

§07102.28 **Corollary** (dieMM with noisy operator §07102.09 continued). *Consider independent noisy versions $\widehat{g}_n = g_n + n^{-1/2}\widehat{\epsilon}_n$ and $\widehat{\mathfrak{s}}_n = \mathfrak{s}_n + k^{-1/2}\widehat{\eta}_n$ defined on $(\mathcal{Z}^{n+k}, \mathcal{Z}^{\otimes(n+k)}, \mathbb{P}_{\theta|\mathfrak{s}}^{n \otimes k} = \mathbb{P}_{\theta|\mathfrak{s}}^{\otimes n} \otimes \mathbb{P}_{\mathfrak{s}}^{\otimes k})$ as in Model §07102.09, where $\psi, \varphi \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ satisfy (dieMM1)–(dieMM3) (Model §07101.08) and (dieMMnO1)–(dieMMnO2) (Model §07102.09) with $\mathfrak{v}_{\theta|\mathfrak{s}|\psi}, \mathfrak{v}_{\mathfrak{s}|\varphi} \in \mathbb{R}_{\geq 1}$, respectively, $\mathfrak{s}_n \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$, $\theta_n \in \mathbb{J}$ and hence $g_n = \mathfrak{s}_n \theta_n \in \text{dom}(M_n) \subseteq \mathbb{J}$. Given Assumption §07102.15 the (infeasible) thresholded OPE $\widehat{\theta}_n^{m_n^\circ} = \widehat{\mathfrak{s}}_n^{(k)\dagger} \widehat{g}_n^{m_n^\circ} \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ with oracle dimension m_n° as in (07.29) satisfies*

$$\mathbb{P}_{\theta|\mathfrak{s}}^{n \otimes k} (\|\widehat{\theta}_n^{m_n^\circ} - \theta_n\|_{\mathfrak{v}_n}^2) \leq 4\mathfrak{v}_{\theta|\mathfrak{s}|\psi} \mathfrak{v}_{\mathfrak{s}|\varphi} R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n) + 8\mathfrak{v}_{\mathfrak{s}|\varphi}^2 \|(1 \vee k \mathfrak{s}_n^2)^{-1/2} \theta_n\|_{\mathfrak{v}_n}^2 \quad \forall n, k \in \mathbb{N} \quad (07.36)$$

where $R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n)$ is the oracle rate in a dieMM §07101.08 (see **Corollary §07101.26**).

§07102.29 **Proof of Corollary §07102.28.** Given in the lecture. \square

§07102.30 **Illustration.** We illustrate the last results considering usual behaviour for $\theta_n, \mathfrak{s}_n, \mathfrak{v}_n \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$. We distinguish again the two cases **(p)** and **(np)** in **Illustration §07101.28**, where in case **(p)** the term $R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n)$ is parametric, that is, $nR_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n) = O(1)$, in case **(np)** it is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n) = \infty$. In case **(np)** we consider again the three specifications **(o-m)**, **(o-s)** and **(s-m)** introduced in **Illustration §07101.28** where also in Table 01 [§07] the order of the oracle dimension m_n° and the oracle rate $R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n)$ as $n \rightarrow \infty$ are given. The next table depicts the oracle rate $R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n)$ and the rate of the additional term $\|(1 \vee k \mathfrak{s}_n^2)^{-1/2} \theta_n\|_{\mathfrak{v}_n}^2$ as $n, k \rightarrow \infty$:

Table 05 [§07]

Order of $R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n)$ and $\|(1 \vee k \mathfrak{s}_n^2)^{-1/2} \theta_n\|_{\mathfrak{v}_n}^2$ as $n, k \rightarrow \infty$

$(j \in \mathcal{J})$	$(a \in \mathbb{R}_{>0})$	$(t \in \mathbb{R}_{>0})$					
$\mathfrak{v}_j^2 = j^{2v}$	θ_j^2	\mathfrak{s}_j^2	$R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n)$	$\theta_j^2 \mathfrak{v}_j^2$	$\mathfrak{s}_j^{-2} \theta_j^2 \mathfrak{v}_j^2$	$\ (1 \vee k \mathfrak{s}_n^2)^{-1/2} \theta_n\ _{\mathfrak{v}_n}^2$	
(o-m)	$v \in (-1/2 - t, a)$	j^{-2a-1}	j^{-2t}	$n^{-\frac{2(a-v)}{2a+2t+1}}$	$j^{-2(a-v)-1}$	$j^{2(t+v-a)-1}$	$k^{-\frac{a-v}{t}}$
	$a - v < t$						$(k / \log k)^{-1}$
	$a - v = t$						k^{-1}
	$a - v > t$						k^{-1}
(o-s)	$a - v \in \mathbb{R}_{>0}$	j^{-2a-1}	$e^{-j^{2t}}$	$(\log n)^{-\frac{a-v}{t}}$	$j^{-2(a-v)-1}$	$j^{-2(a-v)-1} e^{j^{2t}}$	$(\log k)^{-\frac{a-v}{t}}$
(s-m)	$v + t + 1/2 \in \mathbb{R}_{>0}$	$e^{-j^{2a}}$	j^{-2t}	$n^{-1} (\log n)^{\frac{2(t+v)+1}{2a}}$	$j^{2v} e^{-j^{2a}}$	$j^{2(t+v)} e^{-j^{2a}}$	k^{-1}

We note that in case **(o-m)** and **(s-m)** for $v + t < -1/2$ the oracle rate $R_n^\circ(\theta_n, \mathfrak{s}_n, \mathfrak{v}_n)$ is parametric. \square

§07|02|02|02 Maximal global \mathfrak{v} -risk

§07|02.31 **Notation (Reminder).** For sequences $a_n, b_n \in (\mathbb{K})^{\mathbb{N}}$ taking its values in $\mathbb{K} \in \{\mathbb{R}, \mathbb{R}_{>0}, \mathbb{Q}, \mathbb{Z}, \dots\}$ we write $\underline{a}_n \in (\mathbb{K})_{\nearrow}^{\mathbb{N}}$ and $\underline{b}_n \in (\mathbb{K})_{\searrow}^{\mathbb{N}}$ if a_n and b_n , respectively, is monotonically *non-decreasing* and *non-increasing*. If in addition $a_n \rightarrow \infty$ and $b_n \rightarrow 0$ as $n \rightarrow \infty$, then we write $\underline{a}_n \in (\mathbb{K})_{\nearrow\infty}^{\mathbb{N}}$ and $\underline{b}_n \in (\mathbb{K})_{\searrow 0}^{\mathbb{N}}$ for short. For $\mathfrak{w}_n \in \mathbb{L}_{\infty}(\nu)$ we set $\mathfrak{w}_{(0)} := \|\mathfrak{w}_n\|_{\mathbb{L}_{\infty}(\nu)}$ and $\mathfrak{w}_{(\bullet)} = (\mathfrak{w}_{(j)} := \|\mathfrak{w}_n \cdot \mathbb{1}_j^{\perp}\|_{\mathbb{L}_{\infty}(\nu)})_{j \in \mathbb{N}}$, where by construction $\mathfrak{w}_{(\bullet)} \in (\mathbb{R}_{\geq 0})_{\searrow}^{\mathbb{N}}$. \square

§07|02.32 **Assumption.** Consider weights $\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n \in \mathcal{M}_{>0, \nu}(\mathcal{J})$ (i.e. $\nu(\mathcal{N}_{\mathfrak{a}}) = \nu(\mathcal{N}_{\mathfrak{t}}) = 0 = \nu(\mathcal{N}_{\mathfrak{v}})$), such that $\mathfrak{a}_n, \mathfrak{t}_n \in \mathbb{L}_{\infty}(\nu)$, $(\mathfrak{a}\mathfrak{v})_n = \mathfrak{a}_n \mathfrak{v}_n \in \mathbb{L}_{\infty}(\nu)$, $(\mathfrak{a}\mathfrak{v})_{(j)} \in (\mathbb{R}_{\geq 0})_{\searrow 0}^{\mathbb{N}}$, and $\mathfrak{t}_n^{\dagger} \mathbb{1}_n^m, \mathbb{1}_n^m \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ for all $m \in \mathbb{N}$. \square

§07|02.33 **Reminder.** Under Assumption §07|02.32 we have $\mathbb{J}^{\mathfrak{a}} = \mathbb{L}_2(\nu) = \text{dom}(\mathbb{M}_{\mathfrak{a}}) = \mathbb{J}\mathfrak{a}_n \subseteq \mathbb{J}$ and the three measures ν , $\mathfrak{a}_n^2 \nu$ and $\mathfrak{v}_n^2 \nu$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04|01.02). We consider $\mathbb{J}^{\mathfrak{a}}$ endowed with $\|\cdot\|_{\mathfrak{a}_n^{\dagger}} = \|\mathbb{M}_{\mathfrak{a}_n^{\dagger}} \cdot\|_{\mathbb{J}}$ and given a constant $r \in \mathbb{R}_{>0}$ the ellipsoid $\mathbb{J}^{\mathfrak{a}, r} := \{h_n \in \mathbb{J}^{\mathfrak{a}} : \|h_n\|_{\mathfrak{a}_n^{\dagger}} \leq r\} \subseteq \mathbb{J}^{\mathfrak{a}}$. Since $(\mathfrak{a}\mathfrak{v})_n \in \mathbb{L}_{\infty}(\nu)$, and hence $(\mathfrak{a}\mathfrak{v})_{(m)} := \|(\mathfrak{a}\mathfrak{v})_n \cdot \mathbb{1}_n^{m\perp}\|_{\mathbb{L}_{\infty}(\nu)} \in \mathbb{R}_{\geq 0}$ for each $m \in \mathbb{N}$ we have $\mathbb{J}^{\mathfrak{a}} \subseteq \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ (**Property** §04|02.11), and $\|\theta_n \cdot \mathbb{1}_n^{m\perp}\|_{\mathfrak{v}_n} \leq r (\mathfrak{a}\mathfrak{v})_{(m)}$ for all $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$ (**Lemma** §04|02.13). Let in addition $\mathbb{M}_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}$ satisfy a link condition as in **Definition** §04|03.05 with weights $\mathfrak{t}_n \in \mathcal{M}_{>0, \nu}(\mathcal{J}) \cap \mathbb{L}_{\infty}(\nu)$, and radius $\mathfrak{d} \in \mathbb{R}_{>0}$. We set $(\mathfrak{t}\mathfrak{v})_n := (\mathfrak{t}_n^{\dagger} \mathfrak{v}_n)_{j \in \mathcal{J}} = \mathfrak{t}_n^{\dagger} \mathfrak{v}_n \in \mathcal{M}(\mathcal{J})$. Obviously, for each $m \in \mathbb{N}$ the condition $\mathfrak{t}_n^{\dagger} \mathbb{1}_n^m \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ (due to Assumption §07|02.32) implies $\mathfrak{s}_n^{\dagger} \mathbb{1}_n^m \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ too. Consequently, if Assumption §07|02.32, $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$ and $\mathbb{M}_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}$ are satisfied, then Assumption §07|02.15 is also fulfilled. Keep in mind if Assumptions §07|01.11 and §07|01.30 are satisfied, then for $n, m \in \mathbb{N}$ setting

$$\begin{aligned} R_n^m(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) &:= [(\mathfrak{a}\mathfrak{v})_{(m)}^2 \vee n^{-1} \|\mathfrak{t}_n^{\dagger} \mathbb{1}_n^m\|_{\mathfrak{v}_n}^2], \quad m_n^* := \arg \min \{R_n^m(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) : m \in \mathbb{N}\} \\ &\text{and} \quad R_n^*(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) := R_n^{m_n^*}(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) = \min \{R_n^m(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) : m \in \mathbb{N}\} \end{aligned} \quad (07.37)$$

for each $\mathbb{M}_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}$ known in advance, for all $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$, and hence $g_n = \mathfrak{s}_n \theta_n \in \text{dom}(\mathbb{M}_{\mathfrak{s}}) \subseteq \mathbb{J}$, and for each $n \in \mathbb{N}$ the OPE $\widehat{\theta}_n^{m_n^*} := \mathfrak{s}_n^{\dagger} \widehat{g}_n^{m_n^*} \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ fulfils

$$\mathbb{E}_{\theta_n^{\mathfrak{s}}}^n (\|\widehat{\theta}_n^{m_n^*} - \theta_n\|_{\mathfrak{v}_n}^2) \leq (\|\mathfrak{v}_n^{\theta_n^{\mathfrak{s}}}\|_{\mathbb{L}_{\infty}(\nu)} \mathfrak{d}^2 + r^2) R_n^*(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) \quad \forall n \in \mathbb{N}.$$

due to **Corollary** §07|01.35. We shall emphasise that Assumptions §07|02.12 and §07|02.32 contains Assumptions §07|01.11 and §07|01.30, respectively. \square

§07|02.34 **Proposition (Upper bound).** Let Assumptions §07|02.12 and §07|02.32 be satisfied. If $\mathbb{M}_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}$ with $\mathfrak{d} \in \mathbb{R}_{>0}$, and $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$ with $r \in \mathbb{R}_{>0}$, then the thresholded OPE $\widehat{\theta}_n^m = \widehat{\mathfrak{s}}_n^{(k)\dagger} \widehat{g}_n^m \in \mathbb{L}_2(\mathfrak{v}_n^2 \nu)$ $\mathbb{E}_{\theta_n^{\mathfrak{s}}}^{n \otimes k}$ -a.s. for all $n, k, m \in \mathbb{N}$ fulfils

$$\begin{aligned} \mathbb{E}_{\theta_n^{\mathfrak{s}}}^{n \otimes k} (\|\widehat{\theta}_n^m - \theta_n\|_{\mathfrak{v}_n}^2) &\leq (4\|\mathfrak{v}_n^{\mathfrak{s}} \vee \mathbb{1}_n\|_{\mathbb{L}_{\infty}(\nu)} \|\mathfrak{v}_n^{\theta_n^{\mathfrak{s}}}\|_{\mathbb{L}_{\infty}(\nu)} \mathfrak{d}^2 + r^2) R_n^m(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) \\ &\quad + 2(\|\mathfrak{v}_n^{\mathfrak{s}(2)}\|_{\mathbb{L}_{\infty}(\nu)} + 3\|\mathfrak{v}_n^{\mathfrak{s}} \vee \mathbb{1}_n\|_{\mathbb{L}_{\infty}(\nu)}) \mathfrak{d}^2 r^2 \|(1 \vee k \mathfrak{t}_n^{\dagger})^{-1} (\mathfrak{a}\mathfrak{v})_n^2\|_{\mathbb{L}_{\infty}(\nu)} \end{aligned} \quad (07.38)$$

$$\leq (4K_{\mathfrak{s}}^2 K_{\theta_n^{\mathfrak{s}}}^2 \mathfrak{d}^2 + r^2) R_n^m(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) + 8K_{\mathfrak{s}}^4 \mathfrak{d}^2 r^2 \|(1 \vee k \mathfrak{t}_n^{\dagger})^{-1} (\mathfrak{a}\mathfrak{v})_n^2\|_{\mathbb{L}_{\infty}(\nu)}. \quad (07.39)$$

§07|02.35 **Proof of Proposition** §07|02.34. Given in the lecture. \square

§07|02.36 **Remark.** If in addition there exists $\mathfrak{v} \in \mathbb{R}_{>0}$ satisfying $\mathfrak{v} \geq (K_{\mathfrak{s}} \vee K_{\theta_n^{\mathfrak{s}}})$ for all $\theta_n \in \mathbb{J}^{\mathfrak{a}, r}$ and $\mathbb{M}_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}$ then the maximal global \mathfrak{v} -risk of the thresholded OPE $\widehat{\theta}_n^{m_n^*}$ with optimally chosen dimension $m_n^* := \arg \min \{R_n^m(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) : m \in \mathbb{N}\}$ and $R_n^*(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) = \min \{R_n^m(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) : m \in \mathbb{N}\}$ as in (07.37) fulfils

$$\begin{aligned} &\sup \{ \mathbb{E}_{\theta_n^{\mathfrak{s}}}^{n \otimes k} (\|\widehat{\theta}_n^{m_n^*} - \theta_n\|_{\mathfrak{v}_n}^2) : \theta_n \in \mathbb{J}^{\mathfrak{a}, r}, \mathbb{M}_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, \mathfrak{d}} \} \\ &\leq (4\mathfrak{v}^4 \mathfrak{d}^2 + r^2 + 8\mathfrak{v}^4 r^2 \mathfrak{d}^2) R_n^*(\mathfrak{a}_n, \mathfrak{t}_n, \mathfrak{v}_n) \vee \|(\mathfrak{a}\mathfrak{v})_n^2 (1 \vee k \mathfrak{t}_n^{\dagger})^{-1}\|_{\mathbb{L}_{\infty}(\nu)} \text{ for all } n, k \in \mathbb{N}. \end{aligned}$$

Arguing similarly as in **Remark** §07101.21 we note that $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) = o(1)$ as $n \rightarrow \infty$ since $(\dagger \mathbf{v}) \cdot \mathbb{1}^m \in \mathbb{L}_\infty(\nu)$ for all $m \in \mathbb{N}$ and $(\mathbf{a}\mathbf{v})_{(m)} = o(1)$ as $m \rightarrow \infty$ by Assumption §07102.32. Moreover, we have $\|(\mathbf{a}\mathbf{v}) \cdot (1 \vee k \mathbf{t}^\dagger)^{-1}\|_{\mathbb{L}_\infty(\nu)} = o(1)$ as $m \rightarrow \infty$ by dominated convergence. Note that the dimension $m_n^* := m_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ as defined in (07.37) does depend neither on the unknown parameter of interest θ nor on the unknown operator $M_{\mathfrak{s}}$ but on the classes $\mathbb{J}^{\mathbf{a}, \mathbf{r}}$ and $\mathbb{M}_{\mathbf{t}, \mathbf{d}}$ only, and thus also the statistic $\widehat{\theta}^{m_n^*}$. In other words, if the regularity of θ and $M_{\mathfrak{s}}$ is known in advance, then the thresholded OPE $\widehat{\theta}^{m_n^*}$ is a feasible estimator. \square

§07102.37 **Corollary** (GdiSM with noisy operator §07102.05 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{\mathfrak{s}}) = (g + n^{-1/2} \dot{B}, \mathfrak{s} + k^{-1/2} \dot{W}) \sim N_{\theta|\mathfrak{s}}^{n \otimes k} = N_{\theta|\mathfrak{s}}^n \otimes N_{\mathfrak{s}}^k$ as in Model §07102.05, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{W} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ are independent, $\mathfrak{s} \in \mathbb{R}_{\setminus 0} \cap \ell_\infty$, $\theta \in \ell_2$, and hence $g = \mathfrak{s} \cdot \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \ell_2$. Under Assumption §07102.32 the thresholded OPE $\widehat{\theta}^{m_n^*} = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^{m_n^*} \in \ell_2(\mathbf{v}^2)$ with dimension m_n^* as in (07.37) satisfies*

$$\begin{aligned} \sup \{ N_{\theta|\mathfrak{s}}^{n \otimes k} (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbf{v}}^2) : \theta \in \ell_2^{\mathbf{a}, \mathbf{r}}, M_{\mathfrak{s}} \in \mathbb{M}_{\mathbf{t}, \mathbf{d}} \} \\ \leq C_{\mathbf{r}, \mathbf{d}} R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \vee \|(\mathbf{a}\mathbf{v}) \cdot (1 \vee k \mathbf{t}^\dagger)^{-1}\|_{\ell_\infty} \end{aligned} \quad (07.40)$$

with constant $C_{\mathbf{r}, \mathbf{d}} = \mathbf{r}^2 + 4\mathbf{d}^2 + 12\mathbf{r}^2\mathbf{d}^2$.

§07102.38 **Proof of Corollary** §07102.37. Given in the lecture. \square

§07102.39 **Corollary** (diSM with noisy operator §07102.07 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{\mathfrak{s}}) = (g + n^{-1/2} \dot{\xi}, \mathfrak{s} + k^{-1/2} \dot{\eta}) \sim P_{\theta|\mathfrak{s}|\xi}^{n \otimes k} = P_{\theta|\mathfrak{s}}^n \otimes P_{\xi}^k$ as in Model §07102.07, where $\dot{\xi}$ and $\dot{\eta}$ satisfy (iSM1) and (diSMnO1) with $K_\sigma := \|\sigma\|_{\ell_\infty} \vee 1 \in \mathbb{R}_{\geq 1}$ and $K_\xi := \|\xi\|_{\ell_\infty} \vee 1 \in \mathbb{R}_{\geq 1}$, respectively, $\mathfrak{s} \in \mathbb{R}_{\setminus 0} \cap \ell_\infty$ and $\theta \in \ell_2$, and hence $g = \mathfrak{s} \cdot \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \ell_2$. Under Assumption §07102.32 the thresholded OPE $\widehat{\theta}^{m_n^*} = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^{m_n^*} \in \ell_2(\mathbf{v}^2)$ with dimension m_n^* as in (07.37) satisfies*

$$\begin{aligned} \sup \{ P_{\theta|\mathfrak{s}|\xi}^{n \otimes k} (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbf{v}}^2) : \theta \in \ell_2^{\mathbf{a}, \mathbf{r}}, M_{\mathfrak{s}} \in \mathbb{M}_{\mathbf{t}, \mathbf{d}} \} \\ \leq C_{\mathbf{r}, \mathbf{d}, \sigma, \xi} R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \vee \|(\mathbf{a}\mathbf{v}) \cdot (1 \vee k \mathbf{t}^\dagger)^{-1}\|_{\ell_\infty} \end{aligned} \quad (07.41)$$

with constant $C_{\mathbf{r}, \mathbf{d}, \sigma, \xi} = \mathbf{r}^2 + 4K_\xi^2 K_\sigma^2 \mathbf{d}^2 + 8K_\xi^4 \mathbf{r}^2 \mathbf{d}^2$.

§07102.40 **Proof of Corollary** §07102.39. Given in the lecture. \square

§07102.41 **Corollary** (dieMM with noisy operator §07102.09 continued). *Consider independent noisy versions $\widehat{g} = g + n^{-1/2} \dot{\xi}$ and $\widehat{\mathfrak{s}} = \mathfrak{s} + k^{-1/2} \dot{\eta}$ defined on $(\mathcal{Z}^{n+k}, \mathcal{Z}^{\otimes(n+k)}, P_{\theta|\mathfrak{s}}^{n \otimes k} = P_{\theta|\mathfrak{s}}^{\otimes n} \otimes P_{\mathfrak{s}}^{\otimes k})$ as in Model §07102.09, where $\psi, \varphi \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ satisfies (dieMM1)–(dieMM3) (Model §07101.08) and (dieMMnO1)–(dieMMnO2) (Model §07102.09) with $\mathbb{V}_{\theta|\mathfrak{s}|\psi}, \mathbb{V}_{\mathfrak{s}|\varphi} \in \mathbb{R}_{\geq 1}$, respectively, $\mathfrak{s} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$, $\theta \in \mathbb{J}$ and hence $g = \mathfrak{s} \cdot \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \mathbb{J}$. Under Assumption §07102.32 the thresholded OPE $\widehat{\theta}^{m_n^*} = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^{m_n^*} \in \mathbb{L}_2(\mathbf{v}^2 \nu)$ with dimension m_n^* as in (07.37) satisfies*

$$\begin{aligned} \sup \{ P_{\theta|\mathfrak{s}|\xi}^{n \otimes k} (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbf{v}}^2) : \theta \in \mathbb{J}^{\mathbf{a}, \mathbf{r}}, M_{\mathfrak{s}} \in \mathbb{M}_{\mathbf{t}, \mathbf{d}} \} \\ \leq C_{\mathbf{a}, \mathbf{r}, \mathbf{t}, \mathbf{d}} R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \vee \|(\mathbf{a}\mathbf{v}) \cdot (1 \vee k \mathbf{t}^\dagger)^{-1}\|_{\ell_\infty} \end{aligned} \quad (07.42)$$

with constant $C_{\mathbf{a}, \mathbf{r}, \mathbf{t}, \mathbf{d}} = \mathbf{r}^2 + 4\mathbf{d}^2 \sup \{ \mathbb{V}_{\theta|\mathfrak{s}|\psi} \mathbb{V}_{\mathfrak{s}|\varphi} : \theta \in \mathbb{J}^{\mathbf{a}, \mathbf{r}}, M_{\mathfrak{s}} \in \mathbb{M}_{\mathbf{t}, \mathbf{d}} \} + 8\mathbf{r}^2 \mathbf{d}^2 \sup \{ \mathbb{V}_{\mathfrak{s}|\varphi}^2 : M_{\mathfrak{s}} \in \mathbb{M}_{\mathbf{t}, \mathbf{d}} \}$.

§07102.42 **Proof of Corollary** §07102.41. Given in the lecture. \square

§07102.43 **Illustration**. We illustrate the last results considering usual behaviour for $\mathbf{a}, \mathbf{t}, \mathbf{v} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$. We distinguish again the two cases **(p)** and **(np)** in **Illustration** §07101.44, where in case **(p)** the

term $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ is parametric, that is, $nR_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) = O(1)$, in case **(np)** it is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) = \infty$. In case **(np)** we consider again three specifications similar to **(o-m)**, **(o-s)** and **(s-m)** introduced in **Illustration** §07101.44 where also in Table 02 [§07] the order of the dimension m_n^* and the rate $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ as $n \rightarrow \infty$ are given. The next table depicts the rate $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ and the additional term $\|(1 \vee k\mathbf{t}^2)^{-1}(\mathbf{a}\mathbf{v})^2\|_{\mathbb{L}_\infty(\nu)}$ as $n, k \rightarrow \infty$:

Table 06 [§07]

Order of $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ and $\|(1 \vee k\mathbf{t}^2)^{-1}(\mathbf{a}\mathbf{v})^2\|_{\mathbb{L}_\infty(\nu)}$ as $n, k \rightarrow \infty$

$(j \in \mathcal{J})$	$(\mathbf{a} \in \mathbb{R}_{>0}) (\mathbf{t} \in \mathbb{R}_{>0})$		$R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$	$(\mathbf{a}\mathbf{v})_j^2$	$\mathbf{t}_j^{-2}(\mathbf{a}\mathbf{v})_j^2$	$\ (1 \vee k\mathbf{t}^2)^{-1}(\mathbf{a}\mathbf{v})^2\ _{\mathbb{L}_\infty(\nu)}$
$\mathbf{v}_j^2 = j^{2\nu}$	\mathbf{a}_j^2	\mathbf{t}_j^2				
(o-m) $\mathbf{v} \in (-1/2 - \mathbf{t}, \mathbf{a})$	$j^{-2\mathbf{a}}$	$j^{-2\mathbf{t}}$	$n^{-\frac{2(\mathbf{a}-\mathbf{v})}{2\mathbf{a}+2\mathbf{t}+1}}$	$j^{-2(\mathbf{a}-\mathbf{v})}$	$j^{2(\mathbf{t}+\mathbf{v}-\mathbf{a})}$	$k^{-\frac{\mathbf{a}-\mathbf{v}}{\mathbf{t}}}$
$\mathbf{a} - \mathbf{v} \leq \mathbf{t}$						k^{-1}
$\mathbf{a} - \mathbf{v} \geq \mathbf{t}$						
(o-s) $\mathbf{a} - \mathbf{v} \in \mathbb{R}_{>0}$	$j^{-2\mathbf{a}}$	$e^{-j^{2\mathbf{t}}}$	$(\log n)^{-\frac{\mathbf{a}-\mathbf{v}}{\mathbf{t}}}$	$j^{-2(\mathbf{a}-\mathbf{v})}$	$j^{2(\mathbf{v}-\mathbf{a})} e^{j^{2\mathbf{t}}}$	$(\log k)^{-\frac{\mathbf{a}-\mathbf{v}}{\mathbf{t}}}$
(s-m) $\mathbf{v} + \mathbf{t} + 1/2 \in \mathbb{R}_{>0}$	$e^{-j^{2\mathbf{a}}}$	$j^{-2\mathbf{t}}$	$n^{-1}(\log n)^{\frac{2(\mathbf{t}+\mathbf{v})+1}{2\mathbf{a}}}$	$j^{2\mathbf{v}} e^{-j^{2\mathbf{a}}}$	$j^{2(\mathbf{t}+\mathbf{v})} e^{-j^{2\mathbf{a}}}$	k^{-1}

We note that in case **(o-m)** and **(s-m)** for $\mathbf{v} + \mathbf{t} < -1/2$ the rate $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ is parametric. \square

§07|02|03 Local and maximal local ϕ -risk

Secondly, we measure the accuracy of the tOPE $\widehat{\theta}^m := \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^m$ of $\theta^m = \mathfrak{s}^\dagger g^m \in \mathbb{J} \mathbb{1}^m$ with $g = \mathfrak{s} \theta \in \text{dom}(M_{\mathfrak{s}})$ and $\mathfrak{s} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ by the mean of its local ϕ -error introduced in §04103102, i.e. its ϕ -risk.

§07102.44 **Reminder.** If $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\theta \in \text{dom}(\phi\nu)$, then for each $m \in \mathbb{N}$ we have $\theta^m \in \text{dom}(\phi\nu)$ too and $|\phi\nu(\theta) - \phi\nu(\theta^m)| = o(1)$ as $m \rightarrow \infty$ (**Property** §04103.13). \square

§07102.45 **Assumption.** Let $(\widehat{g}, \widehat{\mathfrak{s}}) = (g + n^{-1/2} \dot{\mathfrak{e}}, \mathfrak{s} + k^{-1/2} \dot{\mathfrak{n}}) \sim \mathbb{P}_{\theta_{\mathfrak{s}}}^{n \otimes k} := \mathbb{P}_{\theta_{\mathfrak{s}}}^n \otimes \mathbb{P}_{\mathfrak{s}}^k$ be independent noisy versions satisfying Assumption §07102.01. In addition

(dSIPI1) $\dot{\mathfrak{e}}$ admit a covariance operator, say $\Gamma_{\theta_{\mathfrak{s}}} \in \mathbb{L}(\mathbb{J})$, i.e. $\dot{\mathfrak{e}} \sim P_{(\mathbb{1}, \Gamma_{\theta_{\mathfrak{s}}})}$, $K_{\theta_{\mathfrak{s}}}^2 := 1 \vee \|\Gamma_{\theta_{\mathfrak{s}}}\|_{\mathbb{L}(\mathbb{J})}$,

(dSIPI2) $\dot{\mathfrak{e}} \mathbb{1}^m \in \mathbb{J} = \mathbb{L}_2(\nu)$ $\mathbb{P}_{\theta_{\mathfrak{s}}}^n$ -a.s. for each $m \in \mathbb{N}$, and

(dSIPnO) $\mathbf{v}^{\mathfrak{s}(2)} := \mathbb{P}_{\mathfrak{s}}^k(\dot{\mathfrak{n}}^{\mathfrak{A}}) := (\mathbf{v}_j^{\mathfrak{s}(2)} := \mathbb{P}_{\mathfrak{s}}^k(\dot{\mathfrak{n}}_j^{\mathfrak{A}}))_{j \in \mathcal{J}} \in \mathbb{L}_\infty(\nu)$, $K_{\mathfrak{s}}^4 := 1 \vee \|\mathbf{v}^{\mathfrak{s}(2)}\|_{\mathbb{L}_\infty(\nu)}$.

Moreover, from (dSIPnO) (i.e. $\mathbf{v}^{\mathfrak{s}(2)} = \mathbb{P}_{\mathfrak{s}}^k(\dot{\mathfrak{n}}^{\mathfrak{A}}) \in \mathbb{L}_\infty(\nu)$) follows $\mathbb{P}_{\mathfrak{s}}^k(\dot{\mathfrak{n}}_j^{\mathfrak{A}}) =: \mathbf{v}_j^{\mathfrak{s}} \leq (\mathbf{v}_j^{\mathfrak{s}(2)})^{1/2}$ for ν -a.e. $j \in \mathcal{J}$, and hence $\|\mathbf{v}^{\mathfrak{s}} \vee \mathbb{1}\|_{\mathbb{L}_\infty(\nu)} \leq K_{\mathfrak{s}}^2$. \square

§07102.46 **Notation.** Since $\|\widehat{\mathfrak{s}}^{(k)\dagger}\|_{\mathbb{L}_\infty(\nu)} \leq k^{1/2}$ (**Notation** §07102.03), $\mathfrak{s} \in \mathbb{L}_\infty(\nu)$ and $\mathbb{1}^m \in \mathbb{L}_2(\nu) = \mathbb{J}$ (using $\nu(\llbracket m \rrbracket) \in \mathbb{R}_{\geq 0}$ by Assumption §07100.02) for all $m \in \mathbb{N}$, for $(\widehat{\mathfrak{s}}^{(k)\dagger} \mathfrak{s}) \cdot \mathbb{1}^m := \widehat{\mathfrak{s}}^{(k)\dagger} \mathfrak{s} \cdot \mathbb{1}^m \in \mathcal{M}(\mathcal{A} \otimes \mathcal{J})$ we have $(\widehat{\mathfrak{s}}^{(k)\dagger} \mathfrak{s}) \cdot \mathbb{1}^m \in \mathbb{L}_\infty(\nu)$ for all $m \in \mathbb{N}$ too. If in addition $\mathbb{1}^m \in \mathbb{L}_2(\phi^2\nu)$ for all $m \in \mathbb{N}$ then for $(\widehat{\mathfrak{s}}^{(k)\dagger} \phi) \cdot \mathbb{1}^m := \widehat{\mathfrak{s}}^{(k)\dagger} \phi \cdot \mathbb{1}^m \in \mathcal{M}(\mathcal{A} \otimes \mathcal{J})$ we also have $(\widehat{\mathfrak{s}}^{(k)\dagger} \phi) \cdot \mathbb{1}^m \in \mathbb{J}$ for all $m \in \mathbb{N}$. \square

§07102.47 **Comment.** Under Assumption §07102.45 and $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ if $\mathbb{1}^m \in \mathbb{L}_2(\phi^2\nu)$ for all $m \in \mathbb{N}$ then we have $(\widehat{\mathfrak{s}}^{(k)\dagger} \dot{\mathfrak{e}}) \cdot \mathbb{1}^m \in \text{dom}(\phi\nu)$ $\mathbb{P}_{\theta_{\mathfrak{s}}}^{n \otimes k}$ -a.s. (since $\nu(|(\widehat{\mathfrak{s}}^{(k)\dagger} \dot{\mathfrak{e}}) \cdot \mathbb{1}^m|) \leq \|(\widehat{\mathfrak{s}}^{(k)\dagger} \dot{\mathfrak{e}}) \cdot \mathbb{1}^m\|_{\mathbb{J}} \|\dot{\mathfrak{e}} \cdot \mathbb{1}^m\|_{\mathbb{J}} \in \mathbb{R}_{\geq 0}$ $\mathbb{P}_{\theta_{\mathfrak{s}}}^{n \otimes k}$ -a.s.). If in addition $\theta \in \text{dom}(\phi\nu)$, and hence $\theta^m \in \text{dom}(\phi\nu)$ (**Property** §04103.13), then it follows

$$\widehat{\theta}^m = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g} \cdot \mathbb{1}^m = n^{-1/2} \widehat{\mathfrak{s}}^{(k)\dagger} \dot{\mathfrak{e}} \cdot \mathbb{1}^m + (\widehat{\mathfrak{s}}^{(k)\dagger} \mathfrak{s}) \cdot \theta^m \in \text{dom}(\phi\nu) \quad \mathbb{P}_{\theta_{\mathfrak{s}}}^{n \otimes k}\text{-a.s.} \quad (07.43)$$

If $\mathcal{J} \subseteq \mathbb{Z}$ (at most countable) and $\nu_{\mathcal{J}}$ is the counting measure over the index set \mathcal{J} then Assumption §01101.04 and (dSIPI1) (i.e. $\dot{\mathfrak{e}} \sim P_{(\mathbb{1}, \Gamma_{\theta_{\mathfrak{s}}})}$) implies $\mathbf{v}^{\theta_{\mathfrak{s}}} = \mathbb{P}_{\theta_{\mathfrak{s}}}^n(|\dot{\mathfrak{e}}|^2) \in \mathbb{L}_\infty(\nu_{\mathcal{J}})$ and hence the

additional assumption (dSIPI2) $\dot{\mathfrak{s}} \mathbf{1}^m \in \mathbb{J} = \mathbb{L}_2(\nu_{\mathcal{J}}) \mathbb{P}_{\mathfrak{s}}^n$ -a.s.. However, the last implication does generally not hold, if $\mathcal{J} \in \{\mathbb{R}, \mathbb{R}_{\geq 0}\}$ for example. \square

§07|02|03|01 Local ϕ -risk

§07|02.48 **Assumption.** Let $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$, $\theta \in \text{dom}(\phi\nu)$, $\mathfrak{s}^\dagger \mathbf{1}^m, \mathbf{1}^m \in \mathbb{L}_2(\phi^2\nu)$ for $m \in \mathbb{N}$ be satisfied. \square

§07|02.49 **Definition.** Under Assumptions §07|02.45 and §07|02.48 for $m \in \mathbb{N}$ the *local ϕ -risk* of a thresholded OPE $\widehat{\theta}^m = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^m = \widehat{\mathfrak{s}}^\dagger \mathbf{1}_{\{\widehat{\mathfrak{s}}^2 \geq k^{-1}\}} \widehat{g} \mathbf{1}^m \in \text{dom}(\phi\nu) \mathbb{P}_{\mathfrak{s}}^{n \otimes k}$ -a.s. satisfies

$$\begin{aligned} \mathbb{E}_{\mathfrak{s}}^{n \otimes k} (|\phi\nu(\widehat{\theta}^m - \theta)|^2) &= \mathbb{E}_{\mathfrak{s}}^{n \otimes k} (|\nu((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) (\widehat{g} - g) \mathbf{1}^m|^2) \\ &\quad + \mathbb{E}_{\mathfrak{s}}^k (|\nu((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) \mathbf{1}^m - \mathbf{1} \cdot) \phi \theta|^2). \end{aligned} \quad (07.44)$$

with *variance* $\mathbb{E}_{\mathfrak{s}}^{n \otimes k} (|\nu((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) (\widehat{g} - g) \mathbf{1}^m|^2)$ and *bias* $\mathbb{E}_{\mathfrak{s}}^k (|\nu((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) \mathbf{1}^m - \mathbf{1} \cdot) \phi \theta|^2)$. \square

§07|02.50 **Property.** Under Assumptions §07|02.45 and §07|02.48 (exploiting the independence of $(\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot$ and $\dot{\mathfrak{s}}, \dot{\mathfrak{s}} \sim \mathbb{P}_{(\mathbf{1}, \mathbb{I}_{\mathfrak{s}})}$ with $\Gamma_{\mathfrak{s}} \in \mathbb{L}(\mathbb{J})$, and $(\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot \mathbf{1}^m \in \mathbb{L}_2(\nu)$) we have

$$\begin{aligned} n \mathbb{E}_{\mathfrak{s}}^{n \otimes k} (|\nu((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) (\widehat{g} - g) \mathbf{1}^m|^2) &= \mathbb{E}_{\mathfrak{s}}^{n \otimes k} (|\nu((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) \dot{\mathfrak{s}} \mathbf{1}^m|^2) \\ &= \mathbb{E}_{\mathfrak{s}}^k \langle \Gamma_{\mathfrak{s}}((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) \mathbf{1}^m, (\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot \mathbf{1}^m \rangle_{\mathbb{J}} \leq \|\Gamma_{\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} \nu(\mathbb{E}_{\mathfrak{s}}^k (|\dot{\mathfrak{s}}^{(k)\dagger}|^2) \phi^2 \mathbf{1}^m). \end{aligned}$$

Since $\mathfrak{s} \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ and $\mathfrak{s}^\dagger \mathbf{1}^m \in \mathbb{L}_2(\phi^2\nu)$ the last bound together with **Lemma** §07|02.18 (i) implies

$$\begin{aligned} n \mathbb{E}_{\mathfrak{s}}^{n \otimes k} (|\nu((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) (\widehat{g} - g) \mathbf{1}^m|^2) &\leq \|\Gamma_{\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} \nu(\mathbb{E}_{\mathfrak{s}}^k ((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) (\mathfrak{s}^\dagger \phi)^2 \mathbf{1}^m) \\ &\leq \|\Gamma_{\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} 2(\|\mathfrak{v}^{\mathfrak{s}}\|_{\mathbb{L}_{\infty}(\nu)} + 1) \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\phi}^2 \leq \|\Gamma_{\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} 4\|\mathfrak{v}^{\mathfrak{s}} \vee \mathbf{1}\|_{\mathbb{L}_{\infty}(\nu)} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\phi}^2. \end{aligned} \quad (07.45)$$

Moreover, assuming $\theta \in \text{dom}(\phi\nu)$ and hence $\theta \mathbf{1}^m \in \text{dom}(\phi\nu)$, we obtain (using $\widehat{\mathfrak{s}}^{(k)\dagger} \widehat{\mathfrak{s}} = \mathbf{1}_{\{\widehat{\mathfrak{s}}^2 < k^{-1}\}}$ and applying the generalised Minkowski inequality)

$$\begin{aligned} \frac{1}{3} \mathbb{E}_{\mathfrak{s}}^k (|\nu((\widehat{\mathfrak{s}}^{(k)\dagger}) \cdot) \mathbf{1}^m - \mathbf{1} \cdot) \phi \theta|^2) &- |\phi\nu(\theta \mathbf{1}^{m \perp})|^2 \\ &\leq \mathbb{E}_{\mathfrak{s}}^k (|\nu(\widehat{\mathfrak{s}}^{(k)\dagger}) (\mathfrak{s} - \widehat{\mathfrak{s}}) \phi \theta \mathbf{1}^m|^2) + \mathbb{E}_{\mathfrak{s}}^k (|\nu(\mathbf{1}_{\{\widehat{\mathfrak{s}}^2 < k^{-1}\}} \phi \theta \mathbf{1}^m)|^2) \\ &\leq |\nu(|\phi \theta \mathbf{1}^m| (\mathbb{E}_{\mathfrak{s}}^k (|\mathfrak{s} - \widehat{\mathfrak{s}}|^2 |\widehat{\mathfrak{s}}^{(k)\dagger}|^2))^{1/2})|^2 + |\nu(|\phi \theta \mathbf{1}^m| (\mathbb{E}_{\mathfrak{s}}^k (\widehat{\mathfrak{s}}^2 < k^{-1}))^{1/2})|^2 \\ &\leq 2(\|\mathfrak{v}^{\mathfrak{s}(2)}\|_{\mathbb{L}_{\infty}(\nu)} + \|\mathfrak{v}^{\mathfrak{s}}\|_{\mathbb{L}_{\infty}(\nu)} + 2\|\mathfrak{v}^{\mathfrak{s}} \vee \mathbf{1}\|_{\mathbb{L}_{\infty}(\nu)}) \|\theta \mathbf{1}^m (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi|\nu)}^2 \end{aligned} \quad (07.46)$$

where the last inequality follows from **Lemma** §07|02.18 (ii) and (iii). \square

§07|02.51 **Reminder.** If Assumptions §07|01.46 and §07|01.48 are satisfied then for all $m, n \in \mathbb{N}$ setting

$$\begin{aligned} \mathbb{R}_n^m(\theta, \mathfrak{s}, \phi) &:= |\phi\nu(\theta \mathbf{1}^{m \perp})|^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\phi}^2, \quad m_n^\circ := \arg \min \{ \mathbb{R}_n^m(\theta, \mathfrak{s}, \phi) : m \in \mathbb{N} \} \\ \text{and } \mathbb{R}_n^\circ(\theta, \mathfrak{s}, \phi) &:= \mathbb{R}_n^{m_n^\circ}(\theta, \mathfrak{s}, \phi) := \min \{ \mathbb{R}_n^m(\theta, \mathfrak{s}, \phi) : m \in \mathbb{N} \} \end{aligned} \quad (07.47)$$

the OPE $\widehat{\theta}^{m_n^\circ} := \widehat{\mathfrak{s}}^\dagger \widehat{g}^{m_n^\circ}$ with known $\mathfrak{s} \in \mathbb{L}_{\infty}(\nu)$ fulfills $\mathbb{E}_{\mathfrak{s}}^n (|\phi\nu(\widehat{\theta}^{m_n^\circ} - \theta)|^2) \leq (1 \vee \|\Gamma_{\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})}) \mathbb{R}_n^\circ(\theta, \mathfrak{s}, \phi)$ due to **Proposition** §07|01.51. Keep in mind that Assumption §07|01.46 is part of Assumption §07|02.45 and that Assumption §07|01.48 is part of Assumption §07|02.48. \square

§07|02.52 **Proposition (Upper bound).** Let Assumptions §07|02.45 and §07|02.48 be satisfied. The thresholded OPE $\widehat{\theta}^m = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^m \in \text{dom}(\phi\nu) \mathbb{P}_{\mathfrak{s}}^{n \otimes k}$ -a.s. for all $n, k, m \in \mathbb{N}$ fulfills

$$\begin{aligned} \mathbb{E}_{\mathfrak{s}}^{n \otimes k} (|\phi\nu(\widehat{\theta}^m - \theta)|^2) &\leq 4(\|\Gamma_{\mathfrak{s}}\|_{\mathbb{L}(\mathbb{J})} \vee 1) \|\mathfrak{v}^{\mathfrak{s}} \vee \mathbf{1}\|_{\mathbb{L}_{\infty}(\nu)} \mathbb{R}_n^m(\theta, \mathfrak{s}, \phi) \\ &\quad + 6(\|\mathfrak{v}^{\mathfrak{s}(2)}\|_{\mathbb{L}_{\infty}(\nu)} + 3\|\mathfrak{v}^{\mathfrak{s}} \vee \mathbf{1}\|_{\mathbb{L}_{\infty}(\nu)}) \|\theta \mathbf{1}^m (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi|\nu)}^2 \end{aligned} \quad (07.48)$$

$$\leq 4K_{\mathfrak{s}}^2 K_{\mathfrak{s}}^2 \mathbb{R}_n^m(\theta, \mathfrak{s}, \phi) + 24K_{\mathfrak{s}}^4 \|\theta \mathbf{1}^m (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi|\nu)}^2. \quad (07.49)$$

§0702.53 **Proof** of **Proposition** §0702.52. Given in the lecture. \square

§0702.54 **Comment**. Selecting $m_n^\circ := \arg \min \{R_n^m(\theta, \mathfrak{s}, \phi) : m \in \mathbb{N}\}$ and $R_n^\circ(\theta, \mathfrak{s}, \nu) = R_n^{m_n^\circ}(\theta, \mathfrak{s}, \phi)$ as in (07.47) (**Reminder** §0702.51) from **Proposition** §0702.52 we obtain immediately

$$\mathbb{E}_{\theta|\mathfrak{s}}^{n \otimes k} (|\phi \nu(\widehat{\theta}^{m_n^\circ} - \theta)|^2) \leq 4K_{\mathfrak{s}}^2 K_{\theta|\mathfrak{s}}^2 R_n^\circ(\theta, \mathfrak{s}, \phi) + 24K_{\mathfrak{s}}^4 \|\theta \mathbf{1}^{m_n^\circ} (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi| \nu)}^2. \quad (07.50)$$

We shall emphasise, that $R_n^\circ(\theta, \mathfrak{s}, \phi)$ is the oracle rate (**Property** §0701.55) if $M_{\mathfrak{s}} \in \mathbb{L}(\mathbb{D})$ is *known* in advance. Furthermore, for each $m \in \mathbb{N}$ we have

$$\begin{aligned} \|\theta \mathbf{1}^m (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi| \nu)} &\leq \|\theta (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi| \nu)} \\ &= \|\theta (1 \vee k \mathfrak{s}^2)^{-1/2} \mathbf{1}^m\|_{\mathbb{L}_1(|\phi| \nu)} + \|\theta (1 \vee k \mathfrak{s}^2)^{-1/2} \mathbf{1}^{m \perp}\|_{\mathbb{L}_1(|\phi| \nu)} \\ &\leq \|\theta (1 \vee k \mathfrak{s}^2)^{-1/2} \mathbf{1}^m\|_{\mathbb{L}_1(|\phi| \nu)} + \|\theta \mathbf{1}^{m \perp}\|_{\mathbb{L}_1(|\phi| \nu)}. \end{aligned} \quad (07.51)$$

Consequently, under the assumptions of **Proposition** §0702.52 from (??) (**Proof** §0702.53) follows

$$\begin{aligned} \mathbb{E}_{\theta|\mathfrak{s}}^{n \otimes k} (|\phi \nu(\widehat{\theta}^m - \theta)|^2) &\leq 4K_{\mathfrak{s}}^2 K_{\theta|\mathfrak{s}}^2 R_n^m(\theta, \mathfrak{s}, \phi) + 24K_{\mathfrak{s}}^4 \|\theta \mathbf{1}^m (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi| \nu)}^2 \\ &\leq 4K_{\mathfrak{s}}^2 K_{\theta|\mathfrak{s}}^2 R_n^m(\theta, \mathfrak{s}, \phi) + 24K_{\mathfrak{s}}^4 \|\theta (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi| \nu)}^2 \\ &\leq 28K_{\mathfrak{s}}^4 K_{\theta|\mathfrak{s}}^2 (\|\theta \mathbf{1}^{m \perp}\|_{\mathbb{L}_1(|\phi| \nu)}^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\phi}^2) + 24K_{\mathfrak{s}} \|\theta \mathbf{1}^m (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi| \nu)}^2. \end{aligned}$$

Selecting $m_n^\circ := \arg \min \{\|\theta \mathbf{1}^{m \perp}\|_{\mathbb{L}_1(|\phi| \nu)}^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\phi}^2 : m \in \mathbb{N}\}$ we obtain

$$\begin{aligned} \mathbb{E}_{\theta|\mathfrak{s}}^{n \otimes k} (|\phi \nu(\widehat{\theta}^{m_n^\circ} - \theta)|^2) &\leq 3K_{\mathfrak{s}}^{1/2} K_{\theta|\mathfrak{s}} \min \{\|\theta \mathbf{1}^{m \perp}\|_{\mathbb{L}_1(|\phi| \nu)}^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\phi}^2 : m \in \mathbb{N}\} \\ &\quad + 24K_{\mathfrak{s}} \|\theta (1 \vee k \mathfrak{s}^2)^{-1/2}\|_{\mathbb{L}_1(|\phi| \nu)}^2. \end{aligned}$$

We shall emphasise, that the last upper bound consists (up to the constants) of the sum of the two terms depending each on one of the sample sizes n and k only. However, the first term $\min \{\|\theta \mathbf{1}^{m \perp}\|_{\mathbb{L}_1(|\phi| \nu)}^2 + n^{-1} \|\mathfrak{s}^\dagger \mathbf{1}^m\|_{\phi}^2 : m \in \mathbb{N}\}$ is generally larger than the oracle rate $R_n^\circ(\theta, \mathfrak{s}, \phi)$. \square

§0702.55 **Corollary** (GdiSM with noisy operator §0702.05 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{\mathfrak{s}}) = (g + n^{-1/2} \dot{B}, \mathfrak{s} + k^{-1/2} \dot{W}) \sim N_{\theta|\mathfrak{s}}^{n \otimes k} = N_{\theta|\mathfrak{s}}^n \otimes N_{\mathfrak{s}}^k$ as in Model §0702.05, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{W} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ are independent, $\mathfrak{s} \in \mathbb{R}_{\nu_0}^{\mathbb{N}} \cap \ell_\infty$ and $\theta \in \ell_2$, and hence $g = \mathfrak{s} \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \ell_2$. Given $\phi \in \mathbb{R}_{\nu_0}^{\mathbb{N}}$ and $\theta \in \text{dom}(\phi_{\nu_{\mathbb{N}}})$ the (infeasible) thresholded OPE $\widehat{\theta}^{m_n^\circ} = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^{m_n^\circ} \in \text{dom}(\phi_{\nu_{\mathbb{N}}})$ with oracle dimension m_n° as in (07.47) satisfies*

$$N_{\theta|\mathfrak{s}}^{n \otimes k} (|\phi_{\nu_{\mathbb{N}}}(\widehat{\theta}^{m_n^\circ} - \theta)|^2) \leq 4R_n^\circ(\theta, \mathfrak{s}, \phi) + 36\|(1 \vee k \mathfrak{s}^2)^{-1/2} \theta\|_{\ell_1(|\phi|)}^2 \quad (07.52)$$

where $R_n^\circ(\theta, \mathfrak{s}, \phi)$ is the oracle rate in a GdiSM §0701.03 (see **Corollary** §0701.57).

§0702.56 **Proof** of **Corollary** §0702.55. Given in the lecture. \square

§0702.57 **Corollary** (diSM with noisy operator §0702.07 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{\mathfrak{s}}) = (g + n^{-1/2} \dot{\epsilon}, \mathfrak{s} + k^{-1/2} \dot{\eta}) \sim P_{\theta|\mathfrak{s}|\sigma|\xi}^{n \otimes k} = P_{\theta|\mathfrak{s}|\sigma}^n \otimes P_{\mathfrak{s}|\xi}^k$ as in Model §0702.07, where $\dot{\epsilon}$ and $\dot{\eta}$ satisfy (iSM1) and (diSMnO1) with $K_\sigma := \|\sigma\|_{\ell_\infty} \vee 1 \in \mathbb{R}_{\geq 1}$ and $K_\xi := \|\xi\|_{\ell_\infty} \vee 1 \in \mathbb{R}_{\geq 1}$, respectively, $\mathfrak{s} \in \mathbb{R}_{\nu_0}^{\mathbb{N}} \cap \ell_\infty$ and $\theta \in \ell_2$, and hence $g = \mathfrak{s} \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \ell_2$. Given $\phi \in \mathbb{R}_{\nu_0}^{\mathbb{N}}$ and $\theta \in \text{dom}(\phi_{\nu_{\mathbb{N}}})$ the (infeasible) thresholded OPE $\widehat{\theta}^{m_n^\circ} = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^{m_n^\circ} \in \text{dom}(\phi_{\nu_{\mathbb{N}}})$ with oracle dimension m_n° as in (07.47) satisfies*

$$P_{\theta|\mathfrak{s}|\sigma|\xi}^{n \otimes k} (|\phi_{\nu_{\mathbb{N}}}(\widehat{\theta}^{m_n^\circ} - \theta)|^2) \leq 4K_\sigma^2 K_\xi^2 R_n^\circ(\theta, \mathfrak{s}, \phi) + 24K_\xi^4 \|(1 \vee k \mathfrak{s}^2)^{-1/2} \theta\|_{\ell_1(|\phi|)}^2 \quad \forall n, k \in \mathbb{N} \quad (07.53)$$

where $R_n^\circ(\theta, \mathfrak{s}, \phi)$ is the oracle rate in a diSM §0701.06 (see **Corollary** §0701.59).

§07102.58 **Proof of Corollary §07102.57.** Given in the lecture. \square

§07102.59 **Corollary** (dieMM with noisy operator §07102.09 continued). *Consider independent noisy versions $\widehat{g}_j = g_j + n^{-1/2}\widehat{\epsilon}_j$ and $\widehat{s}_j = s_j + k^{-1/2}\widehat{\eta}_j$ defined on $(\mathcal{Z}^{n+k}, \mathcal{L}^{\otimes(n+k)}, \mathbb{P}_{\theta|s}^{n \otimes k} = \mathbb{P}_{\theta|s}^n \otimes \mathbb{P}_s^k)$ as in Model §07102.09, where $\psi, \varphi \in \mathcal{M}(\mathcal{Z} \otimes \mathcal{J})$ satisfies (dieMM1)–(dieMM3) (Model §07101.08) and (dieMMnO1)–(dieMMnO2) (Model §07102.09) with $v_{\theta|s|\psi}, v_{s|\varphi} \in \mathbb{R}_{>1}$, respectively, $s_* \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J}) \cap \mathbb{L}_{\infty}(\nu)$, $\theta \in \mathbb{J}$ and hence $g_* = s_*\theta \in \text{dom}(M_s) \subseteq \ell_2$. Given Assumption §07102.48 the (infeasible) tOPE $\widehat{\theta}^{m_n^o} = \widehat{s}_*^{(k)\dagger} \widehat{g}_*^{m_n^o} \in \text{dom}(\phi\nu)$ with oracle dimension m_n^o as in (07.29) satisfies*

$$\mathbb{P}_{\theta|s}^{n \otimes k} (|\phi\nu_N(\widehat{\theta}^{m_n^o} - \theta)|^2) \leq 4v_{\theta|s|\psi} v_{s|\varphi} R_n^o(\theta, s_*, \phi) + 24v_{s|\varphi}^2 \|(1 \vee k s_*^2)^{-1/2} \theta\|_{\ell_1(|\phi|)}^2 \quad \forall n, k \in \mathbb{N} \quad (07.54)$$

where $R_n^o(\theta, s_*, \phi)$ is the oracle rate in a dieMM §07101.08 (see Corollary §07101.61).

§07102.60 **Proof of Corollary §07102.59.** Given in the lecture. \square

§07102.61 **Illustration.** We illustrate the last results considering usual behaviour for $\theta, s_*, \phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$. We distinguish again the two cases **(p)** and **(np)** in Illustration §07101.63, where in case **(p)** the term $R_n^o(\theta, s_*, \phi)$ is parametric, that is, $nR_n^o(\theta, s_*, \phi) = O(1)$, in case **(np)** it is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^o(\theta, s_*, \phi) = \infty$. In case **(np)** we consider again the three specifications **(o-m)**, **(o-s)** and **(s-m)** introduced in Illustration §07101.63 where also in Table 03 [§07] the order of the oracle dimension m_n^o and the oracle rate $R_n^o(\theta, s_*, \phi)$ as $n \rightarrow \infty$ are given. The next table depict the oracle rate $R_n^o(\theta, s_*, \phi)$ and the rate of the additional term $\|(1 \vee k s_*^2)^{-1/2} \theta\|_{\ell_1(|\phi|)}^2$ as $n, k \rightarrow \infty$:

Table 07 [§07]

Order of $R_n^o(\theta, s_*, \phi)$ and $\|(1 \vee k s_*^2)^{-1/2} \theta\|_{\ell_1(|\phi|)}^2$ as $n, k \rightarrow \infty$

	$(j \in \mathcal{J})$	$(a \in \mathbb{R}_{>0})$	$(t \in \mathbb{R}_{>0})$			
	$\phi_j = j^{v-1/2}$	θ_j	s_{jt}	$R_n^o(\theta, s_*, \phi)$	$\theta_j \phi_j$	$s_{jt}^\dagger \theta_j \phi_j$
(o-m)	$v \in (-t, a)$	$j^{-a-1/2}$	j^{-2t}	$n^{-\frac{a-v}{a+t}}$	j^{v-a-1}	$j^{t+v-a-1}$
	$a - v < t$					$k^{-\frac{a-v}{t}}$
	$a - v = t$					$k^{-1}(\log k)^2$
	$a - v > t$					k^{-1}
(o-s)	$a - v \in \mathbb{R}_{>0}$	$j^{-a-1/2}$	$e^{-j^{2t}}$	$(\log n)^{-\frac{a-v}{t}}$	j^{v-a-1}	$j^{v-a-1} e^{j^{2t}}$
(s-m)	$v + t \in \mathbb{R}_{>0}$	$e^{-j^{2a}}$	j^{-2t}	$\frac{(\log n)^{\frac{t+v}{a}}}{n}$	$j^{v-1/2} e^{-j^{2a}}$	$j^{t+v-1/2} e^{-j^{2a}}$
						k^{-1}

We note that in case **(o-m)** and **(s-m)** for $v < -t$ the oracle rate $R_n^o(\theta, s_*, \phi)$ is parametric. \square

§07102|03|02 Maximal local ϕ -risk

§07102.62 **Assumption.** Consider weights $\alpha_*, t_* \in \mathcal{M}_{>0, \nu}(\mathcal{J})$ and $\phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$ (i.e. $\nu(\mathcal{N}_{\alpha_*}) = \nu(\mathcal{N}_{t_*}) = 0 = \nu(\mathcal{N}_{\phi})$), such that $\alpha_*, t_* \in \mathbb{L}_{\infty}(\nu)$, $\alpha_* \in \mathbb{L}_2(\phi^2\nu)$, and $t_*^\dagger \mathbb{1}_*^m, \mathbb{1}_*^m \in \mathbb{L}_2(\phi^2\nu)$ for all $m \in \mathbb{N}$. \square

§07102.63 **Reminder.** Under Assumption §07102.62 we have $\mathbb{J}^a = \mathbb{L}_2(\nu) = \text{dom}(M_{\alpha_*}) = \mathbb{J}\alpha_* \subseteq \mathbb{J}$ and the three measures ν , $\alpha_*^{2t_*}\nu$ and $|\phi|\nu$ dominate mutually each other, i.e. they share the same null sets (see Property §04101.02). We consider \mathbb{J}^a endowed with $\|\cdot\|_{\alpha_*} = \|M_{\alpha_*} \cdot\|_{\mathbb{J}}$ and given a constant $r \in \mathbb{R}_{>0}$ the ellipsoid $\mathbb{J}^{a,r} := \{h_* \in \mathbb{J}^a : \|h_*\|_{\alpha_*} \leq r\} \subseteq \mathbb{J}^a$. Since $\alpha_* \in \mathbb{L}_2(\phi^2\nu)$, and hence $\|\alpha_* \mathbb{1}_*^{m\perp}\|_{\phi} = \|(\alpha\phi) \mathbb{1}_*^{m\perp}\|_{\mathbb{J}} \in \mathbb{R}_{\geq 0}$ for each $m \in \mathbb{N}$ ($\|\alpha_* \mathbb{1}_*^{m\perp}\|_{\phi} = o(1)$ as $m \rightarrow \infty$ by dominated convergence) we have $\mathbb{J}^a \subseteq \text{dom}(\phi\nu)$ (Property §04102.23), and $|\phi\nu(\theta_* \mathbb{1}_*^{m\perp})| \leq r \|\alpha_* \mathbb{1}_*^{m\perp}\|_{\phi}$ for all $\theta_* \in \mathbb{J}^{a,r}$ (Lemma §04102.25). Let in addition $M_s \in \mathbb{M}_{t_*, d}$ satisfy a link condition as in Definition §04103.05

with weights $\mathfrak{t}_j \in \mathcal{M}_{>0,\nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$, and radius $d \in \mathbb{R}_{>0}$. We set $(\mathfrak{t}^\dagger \phi)_j := (\mathfrak{t}_j^\dagger \phi_j)_{j \in \mathcal{J}} = \mathfrak{t}^\dagger \phi \in \mathcal{J}$. Obviously, for $m \in \mathbb{N}$ the condition $\mathfrak{t}^\dagger \mathbf{1}^m \in \mathbb{L}_2(\phi^2 \nu)$ due to Assumption §07102.62 implies $\mathfrak{s}^\dagger \mathbf{1}^m \in \mathbb{L}_2(\phi^2 \nu)$ too. Consequently, if Assumption §07102.62, $\theta \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}$ and $M_s \in \mathbb{M}_{\mathfrak{t},d}$ are satisfied, then Assumption §07102.48 is also fulfilled. Keep in mind if Assumptions §07101.48 and §07101.64 are satisfied, then for $n, m \in \mathbb{N}$ setting

$$\begin{aligned} R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) &:= \|\mathfrak{a} \mathbf{1}^{m \perp}\|_\phi^2 + n^{-1} \|\mathfrak{t}^\dagger \mathbf{1}^m\|_\phi^2, \quad m_n^* := \arg \min \{R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) : m \in \mathbb{N}\} \\ &\text{and} \quad R_n^*(\mathfrak{a}, \mathfrak{t}, \phi) := R_n^{m_n^*}(\mathfrak{a}, \mathfrak{t}, \phi) = \min \{R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) : m \in \mathbb{N}\} \end{aligned} \quad (07.55)$$

for all $\theta = \mathfrak{s}^\dagger g \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}$, known $M_s \in \mathbb{M}_{\mathfrak{t},d}$ and $n \in \mathbb{N}$ the OPE $\widehat{\theta}^{m_n^*} := \mathfrak{s}^\dagger \widehat{g}^{m_n^*}$ fulfills

$$\mathbb{E}_{\theta|s}^n (|\phi \nu (\widehat{\theta}^{m_n^*} - \theta)|^2) \leq (d^2 \|\Gamma_{\theta|s}\|_{\mathbb{L}(\mathcal{J})} \vee r^2) R_n^*(\mathfrak{a}, \mathfrak{t}, \phi).$$

due to **Corollary** §07101.69. We shall emphasise that Assumptions §07102.45 and §07102.62 contains Assumptions §07101.46 and §07101.64, respectively. \square

§07102.64 **Proposition Upper bound.** *Let Assumptions §07102.45 and §07102.62 be satisfied. If $M_s \in \mathbb{M}_{\mathfrak{t},d}$ with $d \in \mathbb{R}_{>0}$, and $\theta \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}$ with $r \in \mathbb{R}_{>0}$, then for all $n, k, m \in \mathbb{N}$ the thresholded OPE $\widehat{\theta}^m = \widehat{\mathfrak{s}}^{(k)|\dagger} \widehat{g}^m \in \text{dom}(\phi \nu)$ $\mathbb{E}_{\theta|s}^{n \otimes k}$ -a.s. fulfills*

$$\begin{aligned} \mathbb{E}_{\theta|s}^{n \otimes k} (|\phi \nu (\widehat{\theta}^m - \theta)|^2) &\leq (3r^2 \vee 4(\|\Gamma_{\theta|s}\|_{\mathbb{L}(\mathcal{J})} \vee 1) \|\mathfrak{v}^s \vee \mathbf{1}\|_{\mathbb{L}_\infty(\nu)} d^2) R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) \\ &\quad + 6(\|\mathfrak{v}^{s(2)}\|_{\mathbb{L}_\infty(\nu)} + 3\|\mathfrak{v}^s \vee \mathbf{1}\|_{\mathbb{L}_\infty(\nu)}) d^2 r^2 \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a} \mathbf{1}^m\|_\phi^2 \end{aligned} \quad (07.56)$$

$$\leq (3r^2 \vee 4K_{\theta|s}^2 K_s^2 d^2) R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) + 24K_s^4 d^2 r^2 \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a} \mathbf{1}^m\|_\phi^2. \quad (07.57)$$

§07102.65 **Proof of Proposition** §07102.64. Given in the lecture. \square

§07102.66 **Remark.** Selecting $m_n^* := \arg \min \{R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) : m \in \mathbb{N}\}$ and $R_n^*(\mathfrak{a}, \mathfrak{t}, \phi) = R_n^{m_n^*}(\mathfrak{a}, \mathfrak{t}, \phi)$ as in (07.55) (**Reminder** §07102.63) we obtain

$$\mathbb{E}_{\theta|s}^{n \otimes k} (|\phi \nu (\widehat{\theta}^{m_n^*} - \theta)|^2) \leq (3r^2 \vee 4K_{\theta|s}^2 K_s^2 d^2) R_n^*(\mathfrak{a}, \mathfrak{t}, \phi) + 24K_s^4 d^2 r^2 \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a}\|_\phi^2 \quad (07.58)$$

where $R_n^*(\mathfrak{a}, \mathfrak{t}, \phi)$ is the rate (**Corollary** §07101.69) if $M_s \in \mathbb{M}(\mathbb{J})$ is known in advance. Furthermore, for each $m \in \mathbb{N}$ we have

$$\begin{aligned} \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a} \mathbf{1}^m\|_\phi^2 &\leq \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a}\|_\phi^2 = \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a} \mathbf{1}^m\|_\phi^2 + \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a} \mathbf{1}^{m \perp}\|_\phi^2 \\ &\leq \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a} \mathbf{1}^m\|_\phi^2 + \|\mathfrak{a} \mathbf{1}^{m \perp}\|_\phi^2. \end{aligned} \quad (07.59)$$

Consequently, under the assumptions of **Proposition** §07102.64 from (??) (**Proof** §07102.65) follows immediately

$$\begin{aligned} \mathbb{E}_{\theta|s}^{n \otimes k} (|\phi \nu (\widehat{\theta}^m - \theta)|^2) &\leq (3r^2 \vee 4K_{\theta|s}^2 K_s^2 d^2) R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) + 24K_s^4 d^2 r^2 \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a} \mathbf{1}^m\|_\phi^2 \\ &\leq (3r^2 \vee 4K_{\theta|s}^2 K_s^2 d^2) R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) + 24K_s^4 d^2 r^2 \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a}\|_\phi^2 \\ &\leq (3r^2 \vee 4K_{\theta|s}^2 K_s^2 d^2 + 24K_s^4 d^2 r^2) R_n^m(\mathfrak{a}, \mathfrak{t}, \phi) + 24K_s^4 d^2 r^2 \|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a} \mathbf{1}^m\|_\phi^2 \end{aligned}$$

We shall emphasise, that the upper bound (07.58) consists (up to the constants) of the sum of the two terms $R_n^*(\mathfrak{a}, \mathfrak{t}, \phi)$ and $\|(1 \vee k \mathfrak{t}^\dagger)^{-1/2} \mathfrak{a}\|_\phi^2$ depending each on one of the sample sizes n and k only. If in addition there exists $\mathfrak{v} \in \mathbb{R}_{>0}$ satisfying $\mathfrak{v} \geq (K_{\theta|s} \vee K_s)$ for all $\theta \in \mathbb{J}^{\mathfrak{a},\mathfrak{r}}$ and

$M_s \in \mathbb{M}_{t,d}$ then for all $n, k \in \mathbb{N}$ the maximal local ϕ -risk of the thresholded OPE $\widehat{\theta}^{m_n^*}$ with optimally chosen dimension m_n^* is bounded by

$$\begin{aligned} \sup \{ \mathbb{P}_{\theta|s}^{n \otimes k} (|\phi \nu(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \mathbb{J}^{a,r}, M_s \in \mathbb{M}_{t,d} \} \\ \leq (3r^2 \vee 4K_{\theta|s}^2 K_s^2 d^2 + 24K_s^4 d^2 r^2) R_n^*(\mathbf{a}, \mathbf{t}, \phi) \vee \|(1 \vee k \mathbf{t}^2)^{-1/2} \mathbf{a}\|_{\phi}^2. \end{aligned}$$

Arguing similarly as in Remark §07101.21 we note that $R_n^*(\mathbf{a}, \mathbf{t}, \phi) = o(1)$ as $n \rightarrow \infty$ since $\dagger \mathbf{1}^m \in \mathbb{L}_2(\phi^2 \nu)$ for all $m \in \mathbb{N}$ and $\|\mathbf{a}, \mathbf{1}^{m \perp}\|_{\phi}^2 = o(1)$ as $m \rightarrow \infty$ by Assumption §07102.62. Moreover, we have $\|(1 \vee k \mathbf{t}^2)^{-1/2} \mathbf{a}\|_{\phi}^2 = o(1)$ as $k \rightarrow \infty$ by dominated convergence. Note that the dimension $m_n^* := m_n^*(\mathbf{a}, \mathbf{t}, \phi)$ does depend neither on the unknown parameter of interest θ nor on the unknown operator M_s but on the classes $\mathbb{J}^{a,r}$ and $\mathbb{M}_{t,d}$ only, and thus also the statistic $\widehat{\theta}^{m_n^*}$. In other words, if the regularity of θ and M_s is known in advance, then the OPE $\widehat{\theta}^{m_n^*}$ is a feasible estimator. \square

§07102.67 **Corollary** (GdiSM with noisy operator §07102.05 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{s}) = (g + n^{-1/2} \dot{B}, s + k^{-1/2} \dot{W}) \sim N_{\theta|s}^{n \otimes k} = N_{\theta|s}^n \otimes N_s^k$ as in Model §07102.05, where $\dot{B} \sim N_{(0,1)}^{\otimes N}$ and $\dot{W} \sim N_{(0,1)}^{\otimes N}$ are independent, $s \in \mathbb{R}_{\setminus 0}^N \cap \ell_{\infty}$, $\theta \in \ell_2$, and hence $g = s \cdot \theta \in \text{dom}(M_s) \subseteq \ell_2$. Under Assumption §07102.62 the thresholded OPE $\widehat{\theta}^{m_n^*} = \widehat{s}^{(k)\dagger} \widehat{g}^{m_n^*} \in \text{dom}(\phi_{\mathbb{K}})$ with dimension m_n^* as in (07.55) satisfies*

$$\begin{aligned} \sup \{ N_{\theta|s}^{n \otimes k} (|\phi_{\mathbb{K}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{a,r}, M_s \in \mathbb{M}_{t,d} \} \\ \leq C_{r,d} R_n^*(\mathbf{a}, \mathbf{t}, \phi) \vee \|(1 \vee k \mathbf{t}^2)^{-1/2} \mathbf{a}\|_{\phi}^2. \quad (07.60) \end{aligned}$$

with constant $C_{r,d} = 3r^2 \vee 4d^2 + 36r^2 d^2$.

§07102.68 **Proof of Corollary** §07102.67. Given in the lecture. \square

§07102.69 **Corollary** (diSM with noisy operator §07102.07 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{s}) = (g + n^{-1/2} \dot{\epsilon}, s + k^{-1/2} \dot{\eta}) \sim P_{\theta|s|\sigma|\xi}^{n \otimes k} = P_{\theta|s|\sigma}^n \otimes P_{s|\xi}^k$ as in Model §07102.07, where $\dot{\epsilon}$ and $\dot{\eta}$ satisfy (iSM1) and (diSMnO1) with $K_{\sigma} := \|\sigma\|_{\ell_{\infty}} \vee 1 \in \mathbb{R}_{\geq 1}$ and $K_{\xi} := \|\xi\|_{\ell_{\infty}} \vee 1 \in \mathbb{R}_{\geq 1}$, respectively, $s \in \mathbb{R}_{\setminus 0}^N \cap \ell_{\infty}$ and $\theta \in \ell_2$, and hence $g = s \cdot \theta \in \text{dom}(M_s) \subseteq \ell_2$. Under Assumption §07102.62 the thresholded OPE $\widehat{\theta}^{m_n^*} = \widehat{s}^{(k)\dagger} \widehat{g}^{m_n^*} \in \text{dom}(\phi_{\mathbb{K}})$ with dimension m_n^* as in (07.55) satisfies*

$$\begin{aligned} \sup \{ P_{\theta|s|\sigma|\xi}^{n \otimes k} (|\phi_{\mathbb{K}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{a,r}, M_s \in \mathbb{M}_{t,d} \} \\ \leq C_{r,d,\sigma,\xi} R_n^*(\mathbf{a}, \mathbf{t}, \phi) \vee \|(1 \vee k \mathbf{t}^2)^{-1/2} \mathbf{a}\|_{\phi}^2 \quad (07.61) \end{aligned}$$

with constant $C_{r,d,\sigma,\xi} = 3r^2 \vee 4K_{\xi}^2 K_{\sigma}^2 d^2 + 24K_{\xi}^4 r^2 d^2$.

§07102.70 **Proof of Corollary** §07102.69. Given in the lecture. \square

§07102.71 **Corollary** (dieMM with noisy operator §07102.09 continued). *Consider independent noisy versions $\widehat{g} = g + n^{-1/2} \dot{\epsilon}$ and $\widehat{s} = s + k^{-1/2} \dot{\eta}$ defined on $(\mathcal{Z}^{n+k}, \mathcal{X}^{\otimes(n+k)}, \mathbb{P}_{\theta|s}^{n \otimes k} = \mathbb{P}_{\theta|s}^{\otimes n} \otimes \mathbb{P}_s^{\otimes k})$ as in Model §07102.09, where $\psi, \varphi \in \mathcal{M}(\mathcal{X} \otimes \mathcal{Y})$ satisfies (dieMM1)–(dieMM3) and (dieMMnO1)–(dieMMnO2) with $\nu_{\theta|s|\psi}, \nu_{s|\varphi} \in \mathbb{R}_{\geq 1}$, respectively, $s \in \mathcal{M}_{\neq 0, \nu}(\mathcal{Y}) \cap \mathbb{L}_{\infty}(\nu)$, $\theta \in \mathbb{J}$ and hence $g = s \cdot \theta \in \text{dom}(M_s) \subseteq \mathbb{J}$. Under Assumption §07102.62 the thresholded OPE $\widehat{\theta}^{m_n^*} = \widehat{s}^{(k)\dagger} \widehat{g}^{m_n^*} \in \text{dom}(\phi_{\mathbb{K}})$ with dimension m_n^* as in (07.55) satisfies*

$$\begin{aligned} \sup \{ P_{\theta|s|\sigma|\xi}^{n \otimes k} (|\phi_{\mathbb{K}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \mathbb{J}^{a,r}, M_s \in \mathbb{M}_{t,d} \} \\ \leq C_{r,d} R_n^*(\mathbf{a}, \mathbf{t}, \phi) \vee \|(1 \vee k \mathbf{t}^2)^{-1/2} \mathbf{a}\|_{\phi}^2 \quad (07.62) \end{aligned}$$

with constant $C_{r,d} = 3r^2 \vee 4d^2 \sup \{ \nu_{\theta|s|\psi} \nu_{s|\varphi} : \theta \in \mathbb{J}^{a,r}, M_s \in \mathbb{M}_{t,d} \} + 24r^2 d^2 \sup \{ \nu_{s|\varphi}^2 : M_s \in \mathbb{M}_{t,d} \}$.

§07102.72 **Proof** of **Corollary** §07102.71. Given in the lecture. \square

§07102.73 **Illustration**. We illustrate the last results considering usual behaviour for $\mathbf{a}, \mathbf{t}, \phi \in \mathcal{M}_{\neq 0, \nu}(\mathcal{J})$. We distinguish again the two cases **(p)** and **(np)** in **Illustration** §07102.73 where in case **(p)** the term $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ is parametric, that is, $nR_n^*(\mathbf{a}, \mathbf{t}, \phi) = O(1)$, in case **(np)** it is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^*(\mathbf{a}, \mathbf{t}, \phi) = \infty$. In case **(np)** we consider again the three specifications **(o-m)**, **(o-s)** and **(s-m)** introduced in **Illustration** §07101.78 where also in Table 04 [§07] the order of the dimension m_n^* and the rate $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ as $n \rightarrow \infty$ are given. The next table depicts the rate $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ and the additional term $\|(1 \vee k \mathbf{t}^2)^{-1/2} \mathbf{a}\|_\phi^2$ as $n, k \rightarrow \infty$ only:

Table 08 [§07]

Order of $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ and $\|(1 \vee k \mathbf{t}^2)^{-1/2} \mathbf{a}\|_\phi^2$ as $n, k \rightarrow \infty$

	$(j \in \mathcal{J})$	$(\mathbf{a} \in \mathbb{R}_{>0}) (\mathbf{t} \in \mathbb{R}_{>0})$				
	$\phi_j^2 = j^{2v-1}$	\mathbf{a}_j^2	\mathbf{t}_j^2	$R_n^*(\mathbf{a}, \mathbf{t}, \phi)$	$(\mathbf{a}\phi)_j^2$	$\mathbf{t}_j^{-2}(\mathbf{a}\phi)_j^2$
(o-m)	$v \in (-t, a)$	j^{-2a}	j^{-2t}	$n^{-\frac{a-v}{a+t}}$	$j^{-2(a-v)-1}$	$j^{2(t+v-a)-1}$
	$a - v \leq t$					$k^{-\frac{a-v}{t}}$
	$a - v \geq t$					k^{-1}
(o-s)	$a - v \in \mathbb{R}_{>0}$	j^{-2a}	$e^{-j^{2t}}$	$(\log n)^{-\frac{a-v}{t}}$	$j^{2(v-a)-1}$	$j^{2(v-a)-1} e^{j^{2t}}$
(s-m)	$v + t \in \mathbb{R}_{>0}$	$e^{-j^{2a}}$	j^{-2t}	$n^{-1}(\log n)^{\frac{t+v}{a}}$	$j^{2v-1} e^{-j^{2a}}$	$j^{2(t+v)-1} e^{-j^{2a}}$

We note that in case **(o-m)** and **(s-m)** for $v < -t$ the rate $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ is parametric. \square

§08 (Generalised) Galerkin estimator

§08100.01 **Notation (Reminder)**. Consider $\mathbb{J} = \ell_2 := \mathbb{L}_2(\nu_N) = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_N)$ with counting measure $\nu_N := \sum_{j \in \mathbb{N}} \delta_{\{j\}}$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ and $V \in \mathbb{L}(\mathbb{G}, \ell_2)$. For each $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ and $T_{\bullet, \bullet} := VTU^* \in \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$ (compare **Notation** §01104.03) we identify the kernel (infinite dimensional matrix) $T_{\bullet, \bullet} = (T_{j,j'})_{j,j' \in \mathbb{N}} \in \mathbb{R}^{\mathbb{N}^2}$ and the map from ℓ_2 into itself given by

$$\mathbf{a} \mapsto T_{\bullet, \bullet} \mathbf{a} := \left(\sum_{j' \in \mathbb{N}} T_{j,j'} a_{j'} \right)_{j \in \mathbb{N}} = \langle T_{j, \bullet}, \mathbf{a} \rangle_{\ell_2} = \nu_N(T_{j, \bullet} \mathbf{a})_{j \in \mathbb{N}}$$

(compare **Notation** §01105.01). Moreover, we denote by $\mathbb{L}^{\succ}(\ell_2)$ the subset of all strictly positive definite operator in $\mathbb{L}(\ell_2)$. For each $T_{\bullet, \bullet} \in \mathbb{L}^{\succ}(\ell_2)$ we denote its Moore-Penrose inverse by $T_{\bullet, \bullet}^\dagger : \ell_2 \supseteq \text{dom}(T_{\bullet, \bullet}^\dagger) \rightarrow \ell_2$ (see **Definition** §03100.08). We denote by $\mathbb{L}^{\mathbb{R}}(\ell_2)$ the subset of all *injective* $A_{\bullet, \bullet} \in \mathbb{L}(\ell_2)$ such that $[A_{\bullet, \bullet}]_m \in \mathbb{R}^{(m,m)}$ is *regular* for all $m \in \mathbb{N}$. For each $m \in \mathbb{N}$ and $A_{\bullet, \bullet} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$, the inverse $[A_{\bullet, \bullet}]_m^{-1} \in \mathbb{R}^{(m,m)}$ of $[A_{\bullet, \bullet}]_m \in \mathbb{R}^{(m,m)}$ exists. Note that $\mathbb{L}^{\succ}(\ell_2) \subseteq \mathbb{L}^{\mathbb{R}}(\ell_2)$ (**Lemma** §05101.22). \square

§08100.02 **Assumption**. For $\mathbb{J} = \ell_2$, surjective partial isometries $U \in \mathbb{L}(\mathbb{H}, \ell_2)$ and $V \in \mathbb{L}(\mathbb{G}, \ell_2)$ fixed and presumed to be *known* in advance, the operator $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$ satisfies either $T_{\bullet, \bullet} = VTU^* \in \mathbb{L}^{\succ}(\ell_2) \subseteq \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$ or more generally $T_{\bullet, \bullet} = VTU^* \in \mathbb{L}^{\mathbb{R}}(\ell_2) \subseteq \mathbb{L}(\ell_2) = \mathbb{L}(\ell_2)$. Let $g \in \text{dom}(T_{\bullet, \bullet}^\dagger) = \text{ran}(T_{\bullet, \bullet})$, and hence $\theta = T_{\bullet, \bullet}^\dagger g = T_{\bullet, \bullet}^{-1} g \in \ell_2$. \square

§08100.03 **Reminder**. Under Assumption §08100.02 we consider $T_{\bullet, \bullet} \in \mathbb{L}^{\succ}(\ell_2)$ or more generally $T_{\bullet, \bullet} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ and $g \in \text{dom}(T_{\bullet, \bullet}^\dagger) = \text{ran}(T_{\bullet, \bullet})$, and hence $\theta = T_{\bullet, \bullet}^\dagger g = T_{\bullet, \bullet}^{-1} g \in \ell_2$. For each $m \in \mathbb{N}$ and $A_{\bullet, \bullet} \in \mathbb{L}(\ell_2)$ we write $A_{\bullet, \bullet}^m := M_{1^m} A_{\bullet, \bullet} M_{1^m} \in \mathbb{L}(\ell_2)$, which restricted to an operator from \mathbb{R}^m ($\text{ran}(M_{1^m}) = \ell_2 \mathbf{1}^m$) to itself, can be represented by the matrix $[A_{\bullet, \bullet}]_m \in \mathbb{R}^{(m,m)}$ (see **Notation** §05100.02). If $[A_{\bullet, \bullet}]_m^\dagger \in \mathbb{R}^{(m,m)}$

denotes the Moore-Penrose inverse of $[A_{\cdot, \cdot}]_m$ (as linear map from \mathbb{R}^m into itself), then the Moore-Penrose inverse $A_{\cdot, \cdot}^{m \dagger} = (A_{\cdot, \cdot}^m)^\dagger \in \mathbb{L}(\ell_2)$ of $A_{\cdot, \cdot}^m$ (see Definition §03100.08), restricted to an operator from \mathbb{R}^m to itself can be represented by the matrix $[A_{\cdot, \cdot}^{m \dagger}]_m$. In particular, if $A_{\cdot, \cdot} \in \mathbb{L}^{\cong}(\ell_2)$, i.e. $[A_{\cdot, \cdot}]_m$ is *regular* (invertible), and hence $[A_{\cdot, \cdot}^\dagger]_m = [A_{\cdot, \cdot}]_m^{-1}$, then we have $A_{\cdot, \cdot}^m A_{\cdot, \cdot}^{m \dagger} = M_{\mathbb{1}^m} = A_{\cdot, \cdot}^{m \dagger} A_{\cdot, \cdot}^m$. For $T_{\cdot, \cdot} \in \mathbb{L}^{\cong}(\ell_2)$ and $m \in \mathbb{N}$ we call any element $\theta^m \in \ell_2 \mathbb{1}^m$ i.e. $0 = \theta^m (\mathbb{1} - \mathbb{1}^m) = \theta^m \mathbb{1}^{m \perp}$, satisfying

$$\langle \theta^m, T_{\cdot, \cdot} \theta^m \rangle_{\ell_2} - 2 \langle \theta^m, g \rangle_{\ell_2} \leq \langle a, T_{\cdot, \cdot} a \rangle_{\ell_2} - 2 \langle a, g \rangle_{\ell_2} \quad \text{for all } a \in \ell_2 \mathbb{1}^m$$

a *Galerkin solution* in $\ell_2 \mathbb{1}^m$. Since $T_{\cdot, \cdot} \in \mathbb{L}^{\cong}(\ell_2)$ the Galerkin solution is *uniquely* determined by $[\theta^m]_m = [T_{\cdot, \cdot}]_m^{-1} [g]_m$, and hence $\theta^m = T_{\cdot, \cdot}^{m \dagger} g$ (Lemma §05101.03). More generally, under Assumption §08100.02 with $T_{\cdot, \cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ we call the unique solution $\theta^m = T_{\cdot, \cdot}^{m \dagger} g$ of $[T_{\cdot, \cdot}]_m [\theta^m]_m = [g]_m$ *generalised Galerkin solution* (Definition §05102.01). Keep in mind that $\mathbb{L}^{\cong}(\ell_2) \subseteq \mathbb{L}^{\mathbb{R}}(\ell_2)$ (Lemma §05101.22). \square

§08|01 Non-diagonal statistical inverse problem

§08101.01 **Assumption.** Consider a stochastic process $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathbb{N}}$ on a probability space $(\Omega, \mathcal{A}, \mathbb{P})$ satisfying Assumption §01101.04 (i.e. $\dot{\epsilon} \in \mathcal{M}(\mathcal{A} \otimes 2^{\mathbb{N}})$) with *mean zero* (i.e. $\mathbb{P}(\dot{\epsilon}) = (\mathbb{P}(\dot{\epsilon}_j))_{j \in \mathbb{N}} = 0$), a sample size $n \in \mathbb{N}$ and let Assumption §08100.02 be satisfied where $T_{\cdot, \cdot} \in \mathbb{L}^{\cong}(\ell_2)$ or $T_{\cdot, \cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ is *known* in advance. For $\theta \in \ell_2$ the observable noisy image with mean $g = T_{\cdot, \cdot} \theta \in \ell_2$ takes the form $\hat{g} = g + n^{-1/2} \dot{\epsilon}$. We denote by $\mathbb{P}_{\theta \mid T}^n$ the distribution of \hat{g} . In addition (nSIP) $\dot{\epsilon}$ admits a covariance operator, say $\Gamma_{\theta \mid T} \in \mathbb{L}^{\cong}(\ell_2)$ with $\|\Gamma_{\theta \mid T}\|_{\mathbb{L}(\ell_2)} \leq \nu_{\theta \mid T} \in \mathbb{R}_{\geq 1}$. \square

§08101.02 **Definition.** Under Assumption §08101.01 for $\theta \in \ell_2$ and $T_{\cdot, \cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ consider a noisy version $\hat{g} \sim \mathbb{P}_{\theta \mid T}^n$ of $g = T_{\cdot, \cdot} \theta \in \text{dom}(T_{\cdot, \cdot}^\dagger)$. For each $m \in \mathbb{N}$ we call $\hat{\theta}^m = T_{\cdot, \cdot}^{m \dagger} \hat{g} = T_{\cdot, \cdot}^{m \dagger} \hat{g} \mathbb{1}^m = T_{\cdot, \cdot}^{m \dagger} \hat{g}^m$ (generalised) *Galerkin estimator (GE)* of $\theta = T_{\cdot, \cdot}^\dagger g = T_{\cdot, \cdot}^{-1} g \in \ell_2$. \square

§08101.03 **Comment.** The (generalised) Galerkin solution $\theta^m = T_{\cdot, \cdot}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ does generally not correspond to the orthogonal projection $\mathbb{1}^{m \perp} \theta = (\mathbb{1} - \mathbb{1}^m) \theta$. Moreover, the approximation error $\sup \{\|\theta^j - \theta\|_{\ell_2} : j \in \mathbb{N}_{\geq m}\}$ does generally not converge to zero as $m \rightarrow \infty$ (compare Remark §05101.05). Here and subsequently, we will restrict ourselves to classes of solutions and operators which ensure the convergence. Obviously, this is a minimal regularity condition for us if we aim to estimate the Galerkin solution.

§08|01|01 Examples

§08101.04 **GniSM (§01105.08 continued).** Let Assumption §08100.02 be satisfied where $T_{\cdot, \cdot} \in \mathbb{L}^{\cong}(\ell_2)$ or $T_{\cdot, \cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ is *known* in advance. We illustrate the (generalised) GE in a Gaussian non-diagonal inverse sequence model (GniSM) as in §01105.08. Here the observable stochastic process $\hat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta \mid T}^n$ is a noisy version of $g = T_{\cdot, \cdot} \theta \in \text{dom}(T_{\cdot, \cdot}^\dagger) \subseteq \ell_2$ with $\theta \in \Theta \subseteq \ell_2$ and $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$. Consequently, \hat{g} admits a $N_{\theta \mid T}^n$ -distribution belonging to the family $N_{\Theta \times \{T_{\cdot, \cdot}\}}^n := (N_{\theta \mid T}^n)_{\theta \in \Theta}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, N_{\Theta \times \{T_{\cdot, \cdot}\}}^n)$ where $\Theta \subseteq \ell_2$. \square

§08101.05 **Reminder** (GniSM §08101.04 continued). Due to Property §07101.04 the error process $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ admits as covariance operator $\Gamma_{\theta \mid \mathfrak{s}} = \text{id}_{\ell_2} \in \mathbb{L}^{\cong}(\ell_2)$ and hence Assumption §08101.01 is satisfied. \square

§08101.06 **niSM (§01105.07 continued).** Let Assumption §08100.02 be satisfied where $T_{\cdot, \cdot} \in \mathbb{L}^{\cong}(\ell_2)$ or $T_{\cdot, \cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ is *known* in advanced. We illustrate the (generalised) GE in a Non-diagonal inverse sequence model (niSM) as in §01105.07. Here the observable stochastic process $\hat{g} = g + n^{-1/2} \dot{\epsilon}$ is a

noisy version of $g = T_{\cdot, \cdot} \theta \in \text{dom}(T_{\cdot, \cdot}^\dagger) \subseteq \ell_2$ with $\theta \in \Theta \subseteq \ell_2$ and $\dot{\epsilon} \sim \otimes_{j \in \mathbb{N}} \mathbb{P}^{\dot{\epsilon}_j}$ satisfying (iSM1) and (iSM2) introduced in diSM §07101.06. Under (iSM1) \widehat{g} admits a $\mathbb{P}_{\theta|T|g}^n$ -distribution belonging to the family $\mathbb{P}_{\Theta \times \{T_{\cdot, \cdot}\} \times \Sigma}^n := (\mathbb{P}_{\theta|T|g}^n)_{\theta \in \Theta, \alpha \in \Sigma}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, \mathbb{P}_{\Theta \times \{T_{\cdot, \cdot}\} \times \Sigma}^n)$ where $\Theta \subseteq \ell_2$ and $\Sigma \subseteq \mathbb{R}_{>0} \cap \ell_\infty$. \square

§08101.07 **Reminder** (niSM §08101.06 continued). Due to **Property** §07101.07 (i) under (iSM1) the process $\dot{\epsilon} \sim \otimes_{j \in \mathbb{N}} \mathbb{P}_{(0, \sigma^2)}$ admits as covariance operator $\Gamma_{\theta|g} = M_{\sigma^2} \in \mathbb{L}(\ell_2) \cap \mathbb{L}^{\geq}(\ell_2)$ and hence Assumption §08101.01 is satisfied. \square

§08101.08 **nieMM** (§01105.05 continued). Let Assumption §08100.02 be satisfied where $T_{\cdot, \cdot} \in \mathbb{L}^{\geq}(\ell_2)$ or $T_{\cdot, \cdot} \in \mathbb{L}^{\leq}(\ell_2)$ is known in advanced. We illustrate the (generalised) GE in a Non-diagonal inverse empirical mean model (nieMM) as in §01105.05. Here the observable stochastic process $\widehat{g} = g + n^{-1/2} \dot{\epsilon}$ is a noisy version of $g = T_{\cdot, \cdot} \theta \in \text{dom}(T_{\cdot, \cdot}^\dagger) \subseteq \ell_2$ with $\theta = g \in \Theta \subseteq \ell_2$, and error process $\dot{\epsilon} = n^{1/2}(\widehat{\mathbb{P}}_n(\psi) - \mathbb{P}_{\theta|T}(\psi)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes 2^{\mathbb{N}})$ satisfying Assumption §01101.04. More precisely, on a measurable space $(\mathcal{Z}, \mathcal{Z})$ for $T_{\cdot, \cdot} \in \mathbb{L}^{\leq}(\ell_2)$ and for each $\theta \in \Theta \subseteq \ell_2$ there is a probability measure $\mathbb{P}_{\theta|T} \in \mathcal{W}(\mathcal{Z})$. Consider a stochastic process $\psi = (\psi_j)_{j \in \mathbb{N}} \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}})$ which similar to (dieMM1)–(dieMM4) introduced in dieMM §07101.08 in addition for $T_{\cdot, \cdot} \in \mathbb{L}^{\leq}(\ell_2)$ (or $T_{\cdot, \cdot} \in \mathbb{L}^{\geq}(\ell_2)$) and for each $\theta \in \Theta \subseteq \ell_2$ satisfies

(nieMM1) $\psi_j \in \mathcal{L}_1(\mathbb{P}_{\theta|T}) := \mathcal{L}_1(\mathcal{Z}, \mathcal{Z}, \mathbb{P}_{\theta|T})$ for all $j \in \mathbb{N}$ and $\mathbb{P}_{\theta|T}(\psi) = T_{\cdot, \cdot} \theta = g$,

(dieMM2) for each $m \in \mathbb{N}$ we have $(\psi - \mathbb{P}_{\theta|g}(\psi)) \mathbb{1}^m \in \ell_\infty$ $\mathbb{P}_{\theta|T}$ -a.s. due to (nieMM1),

(nieMM2) there is $\mathfrak{v}_{\theta|T|\psi} \in \mathbb{R}_{\geq 1}$ such that $\|\mathbb{P}_{\theta|T}(\psi^2)\|_{\ell_\infty} \leq \mathfrak{v}_{\theta|T|\psi}$ and

$$\mathbb{P}_{\theta|T}(|\nu_{\mathbb{N}}(h, \psi)|^2) \leq \mathfrak{v}_{\theta|T|\psi} \|h\|_{\ell_2}^2, \quad \forall h \in \ell_2,$$

(nieMM3) $\mathfrak{v}_{\cdot}^{\theta|T} := \mathbb{P}_{\theta|T}(\psi^2) - |\mathbb{P}_{\theta|T}(\psi)|^2 \in \mathbb{R}_{>0} \cap \ell_\infty$, $\|(\mathfrak{v}_{\cdot}^{\theta|T})^{-1}\|_{\ell_\infty} \leq \mathfrak{v}_{\theta|T|\psi}$ and

$$\mathbb{P}_{\theta|T}(|\nu_{\mathbb{N}}(h, \psi)|^2) \geq \mathbb{P}_{\theta|T}(|\nu_{\mathbb{N}}(h, \psi)|^2) - |\mathbb{P}_{\theta|T}(\nu_{\mathbb{N}}(h, \psi))|^2 \geq \mathfrak{v}_{\theta|T|\psi}^{-1} \|h\|_{\ell_2}^2, \quad \forall h \in \ell_2.$$

We consider a statistical product experiment $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, \mathbb{P}_{\Theta \times \{T_{\cdot, \cdot}\}}^{\otimes n} = (\mathbb{P}_{\theta|T}^{\otimes n})_{\theta \in \Theta})$ as in an Empirical mean function §01101.10 where $\Theta \subseteq \ell_2$. \square

§08101.09 **Reminder** (nieMM §08101.08 continued). Due to **Property** §07101.09 (i) under (nieMM1) and (nieMM2) the error process $\dot{\epsilon} = n^{1/2}(\widehat{\mathbb{P}}_n(\psi) - \mathbb{P}_{\theta|T}(\psi)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes 2^{\mathbb{N}})$ admits a covariance operator $\Gamma_{\theta|T} \in \mathbb{L}(\ell_2)$ and hence Assumption §08101.01 is satisfied. \square

§08|01|02 Global and maximal global \mathfrak{v} -risk

We measure first the accuracy of the (generalised) GE $\widehat{\theta}^m := T_{\cdot, \cdot}^{m|\dagger} \widehat{g}$ of the (generalised) Galerkin solution $\theta^m = T_{\cdot, \cdot}^{m|\dagger} g \in \ell_2 \mathbb{1}^m$ with $g = T_{\cdot, \cdot} \theta \in \text{dom}(T_{\cdot, \cdot}^\dagger)$ by the mean of its global \mathfrak{v} -error introduced in §05101101 and §05102101, i.e. its \mathfrak{v} -risk.

§08101.10 **Reminder**. If $\mathfrak{v} \in \mathbb{R}_{>0}^{\mathbb{N}}$ then we have $\mathfrak{v}^2 \mathbb{1}^m \in \ell_\infty$ and $\ell_2 \mathbb{1}^m \subseteq \ell_2(\mathfrak{v}^2)$. Consequently, for each $\theta \in \ell_2(\mathfrak{v}^2)$ the (generalised) Galerkin solution $\theta^m = T_{\cdot, \cdot}^{m|\dagger} g \in \ell_2 \mathbb{1}^m$ satisfies $\theta^m \in \ell_2(\mathfrak{v}^2)$ too. If in addition $C_T := \sup \{ \|M_{\mathfrak{v}} T_{\cdot, \cdot}^{m|\dagger} T_{\cdot, \cdot} M_{\mathbb{1}^m}^\perp\|_{\mathbb{L}(\ell_2)} : m \in \mathbb{N} \} \in \mathbb{R}_{>0}$ then $\|\theta^m - \theta\|_{\mathfrak{v}} \leq (1 + C_T) \|\mathbb{1}^{m|\perp} \theta\|_{\ell_2}$ which implies $\sup \{ \|\theta^j - \theta\|_{\mathfrak{v}} : j \in \mathbb{N}_{\geq m} \} = o(1)$ as $m \rightarrow \infty$ (**Property** §05101.24 and **Property** §05102.08). \square

§08101.11 **Comment**. Under Assumption §08101.01 since $\theta^m, T_{\cdot, \cdot}^{m|\dagger} \mathbb{1}^m \in \ell_2 \mathbb{1}^m$ for each $m \in \mathbb{N}$ we have $T_{\cdot, \cdot}^{m|\dagger} \dot{\epsilon} \in \ell_2 \mathbb{1}^m$ $\mathbb{P}_{\theta|T}^n$ -a.s.. Indeed, $\dot{\epsilon} \sim P_{(0, \Gamma_{\theta|T})}$ with $\Gamma_{\theta|T} \in \mathbb{L}(\ell_2)$ by Assumption §08101.01 (nSIP) implies

$\mathbb{P}_{\theta^T}^n(\dot{\hat{\epsilon}}^2) \in \ell_\infty$, hence $\dot{\hat{\epsilon}} \cdot \mathbf{1}^m \in \ell_\infty$ $\mathbb{P}_{\theta^T}^n$ -a.s. and $\|\mathbb{T}_{\cdot, \cdot}^{m \dagger} \dot{\hat{\epsilon}}\|_{\ell_2} \leq \|\mathbb{T}_{\cdot, \cdot}^{m \dagger} \mathbf{1}^m\|_{\ell_2} \|\dot{\hat{\epsilon}} \cdot \mathbf{1}^m\|_{\ell_\infty} \in \mathbb{R}_{\geq 0}$ $\mathbb{P}_{\theta^T}^n$ -a.s.. Given $\mathfrak{v} \in \mathbb{R}_{\setminus 0}^N$ from $\ell_2 \mathbf{1}^m \subseteq \ell_2(\mathfrak{v}^2)$ (Reminder §08|01.10) it follows

$$\hat{\theta}^m = \mathbb{T}_{\cdot, \cdot}^{m \dagger} \hat{g} = n^{-1/2} \mathbb{T}_{\cdot, \cdot}^{m \dagger} \dot{\hat{\epsilon}} + \theta^m \in \ell_2 \mathbf{1}^m \subseteq \ell_2(\mathfrak{v}^2) \quad \mathbb{P}_{\theta^T}^n\text{-a.s.} \quad \square$$

§08|01|02|01 Global \mathfrak{v} -risk

§08|01.12 **Assumption.** Let $\mathfrak{v} \in \mathbb{R}_{\setminus 0}^N$ and $\theta \in \ell_2(\mathfrak{v}^2)$ be satisfied. □

§08|01.13 **Definition.** Under Assumptions §08|01.01 and §08|01.12 the *global \mathfrak{v} -risk* of a (generalised) GE $\hat{\theta}^m = \mathbb{T}_{\cdot, \cdot}^{m \dagger} \hat{g} \in \ell_2 \mathbf{1}^m \subseteq \ell_2(\mathfrak{v}^2)$ $\mathbb{P}_{\theta^T}^n$ -a.s. satisfies

$$\mathbb{E}_{\theta^T}^n(\|\hat{\theta}^m - \theta\|_{\mathfrak{v}}^2) = \mathbb{E}_{\theta^T}^n\|\mathbb{T}_{\cdot, \cdot}^{m \dagger}(\hat{g} - g)\|_{\mathfrak{v}}^2 + \|\theta^m - \theta\|_{\mathfrak{v}}^2 \quad (08.01)$$

with *variance* $\mathbb{E}_{\theta^T}^n(\|\mathbb{T}_{\cdot, \cdot}^{m \dagger}(\hat{g} - g)\|_{\mathfrak{v}}^2) = n^{-1} \mathbb{E}_{\theta^T}^n(\|\mathbb{T}_{\cdot, \cdot}^{m \dagger} \dot{\hat{\epsilon}}\|_{\mathfrak{v}}^2)$ and *bias* $\|\theta^m - \theta\|_{\mathfrak{v}}$. □

§08|01.14 **Notation (Reminder).** Let $A \in \mathbb{L}(\ell_2)$ be a *Hilbert-Schmidt operator*, $A \in \mathbb{HS}(\ell_2)$ for short, where $\|A\|_{\mathbb{HS}}^2 := \text{tr}(A^*A) = \text{tr}(AA^*) \in \mathbb{R}_{\geq 0}$. If $\Gamma \in \mathbb{L}(\ell_2)$ then $\text{tr}(A^*\Gamma A) \leq \|\Gamma\|_{\mathbb{L}(\ell_2)} \text{tr}(A^*A) = \|\Gamma\|_{\mathbb{L}(\ell_2)} \|A\|_{\mathbb{HS}}^2$. For arbitrary $A \in \mathbb{L}(\ell_2)$ we have $M_{\mathfrak{v}} A^m = M_{\mathfrak{v}}^m A^m \in \mathbb{HS}(\ell_2)$. □

§08|01.15 **Property.** Under Assumptions §08|01.01 and §08|01.12 we have

$$\mathbb{E}_{\theta^T}^n(\|\mathbb{T}_{\cdot, \cdot}^{m \dagger} \dot{\hat{\epsilon}}\|_{\mathfrak{v}}^2) = \text{tr}(M_{\mathfrak{v}} \mathbb{T}_{\cdot, \cdot}^{m \dagger} \Gamma_{\theta^T} (\mathbb{T}_{\cdot, \cdot}^{m \dagger})^* M_{\mathfrak{v}}) = \text{tr}([M_{\mathfrak{v}}]_{\underline{m}} [\mathbb{T}_{\cdot, \cdot}]_{\underline{m}}^{-1} [\Gamma_{\theta^T}]_{\underline{m}} ([\mathbb{T}_{\cdot, \cdot}]_{\underline{m}}^{-1})^* [M_{\mathfrak{v}}]_{\underline{m}}) \quad (08.02)$$

and consequently $\mathbb{E}_{\theta^T}^n(\|\mathbb{T}_{\cdot, \cdot}^{m \dagger}(\hat{g} - g)\|_{\mathfrak{v}}^2) \leq n^{-1} \|\Gamma_{\theta^T}\|_{\mathbb{L}(\ell_2)} \|M_{\mathfrak{v}} \mathbb{T}_{\cdot, \cdot}^{m \dagger}\|_{\mathbb{HS}}^2 \in \mathbb{R}_{\geq 0}$. □

§08|01.16 **Proposition (Upper bound).** Under Assumptions §08|01.01 and §08|01.12 for all $n, m \in \mathbb{N}$ with (generalised) Galerkin solution $\theta^m = \mathbb{T}_{\cdot, \cdot}^{m \dagger} g \in \ell_2 \mathbf{1}^m$ setting

$$\begin{aligned} R_n^m(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) &:= \|\theta^m - \theta\|_{\mathfrak{v}}^2 + n^{-1} \|M_{\mathfrak{v}} \mathbb{T}_{\cdot, \cdot}^{m \dagger}\|_{\mathbb{HS}}^2, \quad m_n^\circ := \arg \min \{R_n^m(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) : m \in \mathbb{N}\} \\ \text{and } R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) &:= R_n^{m_n^\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) = \min \{R_n^m(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) : m \in \mathbb{N}\} \end{aligned} \quad (08.03)$$

we have $\mathbb{E}_{\theta^T}^n(\|\hat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2) \leq (1 \vee \|\Gamma_{\theta^T}\|_{\mathbb{L}(\ell_2)}) R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v})$ for all $n \in \mathbb{N}$.

§08|01.17 **Proof of Proposition §08|01.16.** Given in the lecture. □

§08|01.18 **Comment.** Let $A \in \mathbb{HS}(\ell_2)$ and $\Gamma \in \mathbb{L}^\times(\ell_2)$ be invertible with inverse $\Gamma^{-1} \in \mathbb{L}^\times(\ell_2)$. If we set $\mathfrak{v} := \max(\|\Gamma\|_{\mathbb{L}(\ell_2)}, \|\Gamma^{-1}\|_{\mathbb{L}(\ell_2)}) \in \mathbb{R}_{>0}$, then we have $\mathfrak{v}^{-1} \|A\|_{\mathbb{HS}}^2 \leq \text{tr}(A\Gamma A^*) \leq \mathfrak{v} \|A\|_{\mathbb{HS}}^2$ by using **Notation §08|01.14**. □

§08|01.19 **Oracle inequality.** Under Assumptions §08|01.01 and §08|01.12 if in addition

$$1 \leq \max(\|\Gamma_{\theta^T}\|_{\mathbb{L}(\ell_2)}, \|\Gamma_{\theta^T}^{-1}\|_{\mathbb{L}(\ell_2)}) \leq \mathfrak{v}_{\theta^T} \in \mathbb{R}_{\geq 1}$$

is satisfied, then (08.03) (and **Comment §08|01.18**) implies

$$\begin{aligned} \mathfrak{v}_{\theta^T}^{-1} R_n^m(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) &\leq \mathbb{E}_{\theta^T}^n \|\hat{\theta}^m - \theta\|_{\mathfrak{v}}^2 = n^{-1} \text{tr}(M_{\mathfrak{v}} \mathbb{T}_{\cdot, \cdot}^{m \dagger} \Gamma_{\theta^T} (\mathbb{T}_{\cdot, \cdot}^{m \dagger})^{-1} M_{\mathfrak{v}}) + \|\theta^m - \theta\|_{\mathfrak{v}}^2 \\ &\leq \mathfrak{v}_{\theta^T} R_n^m(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) \quad \forall n, m \in \mathbb{N}. \end{aligned}$$

As a consequence we immediately obtain the following *oracle inequality*

$$\begin{aligned} \mathfrak{v}_{\theta^T}^{-1} R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) &\leq \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta^T}^n(\|\hat{\theta}^m - \theta\|_{\mathfrak{v}}^2) \leq \mathbb{E}_{\theta^T}^n(\|\hat{\theta}^{m_n^\circ} - \theta\|_{\mathfrak{v}}^2) \\ &\leq \mathfrak{v}_{\theta^T} R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathfrak{v}) \leq \mathfrak{v}_{\theta^T}^2 \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta^T}^n(\|\hat{\theta}^m - \theta\|_{\mathfrak{v}}^2) \quad \forall n \in \mathbb{N}, \end{aligned} \quad (08.04)$$

and, hence $R_n^\circ(\theta, T_{\cdot, \cdot}, \mathbf{v})$, m_n° and the statistic $\widehat{\theta}_n^{m_n^\circ}$, respectively, is an *oracle bound*, an *oracle dimension* and *oracle optimal* (up to the constant $\mathfrak{v}_{\theta|T}^2$). \square

§08101.20 **Remark.** Arguing similarly as in **Remark** §07101.21 we note that $\|M_{\mathbf{v}} T_{\cdot, \cdot}^{m|\dagger}\|_{\text{HS}} \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and hence $R_n^\circ(\theta, T_{\cdot, \cdot}, \mathbf{v}) = o(1)$ as $n \rightarrow \infty$, whenever $\|\theta^m - \theta\|_{\mathbf{v}} = o(1)$ as $m \rightarrow \infty$ (see **Reminder** §08101.10). Note that the oracle dimension $m_n^\circ := m_n^\circ(\theta, T_{\cdot, \cdot}, \mathbf{v})$ as defined in **Proposition** §08101.16 depends on the unknown parameter of interest θ , and thus also the oracle optimal statistic $\widehat{\theta}_n^{m_n^\circ}$. In other words $\widehat{\theta}_n^{m_n^\circ}$ is not a feasible estimator. \square

§08101.21 **Corollary** (GniSM §08101.04 continued). Consider $\widehat{g}_\cdot = g_\cdot + n^{-1/2} \dot{B}_\cdot \sim N_{\theta|T}^n$ as in **Model** §08101.04, where $\dot{B}_\cdot \sim N_{(0,1)}^{\otimes \mathbb{N}}$, $T_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\mathbb{R}}(\ell_2)$ or $T_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\mathbb{R}}(\ell_2)$, $\theta \in \ell_2$, and hence $g_\cdot = T_{\cdot, \cdot} \theta \in \text{dom}(T_{\cdot, \cdot}^\dagger) \subseteq \ell_2$. Given **Assumption** §08101.12 the (infeasible, generalised) GE $\widehat{\theta}_\cdot^{m_n^\circ} = T_{\cdot, \cdot}^{m_n^\circ|\dagger} \widehat{g}_\cdot \in \ell_2 \mathbb{1}_\cdot^{m_n^\circ} \subseteq \ell_2(\mathbf{v}^2)$ with oracle dimension m_n° as in (08.03) satisfies

$$N_{\theta|T}^n(\|\widehat{\theta}_\cdot^{m_n^\circ} - \theta\|_{\mathbf{v}}^2) = R_n^\circ(\theta, T_{\cdot, \cdot}, \mathbf{v}) = \inf_{m \in \mathbb{N}} N_{\theta|T}^n(\|\widehat{\theta}_\cdot^m - \theta\|_{\mathbf{v}}^2) \quad \forall n \in \mathbb{N},$$

and hence it is *oracle optimal* (with constant 1).

§08101.22 **Proof of Corollary** §08101.21. Given in the lecture. \square

§08101.23 **Corollary** (niSM §08101.06 continued). Consider $\widehat{g}_\cdot = g_\cdot + n^{-1/2} \dot{\epsilon}_\cdot \sim P_{\theta|T|\sigma}^n$ as in **Model** §08101.06, where $\dot{\epsilon}_\cdot \sim \otimes_{j \in \mathbb{N}} P_{(0,\sigma_j^2)}$ satisfies **(iSM1)** and **(iSM2)** with $\max(\|\sigma_\cdot^{-2}\|_{\ell_\infty}, \|\sigma_\cdot^2\|_{\ell_\infty}) =: \mathfrak{v}_\sigma \in \mathbb{R}_{\geq 1}$, $T_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\mathbb{R}}(\ell_2)$ or $T_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\mathbb{R}}(\ell_2)$, $\theta \in \ell_2$, and hence $g_\cdot = T_{\cdot, \cdot} \theta \in \text{dom}(T_{\cdot, \cdot}^\dagger) \subseteq \ell_2$. Given **Assumption** §08101.12 the (infeasible, generalised) GE $\widehat{\theta}_\cdot^{m_n^\circ} = T_{\cdot, \cdot}^{m_n^\circ|\dagger} \widehat{g}_\cdot \in \ell_2 \mathbb{1}_\cdot^{m_n^\circ} \subseteq \ell_2(\mathbf{v}^2)$ with oracle dimension m_n° as in (08.03) satisfies

$$P_{\theta|T|\sigma}^n(\|\widehat{\theta}_\cdot^{m_n^\circ} - \theta\|_{\mathbf{v}}^2) \leq \mathfrak{v}_\sigma R_n^\circ(\theta, T_{\cdot, \cdot}, \mathbf{v}) \leq \mathfrak{v}_\sigma^2 \inf_{m \in \mathbb{N}} P_{\theta|T|\sigma}^n(\|\widehat{\theta}_\cdot^m - \theta\|_{\mathbf{v}}^2) \quad \forall n \in \mathbb{N},$$

and hence it is *oracle optimal* (with constant \mathfrak{v}_σ).

§08101.24 **Proof of Corollary** §08101.23. Given in the lecture. \square

§08101.25 **Corollary** (nieMM §08101.08 continued). Let $\widehat{g}_\cdot = g_\cdot + n^{-1/2} \dot{\epsilon}_\cdot$ be defined on $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, P_{\theta|T}^{\otimes n})$ as in **Model** §08101.08, where $\psi \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}})$ satisfies **(nieMM1)**–**(nieMM3)** for some $\mathfrak{v}_{\theta|T|\psi} \in \mathbb{R}_{\geq 1}$, $T_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\mathbb{R}}(\ell_2)$ or $T_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\mathbb{R}}(\ell_2)$, $\theta \in \ell_2$, and hence $g_\cdot = T_{\cdot, \cdot} \theta \in \text{dom}(T_{\cdot, \cdot}^\dagger) \subseteq \ell_2$. Given **Assumption** §08101.12 the (infeasible, generalised) GE $\widehat{\theta}_\cdot^{m_n^\circ} = T_{\cdot, \cdot}^{m_n^\circ|\dagger} \widehat{g}_\cdot \in \ell_2 \mathbb{1}_\cdot^{m_n^\circ} \subseteq \ell_2(\mathbf{v}^2)$ $P_{\theta|T}^{\otimes n}$ -a.s. with oracle dimension m_n° as in (08.03) satisfies

$$P_{\theta|T}^n(\|\widehat{\theta}_\cdot^{m_n^\circ} - \theta\|_{\mathbf{v}}^2) \leq \mathfrak{v}_{\theta|T|\psi} R_n^\circ(\theta, T_{\cdot, \cdot}, \mathbf{v}) \leq \mathfrak{v}_{\theta|T|\psi}^2 \inf_{m \in \mathbb{N}} P_{\theta|T}^n(\|\widehat{\theta}_\cdot^m - \theta\|_{\mathbf{v}}^2) \quad \forall n \in \mathbb{N},$$

and hence it is *oracle optimal* (with constant $\mathfrak{v}_{\theta|T|\psi}$).

§08101.26 **Proof of Corollary** §08101.25. Given in the lecture. \square

§08101.27 **Illustration.** We distinguish the following two cases

- (p) $\sup \{\|M_{\mathbf{v}} T_{\cdot, \cdot}^{m|\dagger}\|_{\text{HS}}^2 : m \in \mathbb{N}\} \in \mathbb{R}_{>0}$ or $\sup \{\|\theta^m - \theta\|_{\mathbf{v}}^2 : m \in \mathbb{N}_{>K}\} = 0$ for some $K \in \mathbb{N}$,
- (np) $\sup \{\|M_{\mathbf{v}} T_{\cdot, \cdot}^{m|\dagger}\|_{\text{HS}}^2 : m \in \mathbb{N}\} = \infty$ and $\sup \{\|\theta^m - \theta\|_{\mathbf{v}}^2 : m \in \mathbb{N}_{>K}\} \in \mathbb{R}_{>0}$ for all $K \in \mathbb{N}$.

Note that $\theta \mathbb{1}_\cdot^{K|\perp} = 0$ implies the case **(p)**. Interestingly, in case **(p)** the oracle bound is parametric, that is, $n R_n^\circ(\theta, T_{\cdot, \cdot}, \mathbf{v}) = O(1)$, in case **(np)** the oracle bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} n R_n^\circ(\theta, T_{\cdot, \cdot}, \mathbf{v}) = \infty$. In case **(np)** we consider similar to **(o-m)**, **(o-s)** and **(s-m)** in **Illustration** §07101.28 the following specifications:

Table 01 [§08]

Order of the oracle rate $R_n^{\circ}(\theta, T_{\bullet}, \mathbf{v})$ as $n \rightarrow \infty$

	(squared bias)	(variance)	m_n°	$R_n^{\circ}(\theta, T_{\bullet}, \mathbf{v})$	
$(m \in \mathbb{N})$	$\ \theta^m - \theta\ _{\mathbf{v}}^2$	$\ M_{\mathbf{v}} T_{\bullet}^{m \dagger}\ _{\text{HS}}^2$			
$(\mathbf{v}_m = m^{\mathbf{v}})$	$(a \in \mathbb{R}_{>0})$	$(t \in \mathbb{R}_{>0})$			
(o-m)	$v \in (-1/2 - t, a)$	$m^{-2(a-v)}$	$m^{2(t+v)+1}$	$n^{\frac{1}{2a+2t+1}}$	$n^{-\frac{2(a-v)}{2a+2t+1}}$
	$v+t = -1/2$	$m^{-2a-2t-1}$	$\log m$	$\left(\frac{n}{\log n}\right)^{\frac{1}{2a+2t+1}}$	$\frac{\log n}{n}$
(o-s)	$a-v \in \mathbb{R}_{>0}$	$m^{-2(a-v)}$	$m^{(1-2(t-v))_+} e^{m^{2a}}$	$(\log n)^{\frac{1}{2t}}$	$(\log n)^{-\frac{a-v}{t}}$
(s-m)	$v+t+1/2 \in \mathbb{R}_{>0}$	$m^{(1-2(a-v))_+} e^{-m^{2a}}$	$m^{2(t+v)+1}$	$(\log n)^{\frac{1}{2a}}$	$\frac{(\log n)^{\frac{2t+2v+1}{2a}}}{n}$
	$v+t = -1/2$	$e^{-m^{2a}}$	$\log m$	$(\log n)^{\frac{1}{2a}}$	$\frac{\log \log n}{n}$

We note that in case **(o-m)** and **(s-m)** for $v+t < -1/2$ the oracle rate $R_n^{\circ}(\theta, T_{\bullet}, \mathbf{v})$ is parametric. \square

§08|01|02|02 Maximal global \mathbf{v} -risk

§08|01.28 **Notation (Reminder).** For sequences $a_n, b_n \in (\mathbb{K})^{\mathbb{N}}$ taking its values in $\mathbb{K} \in \{\mathbb{R}, \mathbb{R}_{>0}, \mathbb{Z}, \dots\}$ we write $a_n \in (\mathbb{K})_{\nearrow}^{\mathbb{N}}$ and $b_n \in (\mathbb{K})_{\searrow}^{\mathbb{N}}$ if a_n and b_n , respectively, is monotonically *non-decreasing* and *non-increasing*. If in addition $a_n \rightarrow \infty$ and $b_n \rightarrow 0$ as $n \rightarrow \infty$, then we write $a_n \in (\mathbb{K})_{\nearrow\infty}^{\mathbb{N}}$ and $b_n \in (\mathbb{K})_{\searrow 0}^{\mathbb{N}}$ for short. For $w_n \in \ell_{\infty}$ we set $w_{(0)} := \|w_n\|_{\ell_{\infty}}$ and $w_{(\bullet)} = (w_{(j)} := \|w_n \mathbf{1}_n^{j \perp}\|_{\ell_{\infty}})_{j \in \mathbb{N}}$, where by construction $w_{(\bullet)} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$. \square

§08|01.29 **Assumption.** Consider weights $t_n, \alpha_n \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and $\mathbf{v}_n \in \mathbb{R}_{>0}^{\mathbb{N}}$ such that $(\alpha \mathbf{v})_n := \alpha_n \mathbf{v}_n \in \ell_{\infty}$, $(\alpha \mathbf{v})_{(\bullet)} \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$, and $(t/\mathbf{v})_n = t_n \mathbf{v}_n^{-1} \in \ell_{\infty}$ are satisfied. In addition there exists $C_{(t/\mathbf{v})} \in (0, 1]$ such that for all $m \in \mathbb{N}$

$$(t/\mathbf{v})_{(m-1)}^2 \geq \min \left\{ (t/\mathbf{v})_j^2 : j \in \llbracket m \rrbracket \right\} \geq C_{(t/\mathbf{v})} (t/\mathbf{v})_{(m)}^2 \quad (08.05)$$

or in equal $C_{(t/\mathbf{v})} \|(t/\mathbf{v})_{\bullet}^{-2} \mathbf{1}_m\|_{\ell_{\infty}} \leq (t/\mathbf{v})_{(m)}^{-2}$. \square

§08|01.30 **Reminder.** Under Assumption §08|01.29 we have $\ell_2^{\alpha} = \text{dom}(M_{\alpha^{-1}}) = \ell_2 \alpha \subseteq \ell_2$ and the three measures $\nu_N, \alpha^{-2} \nu_N$ and $\mathbf{v}^2 \nu_N$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04|01.02). We consider ℓ_2^{α} endowed with $\|\cdot\|_{\alpha^{-1}} = \|M_{\alpha^{-1}}\|_{\ell_2}$ and given a constant $r \in \mathbb{R}_{>0}$ the ellipsoid $\ell_2^{\alpha, r} := \{a \in \ell_2^{\alpha} : \|a\|_{\alpha^{-1}} \leq r\} \subseteq \ell_2^{\alpha}$. Since $(\alpha \mathbf{v})_n \in \ell_{\infty}$, and hence $(\alpha \mathbf{v})_{(m)} := \|(\alpha \mathbf{v})_{\bullet} \mathbf{1}_m^{\perp}\|_{\ell_{\infty}} \in \mathbb{R}_{>0}$ for each $m \in \mathbb{N}$ we have $\ell_2^{\alpha} \subseteq \ell_2(\mathbf{v}^2)$ (**Property** §04|02.11). Consequently, if Assumption §08|01.29 and $\theta \in \ell_2^{\alpha, r}$ are satisfied, then Assumption §08|01.12 is also fulfilled. Since $\mathbf{v}_n, t_n \in \mathbb{R}_{>0}^{\mathbb{N}}$ under Assumption §08|01.29, we have $\|t_n^{-1} \mathbf{1}_m\|_{\mathbf{v}} = \|(t/\mathbf{v})_{\bullet} \mathbf{1}_m\|_{\ell_2} \in \mathbb{R}_{>0}$ for each $m \in \mathbb{N}$. Under the Assumptions §08|00.02 and §08|01.29 considering the generalised link condition $T_{\bullet} \in \mathbb{T}_{t, d, D}$ with band $D \in [1, \infty)$ and $d \in [1, D]$ as in **Definition** §05|02.05 we have $\sup_{m \in \mathbb{N}} \{ \|[M_{\mathbf{t}}]_{\mathbf{m}} [T_{\bullet}]_{\mathbf{m}}^{-1}\|_{\text{spec}} \} \leq D$, and hence

$$\begin{aligned} \|M_{\mathbf{v}} T_{\bullet}^{m \dagger}\|_{\text{HS}}^2 &= \text{tr}(M_{\mathbf{v}}^m T_{\bullet}^{m \dagger} (T_{\bullet}^{m \dagger})^* M_{\mathbf{v}}^m) = \text{tr}([M_{\mathbf{v}}]_{\mathbf{m}} [T_{\bullet}]_{\mathbf{m}}^{-1} ([T_{\bullet}]_{\mathbf{m}}^{-1})^* [M_{\mathbf{v}}]_{\mathbf{m}}) \\ &= \text{tr}([M_{(v/t)}]_{\mathbf{m}} [M_{\mathbf{t}}]_{\mathbf{m}} [T_{\bullet}]_{\mathbf{m}}^{-1} ([T_{\bullet}]_{\mathbf{m}}^{-1})^* [M_{\mathbf{t}}]_{\mathbf{m}} [M_{(v/t)}]_{\mathbf{m}}) \leq \|[M_{\mathbf{t}}]_{\mathbf{m}} [T_{\bullet}]_{\mathbf{m}}^{-1}\|_{\text{spec}}^2 \text{tr}([M_{(v/t)}]_{\mathbf{m}}^2) \\ &\leq D^2 \|t^{-1} \mathbf{1}_m\|_{\mathbf{v}}^2 \quad (08.06) \end{aligned}$$

using $\text{tr}([M_{(v/t)}]_{\mathbf{m}}^2) = \|(v/t)_{\bullet} \mathbf{1}_m\|_{\ell_2}^2 = \|t^{-1} \mathbf{1}_m\|_{\mathbf{v}}^2$. Moreover, for each $m \in \mathbb{N}$ the generalised

Galerkin solution $\theta^m := \mathbb{T}_{\cdot, \cdot}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ of $\theta = \mathbb{T}_{\cdot, \cdot}^\dagger g \in \ell_2^{\text{a.r}}$ satisfies (Lemma §05102.09)

$$\|\theta - \theta^m\|_{\mathbb{V}}^2 \leq (D^2 d^2 C_{(v/v)}^{-2} + 1)(\mathbf{a}\mathbf{v})_{(m)}^2 r^2.$$

Note that under Assumptions §08100.02 and §08101.29 the link condition $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d}^{\geq}$ with band $d \in \mathbb{R}_{\geq 1}$ as in Definition §05101.08 implies $\sup_{m \in \mathbb{N}} \{ \|\mathbb{M}_{t, \cdot} [\mathbb{T}_{\cdot, \cdot}]^{-1}\|_{\text{spec}} \} \leq 3d^2$ (Lemma §05101.22), and hence for each $m \in \mathbb{N}$ we have (08.06) with $D = 3d^2$ and the Galerkin solution $\theta^m := \mathbb{T}_{\cdot, \cdot}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ of $\theta = \mathbb{T}_{\cdot, \cdot}^\dagger g \in \ell_2^{\text{a.r}}$ satisfies $\|\theta - \theta^m\|_{\mathbb{V}}^2 \leq (9d^6 C_{(v/v)}^{-2} + 1)(\mathbf{a}\mathbf{v})_{(m)}^2 r^2$ (Lemma §05101.28). \square

§08101.31 **Proposition.** Under Assumptions §08101.01 and §08101.29 setting for $n, m \in \mathbb{N}$

$$\begin{aligned} R_n^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) &:= [(\mathbf{a}\mathbf{v})_{(m)}^2 \vee n^{-1} \|\mathbb{T}_{\cdot, \cdot}^{-1} \mathbb{1}^m\|_{\mathbb{V}}^2], \quad m_n^* := \arg \min \{ R_n^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) : m \in \mathbb{N} \} \\ \text{and } R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) &:= R_n^{m_n^*}(\mathbf{a}, \mathbf{t}, \mathbf{v}) = \min \{ R_n^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) : m \in \mathbb{N} \} \end{aligned} \quad (08.07)$$

and $\|\Gamma_{\theta \mid \mathbb{T}}\|_{\mathbb{L}(\ell_2)} =: \mathbf{v}_{\theta \mid \mathbb{T}} \in \mathbb{R}_{\geq 0}$, for $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d,D}$ and for all $\theta \in \ell_2^{\text{a.r}}$, hence $g = \mathbb{T}_{\cdot, \cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^\dagger) \subseteq \ell_2$, we have

$$\mathbb{E}_{\theta \mid \mathbb{T}}^n (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbb{V}}^2) \leq (D^2 \mathbf{v}_{\theta \mid \mathbb{T}} + 2C_{(v/v)}^{-2} D^2 d^2 r^2) R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \quad \forall n \in \mathbb{N}$$

(or for $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d}^{\geq}$ with $D = 3d^2$).

§08101.32 **Proof of Proposition §08101.31.** Given in the lecture. \square

§08101.33 **Remark.** Under the assumptions of Proposition §08101.31 if there exists in addition $\mathbf{v} \in \mathbb{R}_{>0}$ satisfying $\|\Gamma_{\theta \mid \mathbb{T}}\|_{\mathbb{L}(\ell_2)} \leq \mathbf{v}$ for all $\theta \in \ell_2^{\text{a.r}}$ and $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d,D}$ (or $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d}^{\geq}$), then we have

$$\sup \{ \mathbb{E}_{\theta \mid \mathbb{T}}^n (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbb{V}}^2) : \theta \in \ell_2^{\text{a.r}}, \mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d,D} \} \leq (D^2 \mathbf{v} + 2C_{(v/v)}^{-2} D^2 d^2 r^2) R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \quad \forall n \in \mathbb{N}.$$

Arguing similarly as in Remark §07101.21 we note that $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) = o(1)$ as $n \rightarrow \infty$ since $\|\mathbb{T}_{\cdot, \cdot}^{-1} \mathbb{1}^m\|_{\mathbb{V}} \in \mathbb{R}_{>0}$ for all $m \in \mathbb{N}$, and $(\mathbf{a}\mathbf{v})_{(m)} = o(1)$ as $m \rightarrow \infty$ by Assumption §08101.29. The latter is satisfied, for example, if $(\mathbf{a}\mathbf{v})_{\cdot} \in \ell_2$ (in equal $\mathbf{a}_{\cdot} \in \ell_2(\mathbf{v}^2)$). Note that the dimension $m_n^* := m_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ as defined in (08.07) does not depend on the unknown parameter of interest θ but on the classes $\ell_2^{\text{a.r}}$ and $\mathbb{T}_{t,d,D}$ only, and thus also the statistic $\widehat{\theta}^{m_n^*}$. In other words, if the regularity of θ and $\mathbb{T}_{\cdot, \cdot}$ is known in advance, then the (generalised) GE $\widehat{\theta}^{m_n^*}$ is a feasible estimator. \square

§08101.34 **Corollary** (GniSM §08101.04 continued). Consider $\widehat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta \mid \mathbb{T}}^n$ as in Model §08101.04, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\geq}(\ell_2)$ or $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\mathbb{R}}(\ell_2)$, $\theta \in \ell_2$ and hence $g = \mathbb{T}_{\cdot, \cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^\dagger) \subseteq \ell_2$. Under Assumption §08101.29 the (generalised) GE $\widehat{\theta}^{m_n^*} = \mathbb{T}_{\cdot, \cdot}^{m_n^* \dagger} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^*} \subseteq \ell_2(\mathbf{v}^2)$ with dimension m_n^* as in (08.07) satisfies

$$\sup \{ N_{\theta \mid \mathbb{T}}^n (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbb{V}}^2) : \theta \in \ell_2^{\text{a.r}}, \mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d,D} \} \leq C R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \quad \forall n \in \mathbb{N} \quad (08.08)$$

with constant $C = D^2 + 2C_{(v/v)}^{-2} D^2 d^2 r^2$ (for $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d}^{\geq}$ with $D = 3d^2$).

§08101.35 **Proof of Corollary §08101.34.** Given in the lecture. \square

§08101.36 **Corollary** (niSM §08101.06 continued). Consider $\widehat{g} = g + n^{-1/2} \dot{\epsilon} \sim P_{\theta \mid \mathbb{T} \mid \sigma}^n$ as in Model §08101.06, where $\dot{\epsilon} \sim \otimes_{j \in \mathbb{N}} P_{(0,\sigma^2)}$ satisfies (iSM1) with $\|\sigma^2\|_{\ell_\infty} =: \mathbf{v}_\sigma \in \mathbb{R}_{>0}$, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\geq}(\ell_2)$ or $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\mathbb{R}}(\ell_2)$, $\theta \in \ell_2$ and hence $g = \mathbb{T}_{\cdot, \cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^\dagger) \subseteq \ell_2$. Under Assumption §08101.29 the (generalised) GE $\widehat{\theta}^{m_n^*} = \mathbb{T}_{\cdot, \cdot}^{m_n^* \dagger} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^*} \subseteq \ell_2(\mathbf{v}^2)$ with dimension m_n^* as in (08.07) satisfies

$$\sup \{ P_{\theta \mid \mathbb{T} \mid \sigma}^n (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathbb{V}}^2) : \theta \in \ell_2^{\text{a.r}}, \mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d,D} \} \leq C R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \quad \forall n \in \mathbb{N}$$

with constant $C = D^2 \mathbf{v}_\sigma + 2C_{(v/v)}^{-2} D^2 d^2 r^2$ (for $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t,d}^{\geq}$ with $D = 3d^2$).

§08101.37 **Proof of Corollary** §08101.36. Given in the lecture. □

§08101.38 **Corollary** (nieMM §08101.08 continued). Let $\widehat{g} = g + n^{-1/2}\dot{\epsilon}$ be defined on $(\mathcal{Z}^n, \mathcal{L}^{\otimes n}, \mathbb{P}_{\theta|\mathbb{T}}^{\otimes n})$ as in Model §08101.08, where $\psi \in \mathcal{M}(\mathcal{X} \otimes 2^{\mathbb{N}})$ satisfies (nieMM1) and (nieMM2) for some $\nu_{\theta|\mathbb{T}|\psi} \in \mathbb{R}_{\geq 1}$, $\mathbb{T}_{\bullet} \in \mathbb{L}_{\bullet}^{\geq}(\ell_2)$ or $\mathbb{T}_{\bullet} \in \mathbb{L}_{\bullet}^{\mathbb{R}}(\ell_2)$, $\theta \in \ell_2$ and hence $g = \mathbb{T}_{\bullet}\theta \in \text{dom}(\mathbb{T}_{\bullet}^{\dagger}) \subseteq \ell_2$. Under Assumption §08101.29 the (generalised) GE $\widehat{\theta}^{m_n^*} = \mathbb{T}_{\bullet}^{m_n^* \dagger} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^*} \subseteq \ell_2(\mathfrak{v}^2)$ with dimension m_n^* as in (08.07) satisfies

$$\sup \{ \mathbb{P}_{\theta|\mathfrak{s}}^{\otimes n} (\|\widehat{\theta}^{m_n^*} - \theta\|_{\mathfrak{v}}^2) : \theta \in \ell_2^{\text{a,r}}, \mathbb{T}_{\bullet} \in \mathbb{T}_{\text{t,d,D}} \} \leq C_{\text{a,r,t,d,D}} R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) \quad \forall n \in \mathbb{N}$$

with constant $C_{\text{a,r,t,d,D}} = D^2 \sup \{ \nu_{\theta|\mathbb{T}|\psi} : \theta \in \ell_2^{\text{a,r}}, \mathbb{T}_{\bullet} \in \mathbb{T}_{\text{t,d,D}} \} + 2C_{(\text{t/v})}^{-2} D^2 d^2 r^2$ (for $\mathbb{T}_{\bullet} \in \mathbb{T}_{\text{t,d}}^{\geq}$ with $D = 3d^2$).

§08101.39 **Proof of Corollary** §08101.38. Given in the lecture. □

§08101.40 **Illustration.** We distinguish the following two cases **(p)** $(\mathfrak{v}/\mathfrak{t}) \in \ell_2$, and **(np)** $(\mathfrak{v}/\mathfrak{t}) \notin \ell_2$. Interestingly, in case **(p)** the bound in Proposition §08101.31 is parametric, that is, $nR_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) = O(1)$, in case **(np)** the bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) = \infty$. In case **(np)** we consider similar to **(o-m)**, **(o-s)** and **(s-m)** in Illustration §07101.44 the following three specifications:

Table 02 [§08]

Order of the rate $R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ as $n \rightarrow \infty$

$(j \in \mathbb{N})$ $\mathfrak{v}_j^2 = j^{2\nu}$	$(\mathfrak{a} \in \mathbb{R}_{>0})$ $(\mathfrak{t} \in \mathbb{R}_{>0})$ \mathfrak{a}_j^2 \mathfrak{t}_j^2	(squared bias) $(\mathfrak{a}\mathfrak{v})_{(m)}^2$	(variance) $\ \mathfrak{t}_{\bullet}^{-1} \mathbb{1}_m^m\ _{\mathfrak{v}}^2$	m_n^*	$R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$
(o-m) $\nu \in (-1/2 - \mathfrak{t}, \mathfrak{a})$ $\nu + \mathfrak{t} = -1/2$	$j^{-2\mathfrak{a}}$ $j^{-2\mathfrak{t}}$ $j^{-2\mathfrak{a}}$ $j^{-2\mathfrak{t}}$	$m^{-2(\mathfrak{a}-\nu)}$ $m^{-2\mathfrak{a}-2\mathfrak{t}-1}$	$m^{2\nu+2\mathfrak{t}+1}$ $\log m$	$n^{\frac{1}{2\mathfrak{a}+2\mathfrak{t}+1}}$ $(\frac{n}{\log n})^{\frac{1}{2\mathfrak{a}+2\mathfrak{t}+1}}$	$n^{-\frac{2(\mathfrak{a}-\nu)}{2\mathfrak{a}+2\mathfrak{t}+1}}$ $\frac{\log n}{n}$
(o-s) $\mathfrak{a} - \nu \in \mathbb{R}_{>0}$	$j^{-2\mathfrak{a}}$ $e^{-j^{2\mathfrak{t}}}$	$m^{-2(\mathfrak{a}-\nu)}$	$m^{(1-2(\mathfrak{t}-\nu))_+} e^{m^{2\mathfrak{t}}}$	$(\log n)^{\frac{1}{2\mathfrak{t}}}$	$(\log n)^{-\frac{\mathfrak{a}-\nu}{\mathfrak{t}}}$
(s-m) $\nu + \mathfrak{t} + 1/2 \in \mathbb{R}_{>0}$ $\nu + \mathfrak{t} = -1/2$	$e^{-j^{2\mathfrak{a}}}$ $j^{-2\mathfrak{t}}$ $e^{-j^{2\mathfrak{a}}}$ $j^{-2\mathfrak{t}}$	$m^{2\nu} e^{-m^{2\mathfrak{a}}}$ $m^{2\nu} e^{-m^{2\mathfrak{a}}}$	$m^{2\nu+2\mathfrak{t}+1}$ $\log m$	$(\log n)^{\frac{1}{2\mathfrak{a}}}$ $(\log n)^{\frac{1}{2\mathfrak{a}}}$	$\frac{(\log n)^{\frac{2\mathfrak{t}+2\nu+1}{2\mathfrak{a}}}}{n}$ $\frac{\log \log n}{n}$

We note that in case **(o-m)** and **(s-m)** for $\nu + \mathfrak{t} < -1/2$ the rate $R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ is parametric. □

§08|01|03 Local and maximal local ϕ -risk

Secondly, we measure the accuracy of the (generalised) GE $\widehat{\theta}^m := \mathbb{T}_{\bullet}^{m \dagger} \widehat{g}$ of the (generalised) Galerkin solution $\theta^m = \mathbb{T}_{\bullet}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ with $g = \mathbb{T}_{\bullet}\theta \in \text{dom}(\mathbb{T}_{\bullet}^{\dagger})$ by the mean of its local ϕ -error introduced in §0510102 and §05102102, i.e. its ϕ -risk.

§08101.41 **Reminder.** If $\phi \in \mathbb{R}_{\mathfrak{N}0}^{\mathbb{N}}$ then we have $\phi^2 \mathbb{1}^m \in \ell_2$ and $\ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi_{\mathfrak{N}})$. Consequently, for each $\theta \in \text{dom}(\phi_{\mathfrak{N}})$ the (generalised) Galerkin solution $\theta^m = \mathbb{T}_{\bullet}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ satisfies $\theta^m \in \text{dom}(\phi_{\mathfrak{N}})$ too. If in addition $C_{\mathbb{T}} := \sup \{ \|\mathbb{M}_{\mathbb{T}^{\perp}} \mathbb{T}_{\bullet}^* (\mathbb{T}_{\bullet}^{m \dagger})^* \phi\|_{\ell_2} : m \in \mathbb{N} \} \in \mathbb{R}_{>0}$ then $|\phi_{\mathfrak{N}}(\theta^m - \theta)| \leq (1 + C_{\mathbb{T}}) \|\mathbb{1}^{m \perp} \theta\|_{\ell_2}$ which implies $\sup \{ |\phi_{\mathfrak{N}}(\theta^j - \theta)| : j \in \mathbb{N}_{\geq m} \} = o(1)$ as $m \rightarrow \infty$ (**Property** §05101.31 and **Property** §05102.12). □

§08101.42 **Comment.** Under Assumption §08101.01 since $\theta^m, \mathbb{T}_{\bullet}^{m \dagger} \mathbb{1}^m \in \ell_2 \mathbb{1}^m$ for each $m \in \mathbb{N}$ we have $\mathbb{T}_{\bullet}^{m \dagger} \dot{\epsilon} \in \ell_2 \mathbb{1}^m$ $\mathbb{P}_{\theta|\mathbb{T}}^n$ -a.s.. Indeed, $\dot{\epsilon} \sim \mathbb{P}_{(\mathfrak{u}, \mathfrak{I}_{\theta|\mathbb{T}})}$ with $\mathfrak{I}_{\theta|\mathbb{T}} \in \mathbb{L}^{\geq}(\ell_2)$ by Assumption §08101.01 (**nSIP**) implies $\mathbb{P}_{\theta|\mathbb{T}}^n(\dot{\epsilon}^2) \in \ell_{\infty}$, hence $\dot{\epsilon} \mathbb{1}^m \in \ell_{\infty}$ $\mathbb{P}_{\mathfrak{q}}^n$ -a.s. and $\|\mathbb{T}_{\bullet}^{m \dagger} \dot{\epsilon}\|_{\ell_2} \leq \|\mathbb{T}_{\bullet}^{m \dagger} \mathbb{1}^m\|_{\ell_2} \|\dot{\epsilon} \mathbb{1}^m\|_{\ell_{\infty}} \in \mathbb{R}_{>0}$ $\mathbb{P}_{\theta|\mathbb{T}}^n$ -a.s..

Given $\phi \in \mathbb{R}_{\lambda_0}^{\mathbb{N}}$ from $\ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi_{\mathcal{N}})$ (**Reminder** §08101.41) it follows

$$\widehat{\theta}^m = \mathbb{T}_{\bullet, \bullet}^{m \dagger} \widehat{g} = n^{-1/2} \mathbb{T}_{\bullet, \bullet}^{m \dagger} \dot{\epsilon} + \theta^m \in \ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi_{\mathcal{N}}) \quad \mathbb{P}_{\theta_{\Gamma}^n}^n \text{-a.s.} \quad \square$$

§08101|03101 Local ϕ -risk

§08101.43 **Assumption.** Let $\phi \in \mathbb{R}_{\lambda_0}^{\mathbb{N}}$ and $\theta \in \text{dom}(\phi_{\mathcal{N}})$ be satisfied. □

§08101.44 **Definition.** Under Assumptions §08101.01 and §08101.43 the *local ϕ -risk* of a (generalised) GE $\widehat{\theta}^m = \mathbb{T}_{\bullet, \bullet}^{m \dagger} \widehat{g} \in \ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi_{\mathcal{N}})$ $\mathbb{P}_{\theta_{\Gamma}^n}^n$ -a.s. satisfies

$$\mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\widehat{\theta}^m - \theta)|^2) = \mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\mathbb{T}_{\bullet, \bullet}^{m \dagger}(\widehat{g} - g))|^2) + |\phi_{\mathcal{N}}(\theta^m - \theta)|^2 \quad (08.09)$$

with *variance* $\mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\mathbb{T}_{\bullet, \bullet}^{m \dagger}(\widehat{g} - g))|^2) = n^{-1} \mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\mathbb{T}_{\bullet, \bullet}^{m \dagger} \dot{\epsilon})|^2)$ and *bias* $|\phi_{\mathcal{N}}(\theta^m - \theta)|$. □

§08101.45 **Property.** Under Assumptions §08101.01 and §08101.43 we have

$$\begin{aligned} \mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\mathbb{T}_{\bullet, \bullet}^{m \dagger} \dot{\epsilon})|^2) &= \mathbb{E}_{\theta_{\Gamma}^n}^n(|\langle \phi \mathbb{1}^m, \mathbb{T}_{\bullet, \bullet}^{m \dagger} \dot{\epsilon} \rangle_{\ell_2}|^2) = \mathbb{E}_{\theta_{\Gamma}^n}^n(|\langle (\mathbb{T}_{\bullet, \bullet}^{m \dagger})^* \phi^m, \dot{\epsilon} \rangle_{\ell_2}|^2) \\ &= \langle \Gamma_{\theta_{\Gamma}^n}(\mathbb{T}_{\bullet, \bullet}^{m \dagger})^* \phi^m, (\mathbb{T}_{\bullet, \bullet}^{m \dagger})^* \phi^m \rangle_{\ell_2} = \|(\mathbb{T}_{\bullet, \bullet}^{m \dagger})^* \phi^m\|_{\Gamma_{\theta_{\Gamma}^n}}^2 \end{aligned} \quad (08.10)$$

and consequently $\mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\mathbb{T}_{\bullet, \bullet}^{m \dagger}(\widehat{g} - g))|^2) \leq n^{-1} \|\Gamma_{\theta_{\Gamma}^n}\|_{\mathbb{L}(\ell_2)} \|(\mathbb{T}_{\bullet, \bullet}^{m \dagger})^* \phi^m\|_{\ell_2}^2 \in \mathbb{R}_{\geq 0}$. □

§08101.46 **Proposition (Upper bound).** Under Assumptions §08101.01 and §08101.43 for all $n, m \in \mathbb{N}$ with (generalised) Galerkin solution $\theta^m = \mathbb{T}_{\bullet, \bullet}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ setting

$$\begin{aligned} \mathbb{R}_n^m(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) &:= |\phi_{\mathcal{N}}(\theta^m - \theta)|^2 + n^{-1} \|(\mathbb{T}_{\bullet, \bullet}^{m \dagger})^* \phi^m\|_{\ell_2}^2, \quad m_n^{\circ} := \arg \min \{ \mathbb{R}_n^m(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) : m \in \mathbb{N} \} \\ \text{and } \mathbb{R}_n^{\circ}(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) &:= \mathbb{R}_n^{m_n^{\circ}}(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) = \min \{ \mathbb{R}_n^m(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) : m \in \mathbb{N} \} \end{aligned} \quad (08.11)$$

we have $\mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\widehat{\theta}^{m_n^{\circ}} - \theta)|^2) \leq (1 \vee \|\Gamma_{\theta_{\Gamma}^n}\|_{\mathbb{L}(\ell_2)}) \mathbb{R}_n^{\circ}(\theta, \mathbb{T}_{\bullet, \bullet}, \phi)$ for all $n \in \mathbb{N}$.

§08101.47 **Proof of Proposition** §08101.46. Given in the lecture. □

§08101.48 **Reminder.** If $\Gamma_{\theta_{\Gamma}^n} \in \mathbb{L}^{\geq}(\ell_2)$ is invertible with inverse $\Gamma_{\theta_{\Gamma}^n}^{-1} \in \mathbb{L}(\ell_2)$, i.e. $\Gamma_{\theta_{\Gamma}^n} \Gamma_{\theta_{\Gamma}^n}^{-1} = \text{id}_{\ell_2} = \Gamma_{\theta_{\Gamma}^n}^{-1} \Gamma_{\theta_{\Gamma}^n}$, then we write shortly $1 \leq \max(\|\Gamma_{\theta_{\Gamma}^n}\|_{\mathbb{L}(\ell_2)}, \|\Gamma_{\theta_{\Gamma}^n}^{-1}\|_{\mathbb{L}(\ell_2)}) \leq \mathfrak{v}_{\theta_{\Gamma}^n} \in \mathbb{R}_{\geq 1}$. In this situation for all $h_{\bullet} \in \ell_2$ we have $\mathfrak{v}_{\theta_{\Gamma}^n}^{-1} \|h_{\bullet}\|_{\ell_2}^2 \leq \|h_{\bullet}\|_{\Gamma_{\theta_{\Gamma}^n}}^2 = \langle \Gamma_{\theta_{\Gamma}^n} h_{\bullet}, h_{\bullet} \rangle_{\mathbb{J}} \leq \mathfrak{v}_{\theta_{\Gamma}^n} \|h_{\bullet}\|_{\ell_2}^2$. □

§08101.49 **Oracle inequality.** Under Assumptions §08101.01 and §08101.43 if in addition

$$1 \leq \max(\|\Gamma_{\theta_{\Gamma}^n}\|_{\mathbb{L}(\ell_2)}, \|\Gamma_{\theta_{\Gamma}^n}^{-1}\|_{\mathbb{L}(\ell_2)}) \leq \mathfrak{v}_{\theta_{\Gamma}^n} \in \mathbb{R}_{\geq 1}$$

is satisfied then (08.11) (and **Reminder** §08101.48) implies

$$\begin{aligned} \mathfrak{v}_{\theta_{\Gamma}^n}^{-1} \mathbb{R}_n^m(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) &\leq \mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\widehat{\theta}^m - \theta)|^2) = n^{-1} \|(\mathbb{T}_{\bullet, \bullet}^{m \dagger})^* \phi^m\|_{\Gamma_{\theta_{\Gamma}^n}}^2 + |\phi_{\mathcal{N}}(\theta^m - \theta)|^2 \\ &\leq \mathfrak{v}_{\theta_{\Gamma}^n} \mathbb{R}_n^m(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) \quad \forall n, m \in \mathbb{N}. \end{aligned}$$

As a consequence we immediately obtain the following *oracle inequality*

$$\begin{aligned} \mathfrak{v}_{\theta_{\Gamma}^n}^{-1} \mathbb{R}_n^{\circ}(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) &\leq \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\widehat{\theta}^m - \theta)|^2) \leq \mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\widehat{\theta}^{m_n^{\circ}} - \theta)|^2) \\ &\leq \mathfrak{v}_{\theta_{\Gamma}^n} \mathbb{R}_n^{\circ}(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) \leq \mathfrak{v}_{\theta_{\Gamma}^n}^2 \inf_{m \in \mathbb{N}} \mathbb{E}_{\theta_{\Gamma}^n}^n(|\phi_{\mathcal{N}}(\widehat{\theta}^m - \theta)|^2) \quad \forall n \in \mathbb{N}, \end{aligned} \quad (08.12)$$

and hence $\mathbb{R}_n^{\circ}(\theta, \mathbb{T}_{\bullet, \bullet}, \phi)$, m_n° and the statistic $\widehat{\theta}^{m_n^{\circ}}$, respectively, is an *oracle bound*, an *oracle dimension* and *oracle optimal* (up to the constant $\mathfrak{v}_{\theta_{\Gamma}^n}^2$). □

§08101.50 **Remark.** Arguing similarly as in **Remark** §07101.21 we note that $\|(\mathbb{T}_{\cdot, \cdot}^{m \dagger})^* \phi_{\cdot}^m\|_{\ell_2}^2 \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and hence $R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) = o(1)$ as $n \rightarrow \infty$, whenever $|\phi_{\mathbb{N}}(\theta^m - \theta)|^2 = o(1)$ as $m \rightarrow \infty$ (see **Reminder** §08101.41). Note that the oracle dimension $m_n^\circ := m_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \phi)$ as defined in (08.11) depends on the unknown parameter of interest θ , and thus also the oracle optimal statistic $\widehat{\theta}^{m_n^\circ}$. In other words $\widehat{\theta}^{m_n^\circ}$ is not a feasible estimator. \square

§08101.51 **Corollary** (GniSM §08101.04 continued). Consider $\widehat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta|T}^n$ as in *Model* §08101.04, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\geq}(\ell_2)$ or $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\leq}(\ell_2)$, $\theta \in \ell_2$, and hence $g = \mathbb{T}_{\cdot, \cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^{\dagger}) \subseteq \ell_2$. Given *Assumption* §08101.43 the (infeasible, generalised) GE $\widehat{\theta}^{m_n^\circ} = \mathbb{T}_{\cdot, \cdot}^{m_n^\circ \dagger} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^\circ} \subseteq \text{dom}(\phi_{\mathbb{N}})$ with oracle dimension m_n° as in (08.11) satisfies

$$N_{\theta|T}^n(|\phi_{\mathbb{N}}(\widehat{\theta}^{m_n^\circ} - \theta)|^2) = R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) = \inf_{m \in \mathbb{N}} N_{\theta|T}^n(|\phi_{\mathbb{N}}(\widehat{\theta}^m - \theta)|^2),$$

and hence it is **oracle optimal** (with constant 1).

§08101.52 **Proof of Corollary** §08101.51. Given in the lecture. \square

§08101.53 **Corollary** (niSM §08101.06 continued). Consider $\widehat{g} = g + n^{-1/2} \dot{\epsilon} \sim P_{\theta|T|\sigma}^n$ as in *Model* §08101.06, where $\dot{\epsilon} \sim \otimes_{j \in \mathbb{N}} P_{(0, \sigma_j^2)}$ satisfies (iSM1) and (iSM2) with $\max(\|\sigma_{\cdot}^{-2}\|_{\ell_\infty}, \|\sigma_{\cdot}^2\|_{\ell_\infty}) =: v_\sigma \in \mathbb{R}_{>1}$, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\geq}(\ell_2)$ or $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\leq}(\ell_2)$, $\theta \in \ell_2$, and hence $g = \mathbb{T}_{\cdot, \cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^{\dagger}) \subseteq \ell_2$. Given *Assumption* §08101.43 the (infeasible, generalised) GE $\widehat{\theta}^{m_n^\circ} = \mathbb{T}_{\cdot, \cdot}^{m_n^\circ \dagger} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^\circ} \subseteq \text{dom}(\phi_{\mathbb{N}})$ with oracle dimension m_n° as in (08.11) satisfies

$$P_{\theta|T|\sigma}^n(|\phi_{\mathbb{N}}(\widehat{\theta}^{m_n^\circ} - \theta)|^2) \leq v_\sigma R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) \leq v_\sigma^2 \inf_{m \in \mathbb{N}} P_{\theta|T|\sigma}^n(|\phi_{\mathbb{N}}(\widehat{\theta}^m - \theta)|^2),$$

and hence it is **oracle optimal** (with constant v_σ).

§08101.54 **Proof of Corollary** §08101.53. Given in the lecture. \square

§08101.55 **Corollary** (nieMM §08101.08 continued). Let $\widehat{g} = g + n^{-1/2} \dot{\epsilon} \cdot$ be defined on $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, P_{\theta|T}^{\otimes n})$ as in *Model* §08101.08, where $\psi \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}})$ satisfies (nieMM1)–(nieMM3) for some $v_{\theta|T|\psi} \in \mathbb{R}_{>1}$, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\geq}(\ell_2)$ or $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}_{\cdot}^{\leq}(\ell_2)$, $\theta \in \ell_2$, and hence $g = \mathbb{T}_{\cdot, \cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^{\dagger}) \subseteq \ell_2$. Given *Assumption* §08101.43 the (infeasible, generalised) GE $\widehat{\theta}^{m_n^\circ} = \mathbb{T}_{\cdot, \cdot}^{m_n^\circ \dagger} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^\circ} \subseteq \text{dom}(\phi_{\mathbb{N}})$ with oracle dimension m_n° as in (08.11) satisfies

$$P_{\theta|T}^n(|\phi_{\mathbb{N}}(\widehat{\theta}^{m_n^\circ} - \theta)|^2) \leq v_{\theta|T|\psi} R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \psi) \leq v_{\theta|T|\psi}^2 \inf_{m \in \mathbb{N}} P_{\theta|T}^n(|\phi_{\mathbb{N}}(\widehat{\theta}^m - \theta)|^2),$$

and hence it is **oracle optimal** (with constant $v_{\theta|T|\psi}$).

§08101.56 **Proof of Corollary** §08101.55. Given in the lecture. \square

§08101.57 **Illustration.** We distinguish the following two cases

- (p) $\sup \{ \|(\mathbb{T}_{\cdot, \cdot}^{m \dagger})^* \phi_{\cdot}^m\|_{\ell_2}^2 : m \in \mathbb{N} \} \in \mathbb{R}_{>0}$ or $\sup \{ |\phi_{\mathbb{N}}(\widehat{\theta}^m - \theta)|^2 : m \in \mathbb{N}_{\geq K} \} = 0$ for $K \in \mathbb{N}$,
- (np) $\sup \{ \|(\mathbb{T}_{\cdot, \cdot}^{m \dagger})^* \phi_{\cdot}^m\|_{\ell_2}^2 : m \in \mathbb{N} \} = \infty$ and $\sup \{ |\phi_{\mathbb{N}}(\widehat{\theta}^m - \theta)|^2 : m \in \mathbb{N}_{\geq K} \} \in \mathbb{R}_{>0}$ for all $K \in \mathbb{N}$.

Note that $\theta \mathbb{1}_{\cdot}^{K \perp} = 0$ implies the case (p). Interestingly, in case (p) the oracle bound is parametric, that is, $n R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) = O(1)$, in case (np) the oracle bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} n R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) = \infty$. In case (np) we consider similar to (o-m), (o-s) and (s-m) in **Illustration** §07101.63 the following specifications:

Table 03 [§08]

Order of the oracle rate $R_n^\circ(\theta, T_{\cdot, \cdot}, \phi)$ as $n \rightarrow \infty$

	(squared bias)	(variance)	m_n°	$R_n^\circ(\theta, T_{\cdot, \cdot}, \phi)$	
$(m \in \mathbb{N})$	$ \phi \nu_N(\hat{\theta}^m - \theta) ^2$	$\ (\mathbb{T}_{\cdot, \cdot}^{m \dagger})^* \phi^m\ _{\ell_2}^2$			
$(\phi_m = m^{v-1/2})$	$(a \in \mathbb{R}_{>0})$	$(t \in \mathbb{R}_{>0})$			
(o-m)	$v \in (-t, a)$	$m^{-2(a-v)}$	m^{2t+2v}	$n^{\frac{1}{2a+2t}}$	$n^{-\frac{a-v}{a+t}}$
	$v = -t$	$m^{-2(a+t)}$	$\log m$	$(\frac{n}{\log n})^{\frac{1}{2(a+t)}}$	$\frac{\log n}{n}$
(o-s)	$a - v \in \mathbb{R}_{>0}$	$m^{-2(a-v)}$	$m^{2(v-t)+e^{m^{2t}}}$	$(\log n)^{\frac{1}{2t}}$	$(\log n)^{-\frac{a-v}{t}}$
(s-m)	$v + t \in \mathbb{R}_{>0}$	$m^{(1-4a+2v)+e^{-m^{2a}}}$	m^{2t+v}	$(\log n)^{\frac{1}{2a}}$	$\frac{(\log n)^{\frac{t+v}{a}}}{n}$
	$v = -t$	$m^{(1-4a-2t)+e^{-m^{2a}}}$	$\log m$	$(\log n)^{\frac{1}{2a}}$	$\frac{\log \log n}{n}$

We note that in case **(o-m)** and **(s-m)** for $v < -t$ the oracle rate $R_n^\circ(\theta, T_{\cdot, \cdot}, \phi)$ is parametric. \square

§08|01|03|02 Maximal local ϕ -risk

§08101.58 **Assumption.** Consider weights $\mathfrak{t}, \mathfrak{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and $\phi \in \mathbb{R}_{\searrow 0}^{\mathbb{N}}$ such that $(\mathfrak{a}\phi)_\cdot := \mathfrak{a}_\cdot \phi_\cdot \in \ell_2$ and $(\mathfrak{a}\mathfrak{t})_\cdot := \mathfrak{a}_\cdot \mathfrak{t}_\cdot \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$. \square

§08101.59 **Comment.** Assuming $\mathfrak{t}, \mathfrak{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and hence $(\mathfrak{a}\mathfrak{t})_\cdot^2 \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ is rather weak. If in addition $\liminf_{j \rightarrow \infty} (\mathfrak{a}\mathfrak{t})_j^2 \geq c \in \mathbb{R}_{>0}$ is satisfied, and hence $(\mathfrak{a}\mathfrak{t})_\cdot^2, \mathfrak{a}_\cdot^2, \mathfrak{t}_\cdot^2 \notin (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$, then $\mathfrak{a}_\cdot^2 \notin (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ and the assumption $(\mathfrak{a}\phi)_\cdot \in \ell_2$ implies $\phi \in \ell_2$, which together with $\mathfrak{t}_\cdot^2 \notin (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ implies $(\phi/\mathfrak{t})_\cdot \in \ell_2$, and thus the rate $R_n^*(\mathfrak{a}, \mathfrak{t}, \phi)$ is parametric (**Illustration** §08101.72). Since we are interested in the case of a non-parametric rate, the additional assumption $(\mathfrak{a}\mathfrak{t})_\cdot^2 \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ imposes a rather weak condition satisfied also in **Illustration** §08101.72. \square

§08101.60 **Reminder.** Under Assumption §08101.58 we have $\ell_2^{\mathfrak{a}} = \text{dom}(M_{\mathfrak{a}^{-1}}) = \ell_2 \mathfrak{a} \subseteq \ell_2$ and the three measures $\nu_N, \mathfrak{a}_\cdot^{-2} \nu_N$ and $|\phi| \nu_N$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04101.02). We consider $\ell_2^{\mathfrak{a}}$ endowed with $\|\cdot\|_{\mathfrak{a}^{-1}} = \|M_{\mathfrak{a}^{-1}} \cdot\|_{\ell_2}$ and given a constant $r \in \mathbb{R}_{>0}$ the ellipsoid $\ell_2^{\mathfrak{a}, r} := \{a_\cdot \in \ell_2^{\mathfrak{a}} : \|a_\cdot\|_{\mathfrak{a}^{-1}} \leq r\} \subseteq \ell_2^{\mathfrak{a}}$. Since $(\mathfrak{a}\phi)_\cdot \in \ell_2$ we have $\ell_2^{\mathfrak{a}} \subseteq \text{dom}(\phi \nu_N)$ (**Property** §04102.23). Consequently, if Assumption §08101.58 and $\theta \in \ell_2^{\mathfrak{a}, r}$ are satisfied, then Assumption §08101.43 is also fulfilled. Moreover, from $(\mathfrak{a}\phi)_\cdot \in \ell_2$ follows $\|\mathfrak{a}_\cdot \mathbb{1}^{m \perp}\|_\phi = \|(\mathfrak{a}\phi)_\cdot \mathbb{1}^{m \perp}\|_{\ell_2} = o(1)$ as $m \rightarrow \infty$. For $s \in [0, 1]$ from $(\mathfrak{a}\mathfrak{t})_\cdot = \mathfrak{a}_\cdot \mathfrak{t}_\cdot \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ follows $(\mathfrak{a}\mathfrak{t})_\cdot^s = ((\mathfrak{a}\mathfrak{t})_{(m)}) := (\mathfrak{a}\mathfrak{t})_{m+1} = \|(\mathfrak{a}\mathfrak{t})_\cdot \mathbb{1}^{m \perp}\|_{\ell_\infty} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$. Since $\phi_\cdot, \mathfrak{t}_\cdot \in \mathbb{R}_{\searrow 0}^{\mathbb{N}}$ under Assumption §08101.58, we have $\ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi \nu_N)$ and $\|\mathfrak{t}_\cdot^{-1} \mathbb{1}^m\|_\phi = \|(\phi/\mathfrak{t})_\cdot \mathbb{1}^m\|_{\ell_2} \in \mathbb{R}_{\geq 0}$ for each $m \in \mathbb{N}$. Under the Assumptions §08100.02 and §08101.58 considering the generalised link condition $T_{\cdot, \cdot} \in \mathbb{T}_{t, d, D}$ with band $D \in \mathbb{R}_{\geq 1}$ and $d \in [1, D]$ as in **Definition** §05102.05 we have $\sup_{m \in \mathbb{N}} \{ \|([\mathbb{T}_{\cdot, \cdot}]_m^{-1})^* [M_{\mathfrak{t}}]_{\mathfrak{t}}\|_{\text{spec}} \} \leq D$, and hence

$$\begin{aligned} \|(\mathbb{T}_{\cdot, \cdot}^{m \dagger})^* \phi^m\|_{\ell_2} &= \|([\mathbb{T}_{\cdot, \cdot}]_m^{-1})^* [\phi]_{\mathfrak{t}}\| = \|([\mathbb{T}_{\cdot, \cdot}]_m^{-1})^* [M_{\mathfrak{t}}]_{\mathfrak{t}} [M_{\mathfrak{t}}]_m^{-1} [\phi]_{\mathfrak{t}}\| \\ &\leq \|([\mathbb{T}_{\cdot, \cdot}]_m^{-1})^* [M_{\mathfrak{t}}]_{\mathfrak{t}}\|_{\text{spec}} \| [M_{\mathfrak{t}}]_m^{-1} [\phi]_{\mathfrak{t}} \| \leq D \|\mathfrak{t}_\cdot^{-1} \mathbb{1}^m\|_\phi \end{aligned} \quad (08.13)$$

using $\| [M_{\mathfrak{t}}]_m^{-1} [\phi]_{\mathfrak{t}} \| = \|\mathfrak{t}_\cdot^{-1} \mathbb{1}^m\|_\phi$. Moreover, for each $m \in \mathbb{N}$ the generalised Galerkin solution $\theta^m := \mathbb{T}_{\cdot, \cdot}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ of $\theta = \mathbb{T}_{\cdot, \cdot}^\dagger g \in \ell_2^{\mathfrak{a}, r}$ satisfies (**Lemma** §05102.14)

$$|\phi \nu_N(\theta^m - \theta)|^2 \leq Dd(Dd + 1)r^2 (\|\mathfrak{a}_\cdot \mathbb{1}^{m \perp}\|_\phi^2 + (\mathfrak{a}\mathfrak{t})_{(m)}^2 \|\mathfrak{t}_\cdot^{-1} \mathbb{1}^m\|_\phi^2). \quad (08.14)$$

Under Assumptions §08100.02 and §08101.58 the link condition $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d}^{\geq}$ with band $d \in \mathbb{R}_{\geq 1}$ as in **Definition** §05101.08 implies $\sup_{m \in \mathbb{N}} \{ \| ([\mathbb{T}_{\cdot,\cdot}]_m^{-1})^* [\mathbb{M}_t]_{\underline{m}} \|_{\text{spec}} \} \leq 3d^2$ (**Lemma** §05101.22), and hence for each $m \in \mathbb{N}$ we have (08.13) with $D = 3d^2$ and the Galerkin solution $\theta^m := \mathbb{T}_{\cdot,\cdot}^{m \uparrow} g \in \ell_2 \mathbb{1}^m$ of $\theta = \mathbb{T}_{\cdot,\cdot}^{\uparrow} g \in \ell_2^{\text{a,r}}$ satisfies (08.14) with $D = 3d^2$ (**Lemma** §05101.34). \square

§08101.61 **Lemma.** Under Assumption §08101.58 setting for $n, m \in \mathbb{N}$

$$\begin{aligned} \mathbb{R}_n^m(\mathbf{a}, \mathbf{t}, \phi) &:= \|\mathbf{a} \mathbb{1}^{m \perp}\|_{\phi}^2 + n^{-1} \|\mathbf{t}^{-1} \mathbb{1}^m\|_{\phi}^2, \quad m_n^* := \arg \min \{ \mathbb{R}_n^m(\mathbf{a}, \mathbf{t}, \phi) : m \in \mathbb{N} \} \\ \text{and } \mathbb{R}_n^*(\mathbf{a}, \mathbf{t}, \phi) &:= \mathbb{R}_n^{m_n^*}(\mathbf{a}, \mathbf{t}, \phi) = \min \{ \mathbb{R}_n^m(\mathbf{a}, \mathbf{t}, \phi) : m \in \mathbb{N} \} \end{aligned} \quad (08.15)$$

we have $(\text{at})_{m_n^*}^2 > n^{-1} \geq (\text{at})_{m_n^*+1}^2 = (\text{at})_{(m_n^*)}^2$ for all $n \in \mathbb{N}_{>(\text{at})_2^{-2}}$, i.e. $(\text{at})_2^2 > n^{-1}$ is satisfied.

§08101.62 **Proof of Lemma** §08101.61. Given in the lecture. \square

§08101.63 **Proposition (Upper bound).** Under Assumptions §08101.01 and §08101.58 setting m_n^* and $\mathbb{R}_n^*(\mathbf{a}, \mathbf{t}, \phi)$ for $n \in \mathbb{N}$ as in (08.15) and $\|\Gamma_{\theta|\mathbb{T}}\|_{\mathbb{L}(\ell_2)} =: \mathbb{v}_{\theta|\mathbb{T}} \in \mathbb{R}_{>0}$, for $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D}$ and for all $\theta \in \ell_2^{\text{a,r}}$, hence $g = \mathbb{T}_{\cdot,\cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot,\cdot}^{\uparrow}) \subseteq \ell_2$, we have

$$\mathbb{P}_{\theta|\mathbb{T}}^n (|\phi \nu_{\mathbb{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) \leq D^2 (\mathbb{v}_{\theta|\mathbb{T}} + 2d^2 r^2) \mathbb{R}_n^*(\mathbf{a}, \mathbf{t}, \phi) \quad \forall n \in \mathbb{N}_{>(\text{at})_2^{-2}}$$

(or for $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d}^{\geq}$ with $D = 3d^2$).

§08101.64 **Proof of Proposition** §08101.63. Given in the lecture. \square

§08101.65 **Remark.** Under the assumptions of **Proposition** §08101.63 if there exists in addition $\mathbb{v} \in \mathbb{R}_{>0}$ satisfying $\|\Gamma_{\theta|\mathbb{T}}\|_{\mathbb{L}(\ell_2)} \leq \mathbb{v}$ for all $\theta \in \ell_2^{\text{a,r}}$ and $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D}$ (or $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d}^{\geq}$), then we have

$$\sup \{ \mathbb{P}_{\theta|\mathbb{T}}^n (|\phi \nu_{\mathbb{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{\text{a,r}}, \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D} \} \leq D^2 (\mathbb{v} + 2d^2 r^2) \mathbb{R}_n^*(\mathbf{a}, \mathbf{t}, \phi) \quad \forall n \in \mathbb{N}_{>(\text{at})_2^{-2}}$$

Arguing similarly as in **Remark** §07101.56 we note that $\|\mathbf{t}^{-1} \mathbb{1}^m\|_{\phi} \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and $(\|\phi \mathbb{1}^{m \perp}\|_{\mathbf{a}}^2 = o(1))$ as $m \rightarrow \infty$ (since $(\mathbf{a}\phi) \in \ell_2$), and hence $\mathbb{R}_n^*(\mathbf{a}, \mathbf{t}, \phi) = o(1)$ as $n \rightarrow \infty$. Note that the dimension $m_n^* := m_n^*(\mathbf{a}, \mathbf{t}, \phi)$ as defined in (08.15) does not depend on the unknown parameter of interest θ but on the classes $\ell_2^{\text{a,r}}$ and $\mathbb{T}_{t,d,D}$ only, and thus also the statistic $\widehat{\theta}^{m_n^*}$. In other words, if the regularity of θ and $\mathbb{T}_{\cdot,\cdot}$ is known in advance, then the (generalised) GE $\widehat{\theta}^{m_n^*}$ is a feasible estimator. \square

§08101.66 **Corollary** (GniSM §08101.04 continued). Consider $\widehat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta|\mathbb{T}}^n$ as in **Model** §08101.04, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$, $\mathbb{T}_{\cdot,\cdot} \in \mathbb{L}^{\geq}(\ell_2)$ or $\mathbb{T}_{\cdot,\cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$, $\theta \in \ell_2$ and hence $g = \mathbb{T}_{\cdot,\cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot,\cdot}^{\uparrow}) \subseteq \ell_2$. Under Assumption §08101.58 the (generalised) GE $\widehat{\theta}^{m_n^*} = \mathbb{T}_{\cdot,\cdot}^{m_n^* \uparrow} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^*} \subseteq \text{dom}(\phi \nu_{\mathbb{N}})$ with dimension m_n^* as in (08.15) satisfies

$$\sup \{ N_{\theta|\mathbb{T}}^n (|\phi \nu_{\mathbb{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{\text{a,r}}, \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D} \} \leq C_{r,d,D} \mathbb{R}_n^*(\mathbf{a}, \mathbf{t}, \phi) \quad \forall n \in \mathbb{N}_{>(\text{at})_2^{-2}} \quad (08.16)$$

with constant $C_{r,d,D} = D^2(1 + 2d^2 r^2)$ (for $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d}^{\geq}$ with $D = 3d^2$).

§08101.67 **Proof of Corollary** §08101.66. Given in the lecture. \square

§08101.68 **Corollary** (niSM §08101.06 continued). Consider $\widehat{g} = g + n^{-1/2} \dot{\epsilon} \sim P_{\theta|\mathbb{T}|\sigma}^n$ as in **Model** §08101.06, where $\dot{\epsilon} \sim \otimes_{j \in \mathbb{N}} P_{(0,\sigma^2)}$ satisfies (iSM1) with $\|\sigma^2\|_{\ell_{\infty}} =: \mathbb{v}_{\sigma} \in \mathbb{R}_{>0}$, $\mathbb{T}_{\cdot,\cdot} \in \mathbb{L}^{\geq}(\ell_2)$ or $\mathbb{T}_{\cdot,\cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$, $\theta \in \ell_2$ and hence $g = \mathbb{T}_{\cdot,\cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot,\cdot}^{\uparrow}) \subseteq \ell_2$. Under Assumption §08101.58 the (generalised) GE $\widehat{\theta}^{m_n^*} = \mathbb{T}_{\cdot,\cdot}^{m_n^* \uparrow} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^*} \subseteq \text{dom}(\phi \nu_{\mathbb{N}})$ with dimension m_n^* as in (08.15) satisfies

$$\sup \{ P_{\theta|\mathbb{T}|\sigma}^n (|\phi \nu_{\mathbb{N}}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{\text{a,r}}, \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D} \} \leq C_{r,d,D,\sigma} \mathbb{R}_n^*(\mathbf{a}, \mathbf{t}, \phi) \quad \forall n \in \mathbb{N}_{>(\text{at})_2^{-2}}$$

with constant $C_{r,d,D,\sigma} = D^2(\|\sigma^2\|_{\ell_{\infty}} + 2d^2 r^2)$ (for $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d}^{\geq}$ with $D = 3d^2$).

§08101.69 **Proof of Corollary** §08101.68. Given in the lecture. \square

§08101.70 **Corollary** (nieMM §08101.08 continued). Let $\widehat{g} = g + n^{-1/2}\dot{\varepsilon}$ be defined on $(\mathcal{Z}^n, \mathcal{Z}^{\otimes n}, \mathbb{P}_{\theta|\mathbb{T}}^{\otimes n})$ as in Model §08101.08, where $\psi \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}})$ satisfies (nieMM1) and (nieMM2) for some $\nu_{\theta|\mathbb{T}|\psi} \in \mathbb{R}_{>1}$, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}^{\geq}(\ell_2)$ or $\mathbb{T}_{\cdot, \cdot} \in \mathbb{L}^{\leq}(\ell_2)$, $\theta \in \ell_2$ and hence $g = \mathbb{T}_{\cdot, \cdot}\theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^{\dagger}) \subseteq \ell_2$. Under Assumption §08101.58 the (generalised) GE $\widehat{\theta}^{m_n^*} = \mathbb{T}_{\cdot, \cdot}^{m_n^* \dagger} \widehat{g} \in \ell_2 \mathbb{1}^{m_n^*} \subseteq \text{dom}(\phi_{\nu_n})$ with dimension m_n^* as in (08.15) satisfies

$$\sup \{ \mathbb{P}_{\theta|\mathbb{T}}^{\otimes n} (|\phi_{\nu_n}(\widehat{\theta}^{m_n^*} - \theta)|^2) : \theta \in \ell_2^{\text{a,r}}, \mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{\text{t,d,D}} \} \leq C_{r,\text{a},\text{d},\text{D},\text{t}} R_n^*(\mathbf{a}, \mathbf{t}, \phi) \quad \forall n \in \mathbb{N}_{>(\text{at})^2}$$

with constant $C_{r,\text{a},\text{d},\text{D},\text{t}} = D^2(\sup \{ \nu_{\theta|\mathbb{T}|\psi} : \theta \in \ell_2^{\text{a,r}}, \mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{\text{t,d,D}} \} + 2d^2r^2)$ (for $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{\text{t,d}}^{\geq}$ with $D = 3d^2$).

§08101.71 **Proof of Corollary** §08101.70. Given in the lecture. \square

§08101.72 **Illustration**. We distinguish the following two cases **(p)** $(\phi/\mathbf{t}) \in \ell_2$, and **(np)** $(\phi/\mathbf{t}) \notin \ell_2$. Interestingly, in case **(p)** the bound in Proposition §08101.63 is parametric, that is, $nR_n^*(\mathbf{a}, \mathbf{t}, \phi) = O(1)$, in case **(np)** the bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^*(\mathbf{a}, \mathbf{t}, \phi) = \infty$. In case **(p)** we consider similar to **(o-m)**, **(o-s)** and **(s-m)** in Illustration §07101.78 the following specifications:

Table 04 [§08]

Order of the rate $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ as $n \rightarrow \infty$

	$(j \in \mathbb{N})$	$(\mathbf{a} \in \mathbb{R}_{>0})$	$(\mathbf{t} \in \mathbb{R}_{>0})$	(squared bias)	(variance)		
	$\phi_j^2 = j^{2v-1}$	\mathbf{a}_j^2	\mathbf{t}_j^2	$\ \mathbf{a} \cdot \mathbb{1}^m\ _{\phi}^2$	$\ \mathbf{t}^{-1} \mathbb{1}^m\ _{\phi}^2$	m_n^*	$R_n^*(\mathbf{a}, \mathbf{t}, \phi)$
(o-m)	$v \in (-t, \mathbf{a})$	$j^{-2\mathbf{a}}$	$j^{-2\mathbf{t}}$	$m^{-2(\mathbf{a}-v)}$	$m^{2v+2\mathbf{t}}$	$n^{\frac{1}{2\mathbf{a}+2\mathbf{t}}}$	$n^{-\frac{\mathbf{a}-v}{\mathbf{a}+\mathbf{t}}}$
	$v = -t$	$j^{-2\mathbf{a}}$	$j^{-2\mathbf{t}}$	$m^{-2(\mathbf{a}+\mathbf{t})}$	$\log m$	$\left(\frac{n}{\log n}\right)^{\frac{1}{2(\mathbf{a}+\mathbf{t})}}$	$\frac{\log n}{n}$
(o-s)	$\mathbf{a} - v \in \mathbb{R}_{>0}$	$j^{-2\mathbf{a}}$	$e^{-j^{2\mathbf{t}}}$	$m^{-2(\mathbf{a}-v)}$	$m^{2(v-\mathbf{t})} e^{m^{2\mathbf{t}}}$	$(\log n)^{\frac{1}{2\mathbf{t}}}$	$(\log n)^{-\frac{\mathbf{a}-v}{\mathbf{t}}}$
(s-m)	$v + \mathbf{t} \in \mathbb{R}_{>0}$	$e^{-j^{2\mathbf{a}}}$	$j^{-2\mathbf{t}}$	$e^{-m^{2\mathbf{a}}}$	$m^{2v+2\mathbf{t}}$	$(\log n)^{\frac{1}{2\mathbf{a}}}$	$\frac{(\log n)^{\frac{\mathbf{t}+v}{\mathbf{a}}}}{n}$
	$v = -t$	$e^{-j^{2\mathbf{a}}}$	$j^{-2\mathbf{t}}$	$e^{-m^{2\mathbf{a}}}$	$\log m$	$(\log n)^{\frac{1}{2\mathbf{a}}}$	$\frac{\log \log n}{n}$

We note that in case **(o-m)** and **(s-m)** for $v < -t$ the rate $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ is parametric. \square

§08|02 Non-diagonal statistical inverse problem with noisy operator

§08102.01 **Notation Reminder**. For $\mathbf{A}_{\cdot, \cdot} = (A_{j,j_o})_{j,j_o \in \mathbb{N}} \in \mathbb{R}^{\mathbb{N}^2}$ we denote by $\mathbf{A}_{\cdot, \cdot}^m := \mathbf{M}_{\mathbb{1}^m} \mathbf{A}_{\cdot, \cdot} \mathbf{M}_{\mathbb{1}^m} \in \mathbb{L}(\ell_2)$ with

$$\mathbf{a}_{\cdot} \mapsto \mathbf{A}_{\cdot, \cdot}^m \mathbf{a}_{\cdot} = (\mathbb{1}_j^m \sum_{j_o \in \llbracket m \rrbracket} A_{j,j_o} a_{j_o}) = \mathbb{1}_j^m \langle A_{j, \cdot} \mathbb{1}^m, \mathbf{a}_{\cdot} \rangle_{\ell_2} = \mathbb{1}_j^m \nu_{\mathbb{N}}(A_{j, \cdot} \mathbf{a}_{\cdot} \mathbb{1}^m)_{j \in \mathbb{N}}$$

the operator which restricted to a linear map from \mathbb{R}^m ($\text{ran}(\mathbf{M}_{\mathbb{1}^m}) = \ell_2 \mathbb{1}^m$) into itself is represented by the sub-matrix $[\mathbf{A}_{\cdot, \cdot}]_{\llbracket m \rrbracket} := (A_{j,j_o})_{j,j_o \in \llbracket m \rrbracket} \in \mathbb{R}^{(m,m)}$ (compare Notation §05100.02). Moreover, $\|\cdot\|$ and $\|A\|_{\text{spec}} := \sup\{\|Ax\| : \|x\| \leq 1\}$ denotes, respectively, the Euclidean norm of a vector and the spectral norm of a matrix A . Clearly, we have $\|\mathbf{A}_{\cdot, \cdot}^m\|_{\mathbb{L}(\ell_2)} = \|\mathbf{M}_{\mathbb{1}^m} \mathbf{A}_{\cdot, \cdot} \mathbf{M}_{\mathbb{1}^m}\|_{\mathbb{L}(\ell_2)} = \|[\mathbf{A}_{\cdot, \cdot}]_{\llbracket m \rrbracket}\|_{\text{spec}}$. Furthermore, $\mathbf{A}_{\cdot, \cdot}^m \in \mathbb{L}(\ell_2)$ is a Hilbert-Schmidt operator (§08101.14), i.e. $\mathbf{A}_{\cdot, \cdot}^m \in \text{HS}(\ell_2)$, and $\mathbf{M}_w \mathbf{A}_{\cdot, \cdot}^m = \mathbf{M}_w^m \mathbf{A}^m \in \text{HS}(\ell_2)$ for arbitrary $w \in \mathbb{R}^{\mathbb{N}}$. Moreover, introduce $\ell_p(\mathbb{N}^2) := \mathbb{L}_p(\mathbb{N}^2, 2^{\mathbb{N}^2}, \nu_{\mathbb{N}^2})$ for $p \in \overline{\mathbb{R}}_{\geq 1}$. \square

§08102.02 **Assumption**. Consider a stochastic process $\dot{\varepsilon} = (\dot{\varepsilon}_j)_{j \in \mathbb{N}}$ satisfying Assumption §01101.04 with mean zero and a sample size $n \in \mathbb{N}$, and in addition a stochastic process $\dot{\eta}_{\cdot, \cdot} = (\dot{\eta}_{j,j_o})_{j,j_o \in \mathbb{N}}$ satisfying Assumption §02101.02, with mean zero and a sample size $k \in \mathbb{N}$. Let Assumption §08100.02

be satisfied where $T_{\cdot,\cdot} \in \mathbb{L}^{\dot{\cdot}}(\ell_2)$ or $T_{\cdot,\cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ is *not known* anymore. For $\theta \in \ell_2$ the observable noisy image with mean $g = T_{\cdot,\cdot}\theta \in \ell_2$ and the observable noisy non-diagonal operator with mean kernel $T_{\cdot,\cdot} \in \mathbb{R}^{\mathbb{N}^2}$ takes the form $\widehat{g} = g + n^{-1/2}\dot{\epsilon}$ and $\widehat{T}_{\cdot,\cdot} = T_{\cdot,\cdot} + k^{-1/2}\dot{\eta}_{\cdot,\cdot}$, respectively. We denote by $\mathbb{P}_{\theta|T}^{n,k}$ the joint distribution of $(\widehat{g}, \widehat{T}_{\cdot,\cdot})$. Denoting by $\mathbb{P}_{\theta|T}^n$ and \mathbb{P}_T^k the marginal distribution of \widehat{g} and $\widehat{T}_{\cdot,\cdot}$, respectively, if $\dot{\epsilon}$ and $\dot{\eta}_{\cdot,\cdot}$ are *independent* then we write $\mathbb{P}_{\theta|T}^{n \otimes k} = \mathbb{P}_{\theta|T}^n \otimes \mathbb{P}_T^k$ for the joint product distribution of $(\widehat{g}, \widehat{T}_{\cdot,\cdot})$. In addition $\dot{\epsilon}$ satisfies (nSIP) (Assumption §08I01.01) with $\mathfrak{v}_{\theta|T} \in \mathbb{R}_{>1}$ and $\dot{\eta}_{\cdot,\cdot}$ fulfils

(nSIPnO1) there is $\mathfrak{v}_T \in \mathbb{R}_{>1}$ such that $\dot{\eta}_{\cdot,\cdot}$ for all $m \in \mathbb{N}$ and $a, b \in \ell_2$ satisfies

$$\mathbb{P}_T^k \left(|\langle b, \dot{\eta}_{\cdot,\cdot}^m a \rangle_{\ell_2}|^2 \right) \leq \mathfrak{v}_T \|a\|_{\ell_2}^2 \|b\|_{\ell_2}^2$$

implying $\mathbb{P}_T^k (\dot{\eta}_{j|j_0}^2) =: \mathfrak{v}_{j|j_0}^T \leq \mathfrak{v}_T$ for all $j, j_0 \in \mathbb{N}$, and hence $1 \vee \|\mathfrak{v}_{\cdot,\cdot}^T\|_{\ell_\infty(\mathbb{N}^2)} \leq \mathfrak{v}_T$;

(nSIPnO2) there is $l \in \mathbb{N}$ and $K_T^2 \in \mathbb{R}_{>\mathfrak{v}_T}$ such that $\dot{\eta}_{\cdot,\cdot}$ satisfies

$$\mathfrak{v}_{\cdot,\cdot}^{T(l)} := \mathbb{P}_T^k (\dot{\eta}_{\cdot,\cdot}^{2l}) := (\mathfrak{v}_{j|j_0}^{T(l)} := \mathbb{P}_T^k (\dot{\eta}_{j|j_0}^{2l}))_{j, j_0 \in \mathbb{N}} \in \ell_\infty(\mathbb{N}^2),$$

and $1 \vee \|\mathfrak{v}_{\cdot,\cdot}^{T(l)}\|_{\ell_\infty(\mathbb{N}^2)} \leq K_T^{2l}$ where $1 \vee \|\mathfrak{v}_{\cdot,\cdot}^T\|_{\ell_\infty(\mathbb{N}^2)} \leq \mathfrak{v}_T \leq K_T^2$. \square

§08I02.03 **Lemma.** Let Assumption §08I02.02 (nSIPnO1) and (nSIPnO2) be satisfied, and let $m \in \mathbb{N}$.

(i) Under (nSIPnO1) for any $A_{\cdot,\cdot} \in \text{HS}(\ell_2)$ and $a \in \ell_2$ we have

$$\mathbb{P}_T^k \left(\|A_{\cdot,\cdot} \dot{\eta}_{\cdot,\cdot}^m a\|_{\ell_2}^2 \right) \leq \mathfrak{v}_T \|A_{\cdot,\cdot}\|_{\text{HS}}^2 \|a\|_{\ell_2}^2$$

and in particular, $m^{-1} \mathbb{P}_T^k \left(\|\dot{\eta}_{\cdot,\cdot}^m a\|_{\ell_2}^2 \right) \leq \mathfrak{v}_T \|a\|_{\ell_2}^2$ by using $\|\text{id}_{\cdot,\cdot}\|_{\text{HS}}^2 = m$.

(ii) Under (nSIPnO2) for all $x \in \mathbb{R}_{>0}$ we have

$$\mathbb{P}_T^k \left(\|\dot{\eta}_{\cdot,\cdot}^m\|_{\mathbb{L}(\ell_2)} \geq x \right) \leq m^{2l} x^{-2l} K_T^{2l}$$

and $\mathbb{P}_T^k \left(\|\dot{\eta}_{\cdot,\cdot}^m\|_{\mathbb{L}(\ell_2)}^2 \mathbb{1}_{\{\|\dot{\eta}_{\cdot,\cdot}^m\|_{\mathbb{L}(\ell_2)} \geq x\}} \right) \leq m^{2l} x^{-2(l-1)} K_T^{2l}$.

§08I02.04 **Proof of Lemma §08I02.03.** Given in the lecture. \square

§08I02.05 **Notation.** For each $m \in \mathbb{N}$ and $T_{\cdot,\cdot} \in \mathbb{L}(\ell_2)$ we introduce below an observable event $\Omega_{m,k \wedge n}$ and its complement $\Omega_{m,k \wedge n}^c$ such that on the event $\Omega_{m,k \wedge n}$ the random matrix $[\widehat{T}_{\cdot,\cdot}]_m \in \mathbb{R}^{(m,m)}$ is regular and hence its inverse $[\widehat{T}_{\cdot,\cdot}]_m^{-1} \in \mathbb{R}^{(m,m)}$ always exists. We denote by $\mathbb{1}_{\Omega_{m,k \wedge n}}$ the observable elementary random variable which takes the value 1 on the event $\Omega_{m,k \wedge n}$ and zero otherwise. We denote by $\widehat{T}_{\cdot,\cdot}^m := M_{\mathbb{1}^m} \widehat{T}_{\cdot,\cdot} M_{\mathbb{1}^m} \in \mathbb{L}(\ell_2)$ the random operator which restricted to a linear map from \mathbb{R}^m into itself can be represented by the random matrix $[\widehat{T}_{\cdot,\cdot}]_m$. Note that its Moore-Penrose inverse $\widehat{T}_{\cdot,\cdot}^{m|\dagger} \in \mathbb{L}(\ell_2)$ restricted to a linear map from \mathbb{R}^m into itself can be represented by the Moore-Penrose inverse matrix $[\widehat{T}_{\cdot,\cdot}]_m^\dagger$ of $[\widehat{T}_{\cdot,\cdot}]_m$ (see Definition §03I00.08). On the event $\Omega_{m,k \wedge n}$ we have $\widehat{T}_{\cdot,\cdot}^m \widehat{T}_{\cdot,\cdot}^{m|\dagger} = \widehat{T}_{\cdot,\cdot}^{m|\dagger} \widehat{T}_{\cdot,\cdot}^m = M_{\mathbb{1}^m}$. Let $0_{\cdot,\cdot} \in \mathbb{L}(\ell_2)$ be the zero operator mapping ℓ_2 to $\{0\}$. The random operator, which equals $A_{\cdot,\cdot} \in \mathbb{L}(\ell_2)$ on $\Omega_{m,k \wedge n}$ and $0_{\cdot,\cdot}$ on $\Omega_{m,k \wedge n}^c$, is denoted by $A_{\cdot,\cdot}^{(k \wedge n)} := A_{\cdot,\cdot} \mathbb{1}_{\Omega_{m,k \wedge n}} \in \mathbb{L}(\ell_2)$. Let $\widehat{T}_{\cdot,\cdot}^{m|(k \wedge n)} := \widehat{T}_{\cdot,\cdot}^m \mathbb{1}_{\Omega_{m,k \wedge n}} \in \mathbb{L}(\ell_2)$ and denote its Moore-Penrose inverse by $\widehat{T}_{\cdot,\cdot}^{m|(k \wedge n)|\dagger} \in \mathbb{L}(\ell_2)$ where trivially $\widehat{T}_{\cdot,\cdot}^{m|(k \wedge n)|\dagger} = \widehat{T}_{\cdot,\cdot}^{m|\dagger} \mathbb{1}_{\Omega_{m,k \wedge n}}$. We eventually use the elementary identity $\widehat{T}_{\cdot,\cdot}^{m|\dagger} \widehat{T}_{\cdot,\cdot}^m \mathbb{1}_{\Omega_{m,k \wedge n}} = \widehat{T}_{\cdot,\cdot}^{m|(k \wedge n)|\dagger} \widehat{T}_{\cdot,\cdot}^{m|(k \wedge n)} = \widehat{T}_{\cdot,\cdot}^{m|(k \wedge n)} \widehat{T}_{\cdot,\cdot}^{m|(k \wedge n)|\dagger} = M_{\mathbb{1}^m} \mathbb{1}_{\Omega_{m,k \wedge n}}$. \square

§08I02.06 **Definition.** Under Assumption §08I02.02 let $(\widehat{g}, \widehat{T}_{\cdot,\cdot}) \sim \mathbb{P}_{\theta|T}^{n,k}$ be noisy versions of $g \in \text{dom}(T_{\cdot,\cdot}^\dagger)$ and $T_{\cdot,\cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ (or $T_{\cdot,\cdot} \in \mathbb{L}^{\dot{\cdot}}(\ell_2)$). For each $m \in \mathbb{N}$ we call $\widehat{\theta}^m = \widehat{T}_{\cdot,\cdot}^{m|(k \wedge n)|\dagger} \widehat{g} = \widehat{T}_{\cdot,\cdot}^{m|\dagger} \mathbb{1}_{\Omega_{m,k \wedge n}} \widehat{g} \mathbb{1}^m = \widehat{T}_{\cdot,\cdot}^{m|\dagger} \mathbb{1}_{\Omega_{m,k \wedge n}} \widehat{g}^m$ (generalised) thresholded *Galerkin estimator (tGE)* of $\theta = T_{\cdot,\cdot}^\dagger g \in \ell_2$. \square

§08102.07 **Remark.** Under Assumption §08102.02 we have $\dot{\xi} \mathbb{1}^m \in \ell_\infty \mathbb{P}_{\theta|\mathbb{T}}^n$ -a.s. and $\widehat{\mathbb{T}}_{\cdot,\cdot}^m \in \mathbb{L}(\ell_2)$ with $\text{ran}(\widehat{\mathbb{T}}_{\cdot,\cdot}^m) \subseteq \ell_2 \mathbb{1}^m \mathbb{P}_{\mathbb{T}}^k$ -a.s. for each $m \in \mathbb{N}$. Consequently, $\text{ran}(\widehat{\mathbb{T}}_{\cdot,\cdot}^{m|(k \wedge n)\dagger}) \subseteq \ell_2 \mathbb{1}^m \mathbb{P}_{\mathbb{T}}^k$ -a.s., and $\widehat{\mathbb{T}}_{\cdot,\cdot}^{m|(k \wedge n)\dagger} \dot{\xi} \in \ell_2 \mathbb{1}^m \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k}$ -a.s., and hence

$$\widehat{\theta}^m = \widehat{\mathbb{T}}_{\cdot,\cdot}^{m|(k \wedge n)\dagger} \widehat{g} = n^{-1/2} \widehat{\mathbb{T}}_{\cdot,\cdot}^{m|(k \wedge n)\dagger} \dot{\xi} + \widehat{\mathbb{T}}_{\cdot,\cdot}^{m|(k \wedge n)\dagger} g \in \ell_2 \mathbb{1}^m \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k}\text{-a.s.}$$

Let us recall that the (generalised) Galerkin solution $\theta^m \in \ell_2 \mathbb{1}^m$ does generally not correspond to the orthogonal projection $\mathbb{1}^{m \perp} \theta = (\mathbb{1} - \mathbb{1}^m) \theta$. Moreover, the approximation error $\sup\{\|\theta^m - \theta\|_{\ell_2} : m \geq n\}$ does generally not converge to zero as $n \rightarrow \infty$ (compare Remark §05101.05). Here and subsequently, we will restrict ourselves to classes of solutions and operators which ensure the convergence. Obviously, this is a minimal regularity condition for us if we aim to estimate the Galerkin solution. \square

§08|02|01 Examples

§08102.08 **GniSM with noisy operator (§02102.06 continued).** Let Assumption §08100.02 be satisfied where $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T} \subseteq \mathbb{L}^{\mathbb{R}}(\ell_2)$ or $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T} \subseteq \mathbb{L}^{\mathbb{C}}(\ell_2)$ is *not known* anymore. We illustrate the (generalised) tGE in a Gaussian non-diagonal inverse sequence model (GniSM) with noisy operator as in §02102.06. Here the observable stochastic processes $\widehat{\mathbb{T}}_{\cdot,\cdot} = \mathbb{T}_{\cdot,\cdot} + k^{-1/2} \dot{W}_{\cdot,\cdot} \sim N_{\mathbb{T}}^k$ and $\widehat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta|\mathbb{T}}^n$ are noisy version of $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}$ and $g = \mathbb{T}_{\cdot,\cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot,\cdot}) \subseteq \ell_2$ with $\theta \in \Theta \subseteq \ell_2$, respectively, where $\dot{W}_{\cdot,\cdot} := (\dot{W}_{j|j})_{j,j_0 \in \mathbb{N}} \sim N_{(0,1)}^{\otimes \mathbb{N}^2}$ and $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ are *independent*. Consequently, $(\widehat{g}, \widehat{\mathbb{T}}_{\cdot,\cdot})$ admits a joint $N_{\theta|\mathbb{T}}^{n \otimes k} = N_{\theta|\mathbb{T}}^n \otimes N_{\mathbb{T}}^k$ -distribution belonging to the family $N_{\Theta \times \mathbb{T}}^{n \otimes k} := (N_{\theta|\mathbb{T}}^n \otimes N_{\mathbb{T}}^k)_{\theta \in \Theta, \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}^3}, \mathcal{B}^{\otimes \mathbb{N}^3}, N_{\Theta \times \mathbb{T}}^{n \otimes k})$ where $\Theta \subseteq \ell_2$ and $\mathbb{T} \subseteq \mathbb{L}^{\mathbb{R}}(\ell_2)$ or $\mathbb{T} \subseteq \mathbb{L}^{\mathbb{C}}(\ell_2)$. \square

§08102.09 **Property** (GniSM with noisy operator §08102.08 continued). *Let $\dot{W}_{\cdot,\cdot} \sim N_{(0,1)}^{\otimes \mathbb{N}^2}$ and $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ be independent as in Model §08102.08. Then Assumption §08102.02 is satisfied:*

- (i) *Due to Property §07101.04 \dot{B} admits $\text{id}_{\ell_2} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ as covariance operator with $\|\text{id}_{\ell_2}\|_{\mathbb{L}(\ell_2)} = 1$, i.e. (nSIP) is fulfilled with $\mathfrak{v}_{\theta|\mathbb{T}} = 1$. For all $h_{\cdot} \in \ell_2$ we have $\|h_{\cdot}\|_{\ell_2}^2 = \|h_{\cdot}\|_{\text{id}_{\ell_2}}^2 = \langle \text{id}_{\ell_2} h_{\cdot}, h_{\cdot} \rangle_{\ell_2}$.*
- (ii) *For all $m \in \mathbb{N}$ and $a_{\cdot}, b_{\cdot} \in \ell_2$ we have $\dot{W}_{\cdot,\cdot}^m a_{\cdot} \sim \|a_{\cdot}\|_{\ell_2} \dot{B}^m$, $\langle b_{\cdot}, \dot{B}^m \rangle_{\ell_2} \sim N_{(0, \|a_{\cdot}\|_{\ell_2}^2)}^{\mathbb{R}}$, and hence*

$$N_{(0,1)}^{\otimes \mathbb{N}^2} (|\langle b_{\cdot}, \dot{W}_{\cdot,\cdot}^m a_{\cdot} \rangle_{\ell_2}|^2) = \|a_{\cdot}\|_{\ell_2}^2 \|b_{\cdot}\|_{\ell_2}^2 \leq \|a_{\cdot}\|_{\ell_2}^2 \|b_{\cdot}\|_{\ell_2}^2,$$

i.e. (nSIPnO1) is satisfied with $\mathfrak{v}_{\mathbb{T}} = 1$.

- (iii) *For any $l \in \mathbb{N}$ setting $K_{2l}^{2l} := \prod_{j \in [l]} (2j - 1) =: (2l - 1)!!$ we have $K_{2l}^{2l} \geq 1$ and $1 \vee \|\mathbb{V}_{\cdot,\cdot}^{\mathbb{T}(l)}\|_{\ell_\infty(\mathbb{N}^2)} = K_{2l}^{2l}$, i.e. (nSIPnO2) is satisfied with $K_{\mathbb{T}} = K_{2l}$.* \square

§08102.10 **niSM with noisy operator (§02102.05 continued).** Let Assumption §08100.02 be satisfied where $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T} \subseteq \mathbb{L}^{\mathbb{R}}(\ell_2)$ or $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T} \subseteq \mathbb{L}^{\mathbb{C}}(\ell_2)$ is *not known* anymore. We illustrate the (generalised) GE in a Non-diagonal inverse sequence model (niSM) with noisy operator as in §02102.05. Here the observable stochastic process $\widehat{\mathbb{T}}_{\cdot,\cdot} = \mathbb{T}_{\cdot,\cdot} + k^{-1/2} \dot{\eta}_{\cdot,\cdot} \sim \mathbb{P}_{\mathbb{T}}^k$ and $\widehat{g} = g + n^{-1/2} \dot{\xi} \sim \mathbb{P}_{\theta|\mathbb{T}}^n$ is a noisy version of $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}$ and $g = \mathbb{T}_{\cdot,\cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot,\cdot}) \subseteq \ell_2$ with $\theta \in \Theta \subseteq \ell_2$, respectively, where $\dot{\xi} \sim \otimes_{j \in \mathbb{N}} \mathbb{P}^{\xi}$ and $\dot{\eta}_{\cdot,\cdot} \sim \otimes_{j,j_0 \in \mathbb{N}} \mathbb{P}^{\eta_{j|j_0}}$ are *independent*. In addition, let $\dot{\xi}$ satisfy (iSM1) of Model §07101.06 for $\sigma \in \Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$ and let $\dot{\eta}_{\cdot,\cdot}$ fulfill

(niSMnO1) for $\xi_{\cdot,\cdot} \in \Xi \subseteq (\mathbb{R}_{>0})^{\mathbb{N}^2} \cap \ell_\infty(\mathbb{N}^2)$ we have $\dot{\eta}_{j|j_0} \sim P_{(0, \xi_{j|j_0})} \in \mathcal{W}_2(\mathcal{B})$, for all $j, j_0 \in \mathbb{N}$,

(niSMnO2) for $l \in \mathbb{N}$ and $\xi_{\cdot,\cdot}^{(2l)} \in \Xi^{2l} \subseteq (\mathbb{R}_{>0})^{\mathbb{N}^2} \cap \ell_\infty(\mathbb{N}^2)$ we have $\xi_{\cdot,\cdot}^{(2l)} := (\xi_{j|j_0}^{(2l)})_{j,j_0 \in \mathbb{N}}$.

Under (iSM1) \widehat{g}_\bullet admits a $\mathbb{P}_{\theta|\mathbb{T}_\sigma}^n$ -distribution belonging to the family $\mathbb{P}_{\Theta \times \mathbb{T} \times \Sigma}^n := (\mathbb{P}_{\theta|\mathbb{T}_\sigma}^n)_{\theta \in \Theta, \mathbb{T}_\bullet \in \mathbb{T}, \alpha \in \Sigma}$ and under (niSMnO1), (niSMnO2) $\widehat{\mathbb{T}}_\bullet$ admits a $\mathbb{P}_{\mathbb{T}|\xi|\xi^{(2)}}^k$ -distribution belonging to the family $\mathbb{P}_{\mathbb{T} \times \Xi \times \Xi^{(2)}}^k := (\mathbb{P}_{\mathbb{T}|\xi|\xi^{(2)}}^k)_{\mathbb{T}_\bullet \in \mathbb{T}, \xi_\bullet \in \Xi, \xi_\bullet^{(2)} \in \Xi^{(2)}}$. Summarising $(\widehat{g}_\bullet, \widehat{\mathbb{T}}_\bullet)$ admits a joint $\mathbb{P}_{\theta|\mathbb{T}_\sigma|\xi|\xi^{(2)}}^{n \otimes k} = \mathbb{P}_{\theta|\mathbb{T}_\sigma}^n \otimes \mathbb{P}_{\mathbb{T}|\xi|\xi^{(2)}}^k$ distribution belonging to the family $\mathbb{P}_{\Theta \times \mathbb{T} \times \Sigma \times \Xi \times \Xi^{(2)}}^{n \otimes k} := (\mathbb{P}_{\theta|\mathbb{T}_\sigma}^n \otimes \mathbb{P}_{\mathbb{T}|\xi|\xi^{(2)}}^k)_{\theta \in \Theta, \mathbb{T}_\bullet \in \mathbb{T}, \alpha \in \Sigma, \xi_\bullet \in \Xi, \xi_\bullet^{(2)} \in \Xi^{(2)}}$ and the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}^3}, \mathcal{B}^{\otimes \mathbb{N}^3}, \mathbb{P}_{\Theta \times \mathbb{T} \times \Sigma \times \Xi \times \Xi^{(2)}}^{n \otimes k})$ where $\Theta \subseteq \ell_2$ and $\mathbb{T} \subseteq \mathbb{L}^{\mathbb{Z}}(\ell_2)$ or $\mathbb{T} \subseteq \mathbb{L}^{\mathbb{R}}(\ell_2)$. \square

§08102.11 **Lemma** (niSM with noisy operator §08101.06 continued). *Consider error processes $\dot{\eta}_\bullet$ and $\dot{\xi}_\bullet$ as in Model §08102.10 satisfying (iSM1), (niSMnO1) and (niSMnO2). Then Assumption §08102.02 is satisfied:*

- (i) *Due to Property §07101.07 (i) under (iSM1), $\dot{\xi}_\bullet$ admits $\Gamma_{\theta|\xi} = M_{\sigma^2} \in \mathbb{L}^{\mathbb{M}}(\ell_2) \cap \mathbb{L}^{\mathbb{Z}}(\ell_2)$ as covariance operator with $\|M_{\sigma^2}\|_{\mathbb{L}(\ell_2)} = \|\sigma^2\|_{\ell_\infty} \leq \|\sigma^2\|_{\ell_\infty} \vee 1 =: \nu_\sigma \in \mathbb{R}_{\geq 1}$, i.e. (nSIP) is fulfilled with $\nu_{\theta|\mathbb{T}} = \nu_\sigma$. For all $h_\bullet \in \ell_2$ we have $\|h_\bullet\|_{M_{\sigma^2}}^2 = \langle M_{\sigma^2} h_\bullet, h_\bullet \rangle_{\ell_2} \leq \nu_\sigma \|h_\bullet\|_{\ell_2}^2$.*
- (ii) *Under (niSMnO1) for all $m \in \mathbb{N}$ and $a_\bullet, b_\bullet \in \ell_2$ with $1 \vee \|\xi_\bullet\|_{\ell_\infty(\mathbb{N}^2)} =: \nu_\xi \in \mathbb{R}_{\geq 1}$ we have*

$$\mathbb{P}_{\mathbb{T}|\xi|\xi^{(2)}}^k (|\langle b_\bullet, \dot{\eta}_\bullet^m a_\bullet \rangle_{\ell_2}|^2) \leq \nu_\xi \|a_\bullet\|_{\ell_2}^2 \|b_\bullet\|_{\ell_2}^2$$

i.e. (nSIPnO1) is satisfied with $\nu_{\mathbb{T}} = \nu_\xi$.

- (iii) *Under (niSMnO2) setting $1 \vee \|\xi_\bullet^{(2)}\|_{\ell_\infty(\mathbb{N}^2)} =: K_{\xi^{(2)}}^{2l} \in \mathbb{R}_{\geq 1}$ we have $K_{\xi^{(2)}}^{2l} \geq \nu_\xi$ and $1 \vee \|\mathbf{v}_\bullet^{\mathbb{T}(l)}\|_{\ell_\infty(\mathbb{N}^2)} \leq K_{\xi^{(2)}}^{2l}$, i.e. (nSIPnO2) is satisfied with $K_{\mathbb{T}} = K_{\xi^{(2)}}^{2l}$.*

§08102.12 **Proof of Lemma §08102.11.** Given in the lecture. \square

§08102.13 **nieMM with noisy operator (§02102.04 continued).** Let Assumption §08100.02 be satisfied where $\mathbb{T}_\bullet \in \mathbb{T} \subseteq \mathbb{L}^{\mathbb{Z}}(\ell_2)$ or $\mathbb{T}_\bullet \in \mathbb{T} \subseteq \mathbb{L}^{\mathbb{R}}(\ell_2)$ is *not known* anymore. We illustrate the (generalised) tGE in a Non-diagonal inverse empirical mean model (nieMM) with noisy operator as in §02102.04. Here the observable stochastic processes $\widehat{\mathbb{T}}_\bullet = \mathbb{T}_\bullet + k^{-1/2} \dot{\eta}_\bullet \sim \mathbb{P}_\mathbb{T}^k$ and $\widehat{g}_\bullet = g_\bullet + n^{-1/2} \dot{\xi}_\bullet$ are noisy version of $\mathbb{T}_\bullet \in \mathbb{T}$ and $g_\bullet = \mathbb{T}_\bullet \theta \in \text{dom}(\mathbb{T}_\bullet^\dagger) \subseteq \ell_2$ with $\theta \in \Theta \subseteq \ell_2$, and *independent error processes* $\dot{\xi}_\bullet = n^{1/2}(\widehat{\mathbb{P}}_n(\psi) - \mathbb{E}_{\mathbb{P}_\mathbb{T}}(\psi)) \in \mathcal{M}(\mathcal{Z}^{\otimes n} \otimes 2^{\mathbb{N}})$ and $\dot{\eta}_\bullet = k^{1/2}(\widehat{\mathbb{P}}_k(\varphi_\bullet) - \mathbb{E}_\mathbb{T}(\varphi_\bullet)) \in \mathcal{M}(\mathcal{Z}^{\otimes k} \otimes 2^{\mathbb{N}^2})$ satisfying Assumption §01101.04 and Assumption §02101.02. More precisely, on a measurable space $(\mathcal{Z}, \mathcal{Z})$ for each $\theta \in \Theta \subseteq \ell_2$ and $\mathbb{T}_\bullet \in \mathbb{T}$ there are probability measures $\mathbb{P}_{\theta|\mathbb{T}}, \mathbb{P}_\mathbb{T} \in \mathcal{W}(\mathcal{Z})$. Similar to Model §02102.04 consider stochastic processes $\psi_\bullet \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}})$ and $\varphi_\bullet \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}^2})$. In addition for all $\theta \in \Theta$ and $\mathbb{T}_\bullet \in \mathbb{T}$ the process $\psi_\bullet \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}})$ satisfies (nieMM1) and (nieMM2) of Model §08101.08 for $\nu_{\theta|\mathbb{T}_\psi} \in \mathbb{R}_{\geq 1}$ and the process $\varphi_\bullet \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}^2})$ fulfils

(nieMMnO1) $\varphi_{j_i} \in \mathcal{L}_1(\mathbb{P}_\mathbb{T}) := \mathcal{L}_1(\mathcal{Z}, \mathcal{Z}, \mathbb{P}_\mathbb{T})$ for all $j, j_o \in \mathbb{N}$ and $\mathbb{P}_\mathbb{T}(\varphi_\bullet) = \mathbb{T}_\bullet$,

(nieMMnO2) there is $\nu_{\mathbb{T}_\psi} \in \mathbb{R}_{\geq 1}$ such that φ_\bullet for all $m \in \mathbb{N}$ and $a_\bullet, b_\bullet \in \ell_2$ satisfies

$$\mathbb{P}_\mathbb{T} (|\langle b_\bullet, \varphi_\bullet^m a_\bullet \rangle_{\ell_2}|^2) \leq \nu_{\mathbb{T}_\psi} \|a_\bullet\|_{\ell_2}^2 \|b_\bullet\|_{\ell_2}^2.$$

(nieMMnO3) there is $l \in \mathbb{N}$ and $K_{\mathbb{T}_\psi}^{2l} \in \mathbb{R}_{\geq \nu_{\mathbb{T}_\psi}}$ such that φ_\bullet satisfies

$$\mathbb{P}_\mathbb{T}(\varphi_\bullet^{2l}) := (\mathbb{P}_\mathbb{T}(\varphi_{j_i}^{2l}))_{j, j_o \in \mathbb{N}} \in \ell_\infty(\mathbb{N}^2),$$

and $1 \vee \|\mathbb{P}_\mathbb{T}(|\varphi_\bullet - \mathbb{P}_\mathbb{T} \varphi_\bullet|^{2l})\|_{\ell_\infty(\mathbb{N}^2)} \leq K_{\mathbb{T}_\psi}^{2l}$.

We consider a statistical product experiment $(\mathcal{Z}^{n+k}, \mathcal{Z}^{\otimes(n+k)}, \mathbb{P}_{\Theta \times \mathbb{T}}^{n \otimes k} = (\mathbb{P}_{\theta|\mathbb{T}}^{\otimes n} \otimes \mathbb{P}_\mathbb{T}^{\otimes k})_{\theta \in \Theta, \mathbb{T}_\bullet \in \mathbb{T}})$ as in an Empirical mean function §01101.10 where $\Theta \subseteq \ell_2$ and $\mathbb{T} \subseteq \mathbb{L}^{\mathbb{Z}}(\ell_2)$ or $\mathbb{T} \subseteq \mathbb{L}^{\mathbb{R}}(\ell_2)$. \square

§08102.14 **Lemma** (nieMM with noisy operator §08101.08 continued). *Consider error processes $\hat{\eta}_{\cdot}$ and $\hat{\epsilon}_{\cdot}$ as in Model §08102.13 where $\psi_{\cdot} \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}})$ satisfies (nieMM1) and (nieMM2) and $\varphi_{\cdot} \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}^2})$ fulfils (nieMMnO1)-(nieMMnO3). Then Assumption §08102.02 is satisfied:*

(i) *Due to Property §07101.09 (i) under (nieMM1) and (nieMM2) $\hat{\epsilon}_{\cdot}$ admits a covariance operator $\Gamma_{\theta|T} \in \mathbb{L}^{\geq}(\ell_2)$ satisfying $\|\Gamma_{\theta|T}\|_{\mathbb{L}(\ell_2)} \leq \mathfrak{v}_{\theta|T\psi}$. i.e. (nSIP) is fulfilled with $\mathfrak{v}_{\theta|T} = \mathfrak{v}_{\theta|T\psi}$. For all $h_{\cdot} \in \ell_2$ we have $\|h_{\cdot}\|_{\Gamma_{\theta|T}}^2 = \langle \Gamma_{\theta|T} h_{\cdot}, h_{\cdot} \rangle_{\ell_2} \leq \mathfrak{v}_{\theta|T\psi} \|h_{\cdot}\|_{\ell_2}^2$.*

(ii) *Under (nieMMnO1) and (nieMMnO2) for all $m \in \mathbb{N}$ and $a_{\cdot}, b_{\cdot} \in \ell_2$ we have*

$$\mathbb{E}_T^{\otimes k} (|\langle b_{\cdot}, \hat{\eta}_{\cdot}^m a_{\cdot} \rangle_{\ell_2}|^2) \leq \mathfrak{v}_{T|\varphi} \|a_{\cdot}\|_{\ell_2}^2 \|b_{\cdot}\|_{\ell_2}^2$$

i.e. (nSIPnO1) is satisfied with $\mathfrak{v}_T = \mathfrak{v}_{T|\varphi}$.

(iii) *Under (nieMMnO1) and (nieMMnO3) there exists a constant $C_{2l} \in \mathbb{R}_{\geq 1}$ depending on $l \in \mathbb{N}$ only such that we have $1 \vee \|\mathfrak{v}_{\cdot}^{T(l)}\|_{\mathbb{L}_{\infty}(\mathbb{N}^2)} \leq C_{2l} K_{T|\varphi}^{2l}$, i.e. (nSIPnO2) is satisfied with $K_T = C_{2l}^{1/2l} K_{T|\varphi} \in \mathbb{R}_{\geq 1}$.*

§08102.15 **Proof of Lemma §08102.14.** Given in the lecture. □

§08|02|02 Global and maximal global \mathfrak{v} -risk

We measure first the accuracy of the thresholded (generalised) GE $\hat{\theta}^m := \hat{T}_{\cdot}^{m|(n \wedge k)\dagger} \hat{g}$ of the (generalised) Galerkin solution $\theta^m = T_{\cdot}^{m\dagger} g \in \ell_2 \mathbb{1}^m$ with $g_{\cdot} = T_{\cdot} \theta \in \text{dom}(T_{\cdot}^{\dagger})$ and $T_{\cdot} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ by the mean of its global \mathfrak{v} -error introduced in §0510101 and §05102101, i.e. its \mathfrak{v} -risk.

§08102.16 **Reminder.** If $\mathfrak{v}_{\cdot} \in (\mathbb{R}_{\setminus 0})^{\mathbb{N}}$ then we have $\mathfrak{v}_{\cdot}^2 \mathbb{1}^m \in \ell_{\infty}$ and $\ell_2 \mathbb{1}^m \subseteq \ell_2(\mathfrak{v}_{\cdot}^2)$. Consequently, for each $\theta_{\cdot} \in \ell_2(\mathfrak{v}_{\cdot}^2)$ the (generalised) Galerkin solution $\theta^m = T_{\cdot}^{m\dagger} g \in \ell_2 \mathbb{1}^m$ satisfies $\theta^m \in \ell_2(\mathfrak{v}_{\cdot}^2)$ too. If in addition $C_T := \sup \{ \|M_{\mathfrak{v}} T_{\cdot}^{m\dagger} T_{\cdot} M_{\mathbb{1}^m}\|_{\mathbb{L}(\ell_2)}; m \in \mathbb{N} \} \in \mathbb{R}_{\geq 0}$ then $\|\theta^m - \theta_{\cdot}\|_{\mathfrak{v}} \leq (1 + C_T) \|\mathbb{1}^{m \perp} \theta_{\cdot}\|_{\ell_2}$ which implies $\sup \{ \|\theta^j - \theta_{\cdot}\|_{\mathfrak{v}}; j \in \mathbb{N}_{\geq m} \} = o(1)$ as $m \rightarrow \infty$ (Property §05101.24 and Property §05102.08). □

§08102.17 **Comment.** Under Assumption §08102.02 we have $\hat{\epsilon}_{\cdot} \mathbb{1}^m \in \ell_{\infty} \mathbb{P}_{\theta|T}^m$ -a.s. and $\hat{T}_{\cdot}^m \in \mathbb{L}(\ell_2)$ with $\text{ran}(\hat{T}_{\cdot}^m) \subseteq \ell_2 \mathbb{1}^m \mathbb{P}_T^k$ -a.s. for each $m \in \mathbb{N}$. Consequently, $\text{ran}(\hat{T}_{\cdot}^{m|(k \wedge n)\dagger}) \subseteq \ell_2 \mathbb{1}^m \mathbb{P}_T^k$ -a.s., and $\hat{T}_{\cdot}^{m|(k \wedge n)\dagger} \hat{\epsilon}_{\cdot} \in \ell_2 \mathbb{1}^m \mathbb{P}_{\theta|T}^{n \otimes k}$ -a.s., and hence

$$\hat{\theta}^m = \hat{T}_{\cdot}^{m|(k \wedge n)\dagger} \hat{g} = n^{-1/2} \hat{T}_{\cdot}^{m|(k \wedge n)\dagger} \hat{\epsilon}_{\cdot} + \hat{T}_{\cdot}^{m|(k \wedge n)\dagger} g_{\cdot} \in \ell_2 \mathbb{1}^m \subseteq \ell_2(\mathfrak{v}_{\cdot}^2) \quad \mathbb{P}_{\theta|T}^{n \otimes k}\text{-a.s.} \quad \square$$

§08|02|02|01 Global \mathfrak{v} -risk

§08102.18 **Assumption.** Let $\mathfrak{v}_{\cdot} \in \mathbb{R}_{\setminus 0}^{\mathbb{N}}$ and $\theta_{\cdot} \in \ell_2(\mathfrak{v}_{\cdot}^2)$ be satisfied. □

§08102.19 **Definition.** Under Assumptions §08102.02 and §08102.18 the *global \mathfrak{v} -risk* of a (generalised) tGE $\hat{\theta}^m = \hat{T}_{\cdot}^{m|(k \wedge n)\dagger} \hat{g} \in \ell_2 \mathbb{1}^m \subseteq \ell_2(\mathfrak{v}_{\cdot}^2) \mathbb{P}_{\theta|T}^{n \otimes k}$ -a.s. satisfies

$$\mathbb{E}_{\theta|T}^{n \otimes k} (\|\hat{\theta}^m - \theta_{\cdot}\|_{\mathfrak{v}}^2) = \mathbb{E}_{\theta|T}^{n \otimes k} (\|\hat{T}_{\cdot}^{m|(k \wedge n)\dagger} (\hat{g} - g_{\cdot})\|_{\mathfrak{v}}^2) + \mathbb{E}_T^k (\|\hat{T}_{\cdot}^{m|(k \wedge n)\dagger} g_{\cdot} - \theta_{\cdot}\|_{\mathfrak{v}}^2) \quad (08.17)$$

with $\mathbb{E}_{\theta|T}^{n \otimes k} (\|\hat{T}_{\cdot}^{m|(k \wedge n)\dagger} (\hat{g} - g_{\cdot})\|_{\mathfrak{v}}^2) = n^{-1} \mathbb{E}_T^k (\text{tr}(M_{\mathfrak{v}} \hat{T}_{\cdot}^{m\dagger} \Gamma_{\theta|T} (\hat{T}_{\cdot}^{m\dagger})^* M_{\mathfrak{v}}) \mathbb{1}_{\Omega_{m,k \wedge n}})$ (see Property §08101.15). □

§08102.20 **Property.** Under Assumption §08102.02 we have

$$\begin{aligned} \mathbb{E}_T^k (\|\hat{T}_{\cdot}^{m|(k \wedge n)\dagger} g_{\cdot} - \theta_{\cdot}\|_{\mathfrak{v}}^2) &= \mathbb{E}_T^k (\|\hat{T}_{\cdot}^{m\dagger} (T_{\cdot}^m - \hat{T}_{\cdot}^m) \theta^m + (\theta^m - \theta_{\cdot})\|_{\mathfrak{v}}^2 \mathbb{1}_{\Omega_{m,k \wedge n}}) + \|\theta_{\cdot}\|_{\mathfrak{v}}^2 \mathbb{P}_T^k(\Omega_{m,k \wedge n}^c) \\ &\leq 2 \mathbb{E}_T^k (\|\hat{T}_{\cdot}^{m\dagger} (T_{\cdot}^m - \hat{T}_{\cdot}^m) \theta^m\|_{\mathfrak{v}}^2 \mathbb{1}_{\Omega_{m,k \wedge n}}) + 2 \|\theta^m - \theta_{\cdot}\|_{\mathfrak{v}}^2 + \|\theta_{\cdot}\|_{\mathfrak{v}}^2 \mathbb{P}_T^k(\Omega_{m,k \wedge n}^c) \end{aligned}$$

(since $\widehat{\mathbf{T}}_{\bullet}^{m|\dagger} \widehat{\mathbf{T}}_{\bullet}^m \mathbf{1}_{\Omega_{m,k \wedge n}} = \widehat{\mathbf{T}}_{\bullet}^{m|(k \wedge n)|\dagger} \widehat{\mathbf{T}}_{\bullet}^{m|(k \wedge n)} = \mathbf{M}_{\mathbf{1}^m} \mathbf{1}_{\Omega_{m,k \wedge n}}$). \square

§08102.21 **Notation (Reminder).** Let $A \in \mathbb{L}(\ell_2)$ be a *Hilbert-Schmidt operator*, $A \in \mathbb{HS}(\ell_2)$ for short, where $\|A\|_{\mathbb{HS}}^2 := \text{tr}(A^*A) = \text{tr}(AA^*) \in \mathbb{R}_{\geq 0}$. If $\Gamma \in \mathbb{L}(\ell_2)$ then $\text{tr}(A^*\Gamma A) \leq \|\Gamma\|_{\mathbb{L}(\ell_2)} \text{tr}(A^*A) = \|\Gamma\|_{\mathbb{L}(\ell_2)} \|A\|_{\mathbb{HS}}^2$. For arbitrary $A \in \mathbb{L}(\ell_2)$ we have $\mathbf{M}_v A^m = \mathbf{M}_v^m A^m \in \mathbb{HS}(\ell_2)$. \square

§08102.22 **Notation.** For each $m \in \mathbb{N}$ and \mathbf{T}_{\bullet} in $\mathbb{R}(\ell_2)$ we consider the observable event and its complement

$$\Omega_{m,k \wedge n} := \{\|[\mathbf{M}_v]_{\mathbb{m}} [\widehat{\mathbf{T}}_{\bullet}^m]_{\mathbb{m}}^{-1}\|_{\mathbb{HS}}^2 \leq k \wedge n\} \quad \text{and} \quad \Omega_{m,k \wedge n}^c := \{\|[\mathbf{M}_v]_{\mathbb{m}} [\widehat{\mathbf{T}}_{\bullet}^m]_{\mathbb{m}}^{-1}\|_{\mathbb{HS}}^2 > k \wedge n\}. \quad (08.18)$$

On the event $\Omega_{m,k \wedge n}$ the random matrix $[\widehat{\mathbf{T}}_{\bullet}^m]_{\mathbb{m}} \in \mathbb{R}^{(m,m)}$ is regular with inverse $[\widehat{\mathbf{T}}_{\bullet}^m]_{\mathbb{m}}^{-1} \in \mathbb{R}^{(m,m)}$. Moreover, setting $A_{\bullet}^m := \mathfrak{r}_{\bullet}^m \mathbf{T}_{\bullet}^{m|\dagger}$ we introduce an unobserved event and its complement

$$\mathcal{U}_{m,k} := \{4m \|A_{\bullet}^m\|_{\mathbb{L}(\ell_2)}^2 \leq k\} \quad \text{and} \quad \mathcal{U}_{m,k}^c := \{4m \|A_{\bullet}^m\|_{\mathbb{L}(\ell_2)}^2 > k\}. \quad (08.19)$$

Note that $\mathbf{1}_{\mathcal{U}_{m,k}} = \mathbf{1}_{\{4m \|A_{\bullet}^m\|_{\mathbb{L}(\ell_2)}^2 \leq k\}}$ denotes an unobserved elementary random variable. \square

§08102.23 **Lemma.** Under Assumptions §08102.02 and §08102.18 for all $m, k, n \in \mathbb{N}$ we have

- (i) if $4\|\mathbf{M}_v \mathbf{T}_{\bullet}^{m|\dagger}\|_{\mathbb{HS}}^2 \leq k \wedge n$ then $\mathcal{U}_{m,k} \subseteq \Omega_{m,k \wedge n}$,
 - (ii) $\mathbb{P}_T^k(\text{tr}(\mathbf{M}_v \widehat{\mathbf{T}}_{\bullet}^{m|\dagger} \Gamma_{\theta|T} (\widehat{\mathbf{T}}_{\bullet}^{m|\dagger})^* \mathbf{M}_v) \mathbf{1}_{\Omega_{m,k \wedge n}}) \leq \mathfrak{v}_{\theta|T} (4\|\mathbf{M}_v \mathbf{T}_{\bullet}^{m|\dagger}\|_{\mathbb{HS}}^2 + (k \wedge n) \mathbb{P}_T^k(\mathcal{U}_{m,k}^c))$, and
 - (iii) $\mathbb{P}_T^k(\|\widehat{\mathbf{T}}_{\bullet}^{m|\dagger} (\mathbf{T}_{\bullet}^m - \widehat{\mathbf{T}}_{\bullet}^m) \theta^m\|_v^2 \mathbf{1}_{\Omega_{m,k \wedge n}}) \leq 4k^{-1} \mathfrak{v}_T \|\mathbf{M}_v \mathbf{T}_{\bullet}^{m|\dagger}\|_{\mathbb{HS}}^2 \|\theta^m\|_{\ell_2}^2 + \mathbb{P}_T^k(\|\mathfrak{r}_{\bullet}^m \theta^m\|_{\ell_2}^2 \mathbf{1}_{\mathcal{U}_{m,k}^c})$.
- with $\Omega_{m,k \wedge n}$ and $\mathcal{U}_{m,k}$ as in (08.18) and (08.19), respectively.

§08102.24 **Proof of Lemma §08102.23.** Given in the lecture. \square

§08102.25 **Proposition (Upper bound).** Under Assumptions §08102.02 and §08102.18 for all $n, m \in \mathbb{N}$ with (generalised) Galerkin solution $\theta^m = \mathbf{T}_{\bullet}^{m|\dagger} g \in \ell_2 \mathbf{1}^m$ setting similar to (08.03)

$$\begin{aligned} R_{n \wedge k}^m(\theta, \mathbf{T}_{\bullet}, \mathbf{v}) &:= \|\theta^m - \theta\|_v^2 + (n \wedge k)^{-1} \|\mathbf{M}_v \mathbf{T}_{\bullet}^{m|\dagger}\|_{\mathbb{HS}}^2, \\ m_{n \wedge k}^{\circ} &:= \arg \min \{R_{n \wedge k}^m(\theta, \mathbf{T}_{\bullet}, \mathbf{v}) : m \in \mathbb{N}\} \quad \text{and} \\ R_{n \wedge k}^{\circ}(\theta, \mathbf{T}_{\bullet}, \mathbf{v}) &:= R_{n \wedge k}^{m_{n \wedge k}^{\circ}}(\theta, \mathbf{T}_{\bullet}, \mathbf{v}) = \min \{R_{n \wedge k}^m(\theta, \mathbf{T}_{\bullet}, \mathbf{v}) : m \in \mathbb{N}\} \end{aligned} \quad (08.20)$$

for all $m \in \mathbb{N}$ the (generalised) tGE $\widehat{\theta}^m = \widehat{\mathbf{T}}_{\bullet}^{m|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbf{1}^m$ $\mathbb{P}_{\theta|T}^{n \otimes k}$ -a.s. satisfies

$$\begin{aligned} \mathbb{P}_{\theta|T}^{n \otimes k}(\|\widehat{\theta}^m - \theta\|_v^2) &\leq (4\mathfrak{v}_{\theta|T} + 8\mathfrak{v}_T \|\theta^m\|_{\ell_2}^2) R_{n \wedge k}^m(\theta, \mathbf{T}_{\bullet}, \mathbf{v}) \\ &\quad + \mathfrak{v}_{\theta|T} \mathbb{P}_T^k(\mathcal{U}_{m,k}^c) + 2\mathbb{P}_T^k(\|\mathfrak{r}_{\bullet}^m \theta^m\|_{\ell_2}^2 \mathbf{1}_{\mathcal{U}_{m,k}^c}) + \|\theta\|_v^2 \mathbb{P}_T^k(\Omega_{m,k \wedge n}^c) \end{aligned} \quad (08.21)$$

with $\Omega_{m,k \wedge n}$ and $\mathcal{U}_{m,k}$ as in (08.18) and (08.19), respectively.

§08102.26 **Proof of Proposition §08102.25.** Given in the lecture. \square

§08102.27 **Corollary.** Under the assumptions of **Proposition §08102.25** the (infeasible, generalised) tGE $\widehat{\theta}_{n \wedge k}^{\circ} = \widehat{\mathbf{T}}_{\bullet}^{m_{n \wedge k}^{\circ}|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbf{1}^{m_{n \wedge k}^{\circ}} \subseteq \ell_2(\mathbf{v}^2)$ $\mathbb{P}_{\theta|T}^{n \otimes k}$ -a.s. with $\Omega_{m_{n \wedge k}^{\circ}, k \wedge n}$ as in (08.18) and (infeasible) dimension $m_{n \wedge k}^{\circ}$ as in (08.20) for each $k, n \in \mathbb{N}$ with $R_{n \wedge k}^{\circ}(\theta, \mathbf{T}_{\bullet}, \mathbf{v}) \leq 1/4$ satisfies

$$\begin{aligned} \mathbb{P}_{\theta|T}^{n \otimes k}(\|\widehat{\theta}_{n \wedge k}^{\circ} - \theta\|_v^2) &\leq (4\mathfrak{v}_{\theta|T} + 8\mathfrak{v}_T \|\theta^{m_{n \wedge k}^{\circ}}\|_{\ell_2}^2) R_{n \wedge k}^{\circ}(\theta, \mathbf{T}_{\bullet}, \mathbf{v}) \\ &\quad + 2^{2l} K_T^{2l} \{ (\mathfrak{v}_{\theta|T} + \|\theta\|_v^2) k^{-1} m_{n \wedge k}^{\circ} \|\mathbf{T}_{\bullet}^{m_{n \wedge k}^{\circ}|}\|_{\mathbb{L}(\ell_2)}^2 + \|\theta^{m_{n \wedge k}^{\circ}}\|_{\mathbb{L}(\ell_2)}^2 / 4 \} \\ &\quad \times (m_{n \wedge k}^{\circ})^2 (k^{-1} (m_{n \wedge k}^{\circ})^3 \|\mathbf{T}_{\bullet}^{m_{n \wedge k}^{\circ}|}\|_{\mathbb{L}(\ell_2)}^2)^{l-1} \end{aligned} \quad (08.22)$$

and if in addition

$$(m_{n \wedge k}^\circ)^2 (k^{-1} (m_{n \wedge k}^\circ)^3 \|\mathbb{T}_{\cdot, \cdot}^{m_{n \wedge k}^\circ} \uparrow\|_{\mathbb{L}(\ell_2)}^2)^{l-1} \leq R_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v}) \leq 1/4 \quad (08.23)$$

then we have

$$\begin{aligned} \mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot}^{m_{n \wedge k}^\circ} - \theta_{\cdot}\|_{\mathbf{v}}^2) &\leq R_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v}) \\ &\times \left\{ (4 + 2^{2l} K_{\mathbb{T}}^{2l} (m_{n \wedge k}^\circ)^{-2}) \mathbb{v}_{\theta|\mathbb{T}} + (8 \mathbb{v}_{\mathbb{T}} + 2^{2l-2} K_{\mathbb{T}}^{2l}) \|\theta_{\cdot}^{m_{n \wedge k}^\circ}\|_{\ell_2}^2 + 2^{2l} K_{\mathbb{T}}^{2l} (m_{n \wedge k}^\circ)^{-2} \|\theta_{\cdot}\|_{\mathbf{v}}^2 \right\} \\ &\leq 2^{2l+2} K_{\mathbb{T}}^{2l} (\mathbb{v}_{\theta|\mathbb{T}} + \|\theta_{\cdot}^{m_{n \wedge k}^\circ}\|_{\ell_2}^2 + \|\theta_{\cdot}\|_{\mathbf{v}}^2) R_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v}) \quad (08.24) \end{aligned}$$

§08102.28 **Proof of Corollary §08102.27.** Given in the lecture. \square

§08102.29 **Remark.** Consider $m_{n \wedge k}^\circ = \arg \min \{R_{n \wedge k}^m(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v}) : m \in \mathbb{N}\}$ and $R_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v}) = R_{n \wedge k}^{m_{n \wedge k}^\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v})$ as in (08.20). Arguing similarly as in Remark §07101.21 we note that $\|\mathbb{M}_{\mathbf{v}} \mathbb{T}_{\cdot, \cdot}^{m \uparrow}\|_{\text{HS}} \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and hence $R_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v}) = o(1)$ as $n \wedge k \rightarrow \infty$, whenever $\|\theta_{\cdot}^m - \theta_{\cdot}\|_{\mathbf{v}} = o(1)$ as $m \rightarrow \infty$ (c.f. Remark §05101.05). In this situation if $\sup\{\|\theta_{\cdot}^m\|_{\ell_2}^2 : m \in \mathbb{N}\} \leq K_{\theta|\mathbb{T}}^2 \in \mathbb{R}_{>0}$ then from (08.24) in Corollary §08102.27 follows

$$\mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot}^{m_{n \wedge k}^\circ} - \theta_{\cdot}\|_{\mathbf{v}}^2) \leq 2^{2l+2} K_{\mathbb{T}}^{2l} (\mathbb{v}_{\theta|\mathbb{T}} + K_{\theta|\mathbb{T}}^2 + \|\theta_{\cdot}\|_{\mathbf{v}}^2) R_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v}).$$

However, the dimension $m_{n \wedge k}^\circ = m_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v})$ as defined in (08.03) depends on the unknown parameter of interest θ and the nuisance parameter $\mathbb{T}_{\cdot, \cdot}$, and thus also the statistic $\widehat{\theta}_{\cdot}^{m_{n \wedge k}^\circ}$. In other words $\widehat{\theta}_{\cdot}^{m_{n \wedge k}^\circ}$ is not a feasible estimator. \square

§08102.30 **Corollary** (GniSM with noisy operator §08102.08 continued). Consider independent noisy versions $(\widehat{g}_{\cdot}, \widehat{\mathbb{T}}_{\cdot, \cdot}) = (g_{\cdot} + n^{-1/2} \mathring{B}_{\cdot}, \mathbb{T}_{\cdot, \cdot} + k^{-1/2} \mathring{W}_{\cdot, \cdot}) \sim N_{\theta|\mathbb{T}}^{n \otimes k} = N_{\theta|\mathbb{T}}^n \otimes N_{\mathbb{T}}^k$ as in Model §08102.08, where $\mathring{B}_{\cdot} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ and $\mathring{W}_{\cdot, \cdot} \sim N_{(0,1)}^{\otimes \mathbb{N}^2}$ are independent, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}$ and $\theta_{\cdot} \in \ell_2$, and hence $g_{\cdot} = \mathbb{T}_{\cdot, \cdot} \theta_{\cdot} \in \text{dom}(\mathbb{T}_{\cdot, \cdot} \uparrow) \subseteq \ell_2$. Given Assumption §08102.18 for each $k, n \in \mathbb{N}$ fulfilling (08.23) the (infeasible, generalised) tGE $\widehat{\theta}_{\cdot}^{m_{n \wedge k}^\circ} = \widehat{\mathbb{T}}_{\cdot, \cdot}^{m_{n \wedge k}^\circ | (k \wedge n) \uparrow} \widehat{g}_{\cdot} \in \ell_2 \mathbb{1}_{m_{n \wedge k}^\circ} \subseteq \ell_2(\mathbf{v}^2)$ satisfies

$$N_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot}^{m_{n \wedge k}^\circ} - \theta_{\cdot}\|_{\mathbf{v}}^2) \leq 2^{2l+2} ((2l-1)!!) (1 + \|\theta_{\cdot}^{m_{n \wedge k}^\circ}\|_{\ell_2}^2 + \|\theta_{\cdot}\|_{\mathbf{v}}^2) R_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v})$$

where $R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v})$ is the oracle rate in a GniSM §08101.04 (see Corollary §08101.21).

§08102.31 **Proof of Corollary §08102.30.** Given in the lecture. \square

§08102.32 **Corollary** (niSM with noisy operator §08102.10 continued). Consider independent noisy versions $(\widehat{g}_{\cdot}, \widehat{\mathfrak{s}}_{\cdot}) = (g_{\cdot} + n^{-1/2} \mathring{\epsilon}_{\cdot}, \mathbb{T}_{\cdot, \cdot} + k^{-1/2} \mathring{\eta}_{\cdot, \cdot}) \sim P_{\theta|\mathbb{T}|\sigma|\xi|\xi^{(2l)}}$ as in Model §08102.10, where $\mathring{\epsilon}_{\cdot}$ and $\mathring{\eta}_{\cdot, \cdot}$ satisfies (iSM1) with $\mathbb{v}_{\sigma} = \|\sigma_{\cdot}^2\|_{\ell_{\infty}} \vee 1$ and (niSMnO1)–(niSMnO2) with $K_{\xi^{(2l)}}^{2l} := 1 \vee \|\xi_{\cdot}^{(2l)}\|_{\ell_{\infty}(\mathbb{N}^2)}$, respectively, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}$ and $\theta_{\cdot} \in \ell_2$, and hence $g_{\cdot} = \mathbb{T}_{\cdot, \cdot} \theta_{\cdot} \in \text{dom}(\mathbb{T}_{\cdot, \cdot} \uparrow) \subseteq \ell_2$. Given Assumption §08102.18 for each $k, n \in \mathbb{N}$ fulfilling (08.23) the (infeasible, generalised) tGE $\widehat{\theta}_{\cdot}^{m_{n \wedge k}^\circ} = \widehat{\mathbb{T}}_{\cdot, \cdot}^{m_{n \wedge k}^\circ | (k \wedge n) \uparrow} \widehat{g}_{\cdot} \in \ell_2 \mathbb{1}_{m_{n \wedge k}^\circ} \subseteq \ell_2(\mathbf{v}^2)$ satisfies

$$P_{\theta|\mathbb{T}|\sigma|\xi|\xi^{(2l)}}^{n \otimes k} (\|\widehat{\theta}_{\cdot}^{m_{n \wedge k}^\circ} - \theta_{\cdot}\|_{\mathbf{v}}^2) \leq 2^{2l+2} K_{\xi^{(2l)}}^{2l} (\mathbb{v}_{\sigma} + \|\theta_{\cdot}^{m_{n \wedge k}^\circ}\|_{\ell_2}^2 + \|\theta_{\cdot}\|_{\mathbf{v}}^2) R_{n \wedge k}^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v})$$

where $R_n^\circ(\theta, \mathbb{T}_{\cdot, \cdot}, \mathbf{v})$ is the oracle rate in a niSM §08101.06 (see Corollary §08101.23).

§08102.33 **Proof of Corollary §08102.32.** Given in the lecture. \square

§08102.34 **Corollary** (nieMM with noisy operator §08102.13 continued). Consider independent noisy versions $(\widehat{g}_\bullet, \widehat{T}_{\bullet,\bullet}) = (g_\bullet + n^{-1/2}\dot{\epsilon}_\bullet, T_{\bullet,\bullet} + k^{-1/2}\dot{\eta}_{\bullet,\bullet})$ defined on $(\mathcal{Z}^{n+k}, \mathcal{X}^{\otimes(n+k)}, \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k})$ as in Model §08102.13, where $\psi_\bullet \in \mathcal{M}(\mathcal{X} \otimes 2^{\mathbb{N}})$ and $\varphi_\bullet \in \mathcal{M}(\mathcal{X} \otimes 2^{\mathbb{N}^2})$ satisfies (nieMM1)-(nieMM2) for $\nu_{\theta|\mathbb{T}|\psi} \in \mathbb{R}_{\geq 1}$ and (nieMMnO1)-(nieMMnO3) for $K_{\mathbb{T}|\varphi} \in \mathbb{R}_{\geq 1}$, respectively, $T_{\bullet,\bullet} \in \mathbb{T}$ and $\theta_\bullet \in \ell_2$, and hence $g_\bullet = T_{\bullet,\bullet}\theta_\bullet \in \text{dom}(T_{\bullet,\bullet}^\dagger) \subseteq \ell_2$. Given Assumption §08102.18 for each $k, n \in \mathbb{N}$ fulfilling (08.23) the (infeasible, generalised) tGE $\widehat{\theta}_{n \wedge k}^{m_{n \wedge k}^\circ} = \widehat{T}_{\bullet,\bullet}^{m_{n \wedge k}^\circ | (k \wedge n) | \dagger} \widehat{g}_\bullet \in \ell_2 \mathbf{1}_{m_{n \wedge k}^\circ} \subseteq \ell_2(\nu_\bullet^2)$ satisfies

$$\mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{n \wedge k}^{m_{n \wedge k}^\circ} - \theta_\bullet\|_{\nu_\bullet}^2) \leq C_{2l} K_{\mathbb{T}|\varphi}^{2l} (\nu_{\theta|\mathbb{T}|\psi} + \|\theta_\bullet^{m_{n \wedge k}^\circ}\|_{\ell_2}^2 + \|\theta_\bullet\|_{\nu_\bullet}^2) R_{n \wedge k}^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet)$$

where $C_{2l} \in \mathbb{R}_{\geq 1}$ is a constant depending on $l \in \mathbb{N}$ only and $R_n^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet)$ is the oracle rate in a nieMM §08101.08 (see Corollary §08101.25).

§08102.35 **Proof of Corollary** §08102.34. Given in the lecture. □

§08102.36 **Illustration**. We distinguish as in Illustration §08101.27 the two cases **(p)** and **(np)**, where $\theta_\bullet \mathbf{1}_{\bullet}^{K_{1 \perp}} = 0$ implies the case **(p)**. In case **(p)** the oracle bound is parametric, that is, $nR_n^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet) = O(1)$, in case **(np)** the oracle bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet) = \infty$. In case **(np)** consider similar to **(o-m)**, **(o-s)** and **(s-m)** in Illustration §08101.27 the following specifications:

Table 05 [§08]

Order of the rate $R_{n \wedge k}^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet)$ as $n \wedge k \rightarrow \infty$

	(squared bias)	(variance)	$m_{n \wedge k}^\circ$	$R_{n \wedge k}^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet)$	
$(m \in \mathbb{N})$ $(\nu_m = m^\nu)$	$\ \theta_\bullet^m - \theta_\bullet\ _{\nu_\bullet}^2$ ($a \in \mathbb{R}_{>0}$)	$\ \mathbb{M}_\nu T_{\bullet,\bullet}^{m \dagger}\ _{\text{HS}}^2$ ($t \in \mathbb{R}_{>0}$)			
(o-m)	$\nu \in (-1/2 - t, a)$ $\nu + t = -1/2$	$m^{-2(a-\nu)}$ $m^{-2a-2t-1}$	$m^{2(t+\nu)+1}$ $\log m$	$(n \wedge k)^{-\frac{1}{2a+2t+1}}$ $(\frac{n \wedge k}{\log n \wedge k})^{-\frac{1}{2a+2t+1}}$	$(n \wedge k)^{-\frac{2(a-\nu)}{2a+2t+1}}$ $\frac{\log n \wedge k}{n \wedge k}$
(o-s)	$a - \nu \in \mathbb{R}_{>0}$	$m^{-2(a-\nu)}$	$m^{(1-2(t-\nu))+e^{m^{2t}}}$	$(\log n \wedge k)^{\frac{1}{2t}}$	$(\log n \wedge k)^{-\frac{a-\nu}{t}}$
(s-m)	$\nu + t + 1/2 \in \mathbb{R}_{>0}$ $\nu + t = -1/2$	$m^{(1-2(a-\nu))+e^{-m^{2a}}}$ $e^{-m^{2a}}$	$m^{2(t+\nu)+1}$ $\log m$	$(\log n \wedge k)^{\frac{1}{2a}}$ $(\log n \wedge k)^{\frac{1}{2a}}$	$\frac{(\log n \wedge k)^{\frac{2t+2\nu+1}{2a}}}{n \wedge k}$ $\frac{\log \log n \wedge k}{n \wedge k}$

We note that in case **(o-m)** and **(s-m)** for $\nu + t < -1/2$ the rate $R_{n \wedge k}^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet)$ is parametric. The tGE attains the rate $R_{n \wedge k}^\circ := R_{n \wedge k}^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet)$ due to Corollary §08102.27 under the additional condition

$$(k^{-1}(m_{n \wedge k}^\circ)^3 \|\mathbb{T}_{\bullet,\bullet}^{m_{n \wedge k}^\circ | \dagger}\|_{\mathbb{L}(\ell_2)}^2)^{l-1} \leq (m_{n \wedge k}^\circ)^{-2} R_{n \wedge k}^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet). \tag{08.25}$$

Since $(m_{n \wedge k}^\circ)^{-2} R_{n \wedge k}^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet) = o(1)$ also $k^{-1}(m_{n \wedge k}^\circ)^3 \|\mathbb{T}_{\bullet,\bullet}^{m_{n \wedge k}^\circ | \dagger}\|_{\mathbb{L}(\ell_2)}^2 = o(1)$ is necessary as $n \wedge k \rightarrow \infty$. The next table depicts the order of both terms in case **(o-m)**, **(o-s)** and **(s-m)**.

Table 06 [§08]

Order as $n \wedge k \rightarrow \infty$

	(o-m)	(o-s)	(s-m)
	$\nu \in (-1/2 - t, a)$	$a - \nu \in \mathbb{R}_{>0}$	$\nu + t + 1/2 \in \mathbb{R}_{>0}$
$(m_{n \wedge k}^\circ)^{-2} R_{n \wedge k}^\circ(\theta_\bullet, T_{\bullet,\bullet}, \nu_\bullet)$	$(n \wedge k)^{-\frac{2(a-\nu)+2}{2a+2t+1}}$	$(\log n \wedge k)^{-\frac{2a-2\nu+2}{2t}}$	$\frac{(\log n \wedge k)^{\frac{2t+2\nu-1}{2a}}}{n \wedge k}$
$(n \wedge k)^{-1} (m_{n \wedge k}^\circ)^3 \ \mathbb{T}_{\bullet,\bullet}^{m_{n \wedge k}^\circ \dagger}\ _{\mathbb{L}(\ell_2)}^2$	$(n \wedge k)^{-\frac{2a-2}{2a+2t+1}}$	$(n \wedge k)^{-c}$	$\frac{(\log n \wedge k)^{\frac{2t+3}{2a}}}{n \wedge k}$

In case **(o-s)** a value $l \geq 2$ and **(s-m)** a value $l \geq 3$ is sufficient to ensure (08.25) as $n \wedge k \rightarrow \infty$. In case **(o-m)** assuming $a > 1$ we have $k^{-1}(m_{n \wedge k}^\circ)^3 \|\mathbb{T}_{\bullet,\bullet}^{m_{n \wedge k}^\circ | \dagger}\|_{\mathbb{L}(\ell_2)}^2 = o(1)$ as $n \wedge k \rightarrow \infty$. In this situation we have (08.25) if $2(a-1)(l-1) > 2(a-\nu) + 2$ or in equal $l > (2a-\nu)/(a-1)$. □

§08|02|02|02 Maximal global v-risk

§08|02.37 **Notation (Reminder).** For sequences $a_n, b_n \in (\mathbb{K})^{\mathbb{N}}$ taking its values in $\mathbb{K} \in \{\mathbb{R}, \mathbb{R}_{>0}, \mathbb{Z}, \dots\}$ we write $a_n \in (\mathbb{K})_{\nearrow}^{\mathbb{N}}$ and $b_n \in (\mathbb{K})_{\searrow}^{\mathbb{N}}$ if a_n and b_n , respectively, is monotonically *non-decreasing* and *non-increasing*. If in addition $a_n \rightarrow \infty$ and $b_n \rightarrow 0$ as $n \rightarrow \infty$, then we write $a_n \in (\mathbb{K})_{\nearrow\infty}^{\mathbb{N}}$ and $b_n \in (\mathbb{K})_{\searrow 0}^{\mathbb{N}}$ for short. For $w_n \in \ell_\infty$ we set $w_{(0)} := \|w_n\|_{\ell_\infty}$ and $w_{(\cdot)} = (w_{(j)} := \|w_n \mathbf{1}_n^{j \perp}\|_{\ell_\infty})_{j \in \mathbb{N}}$, where by construction $w_{(\cdot)} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$. \square

§08|02.38 **Assumption.** Consider weights $t_n, a_n \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and $v_n \in \mathbb{R}_{>0}^{\mathbb{N}}$ such that $(av)_n := a_n v_n \in \ell_\infty$, $(av)_{(\cdot)} \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$, and $(t/v)_n = t_n v_n^{-1} \in \ell_\infty$ are satisfied. In addition there exists $C_{(t/v)} \in (0, 1]$ such that for all $m \in \mathbb{N}$

$$(t/v)_{(m-1)}^2 \geq \min \{ (t/v)_j^2 : j \in \llbracket m \rrbracket \} \geq C_{(t/v)} (t/v)_{(m)}^2 \quad (08.26)$$

or in equal $C_{(t/v)} \|(t/v)_n^{-2} \mathbf{1}_n^m\|_{\ell_\infty} \leq (t/v)_{(m)}^{-2}$. \square

§08|02.39 **Reminder.** Under Assumption §08|02.38 we have $\ell_2^a = \text{dom}(M_{a_n}) = \ell_2 a_n \subseteq \ell_2$ and the three measures ν_N , $a_n^{-2} \nu_N$ and $v_n^2 \nu_N$ dominate mutually each other, i.e. they share the same null sets (see **Property** §04|01.02). We consider ℓ_2^a endowed with $\|\cdot\|_{a_n^{-1}} = \|M_{a_n} \cdot\|_{\ell_2}$ and given a constant $r \in \mathbb{R}_{>0}$ the ellipsoid $\ell_2^{a,r} := \{a_n \in \ell_2^a : \|a_n\|_{a_n^{-1}} \leq r\} \subseteq \ell_2^a$. Since $(av)_n \in \ell_\infty$, and hence $(av)_{(m)} := \|(av)_n \mathbf{1}_n^{m \perp}\|_{\ell_\infty} \in \mathbb{R}_{>0}$ for each $m \in \mathbb{N}$ we have $\ell_2^a \subseteq \ell_2(v_n^2)$ (**Property** §04|02.11). Consequently, if Assumption §08|02.38 and $\theta_n \in \ell_2^{a,r}$ are satisfied, then Assumption §08|02.18 is also fulfilled. Since $v_n, t_n \in \mathbb{R}_{>0}^{\mathbb{N}}$ under Assumption §08|02.38, we have $\|t_n^{-1} \mathbf{1}_n^m\|_{v_n} = \|(v_n/t_n) \mathbf{1}_n^m\|_{\ell_2} \in \mathbb{R}_{>0}$ for each $m \in \mathbb{N}$. Under the Assumptions §08|00.02 and §08|02.38 considering the generalised link condition $T_{\cdot, \cdot} \in \mathbb{T}_{t,d,D}$ with band $D \in [1, \infty)$ and $d \in [1, D]$ as in **Definition** §05|02.05 we have $\sup_{m \in \mathbb{N}} \{ \|[M_{t_n}]_{\llbracket m \rrbracket} [T_{\cdot, \cdot}]_{\llbracket m \rrbracket}^{-1}\|_{\text{spec}} \} \leq D$, and hence $\|T_{\cdot, \cdot}^{m \dagger}\|_{\mathbb{L}(\ell_2)} \leq D t_n^{-1}$ and $\|M_{v_n} T_{\cdot, \cdot}^{m \dagger}\|_{\text{HS}} \leq D \|t_n^{-1} \mathbf{1}_n^m\|_{v_n}$ shown in (08.06) using $\text{tr}([M_{(v/t)_n}^m]_{\llbracket m \rrbracket}^2) = \|(v/t)_n \mathbf{1}_n^m\|_{\ell_2}^2 = \|t_n^{-1} \mathbf{1}_n^m\|_{v_n}^2$. Moreover, for each $m \in \mathbb{N}$ the generalised Galerkin solution $\theta_n^m := T_{\cdot, \cdot}^{m \dagger} g_n \in \ell_2 \mathbf{1}_n^m$ of $\theta_n = T_{\cdot, \cdot}^\dagger g_n \in \ell_2^{a,r}$ satisfies (**Lemma** §05|02.09)

$$\|\theta_n^m\|_{\ell_2} \leq a_n \|\theta_n^m\|_{a_n^{-1}} \leq a_n D d r \quad \text{and} \quad \|\theta_n - \theta_n^m\|_{v_n}^2 \leq (D^2 d^2 C_{(t/v)}^{-2} + 1) (av)_{(m)}^2 r^2.$$

Note that under Assumptions §08|00.02 and §08|02.38 the link condition $T_{\cdot, \cdot} \in \mathbb{T}_{t,d}^{\geq}$ with band $d \in \mathbb{R}_{>1}$ as in **Definition** §05|01.08 implies $\sup_{m \in \mathbb{N}} \{ \|[M_{t_n}]_{\llbracket m \rrbracket} [T_{\cdot, \cdot}]_{\llbracket m \rrbracket}^{-1}\|_{\text{spec}} \} \leq 3d^2$ (**Lemma** §05|01.22), and hence for each $m \in \mathbb{N}$ we have $\|M_{v_n} T_{\cdot, \cdot}^{m \dagger}\|_{\text{HS}}^2 \leq 9d^4 \|t_n^{-1} \mathbf{1}_n^m\|_{v_n}^2$ and the Galerkin solution $\theta_n^m := T_{\cdot, \cdot}^{m \dagger} g_n \in \ell_2 \mathbf{1}_n^m$ of $\theta_n = T_{\cdot, \cdot}^\dagger g_n \in \ell_2^{a,r}$ satisfies $\|\theta_n^m\|_{\ell_2} \leq 3a_n d^3 r$ and $\|\theta_n - \theta_n^m\|_{v_n}^2 \leq (9d^6 C_{(t/v)}^{-2} + 1) (av)_{(m)}^2 r^2$ (**Lemma** §05|01.28). \square

§08|02.40 **Corollary.** Under Assumptions §08|02.02 and §08|02.38 let $\theta_n := T_{\cdot, \cdot}^\dagger g_n \in \ell_2^{a,r}$ and $T_{\cdot, \cdot} \in \mathbb{T}_{t,d,D}$ (or $T_{\cdot, \cdot} \in \mathbb{T}_{t,d}^{\geq}$ with $D = 3d^2$), for all $m, k, n \in \mathbb{N}$ we have

- (i) if $4D^2 \|t_n^{-1} \mathbf{1}_n^m\|_{v_n}^2 \leq k \wedge n$ then $\mathcal{U}_{m,k} \subseteq \Omega_{m,k \wedge n}$,
 - (ii) $\mathbb{E}_T^k (\text{tr}(M_{v_n} \widehat{T}_{\cdot, \cdot}^{m \dagger} \Gamma_{\theta|T} (\widehat{T}_{\cdot, \cdot}^{m \dagger})^* M_{v_n}) \mathbf{1}_{\Omega_{m,k \wedge n}}) \leq v_{\theta|T} (4D^2 \|t_n^{-1} \mathbf{1}_n^m\|_{v_n}^2 + (k \wedge n) \mathbb{E}_T^k (\mathcal{U}_{m,k}^c))$, and
 - (iii) $\mathbb{E}_T^k (\|\widehat{T}_{\cdot, \cdot}^{m \dagger} (T_{\cdot, \cdot}^m - \widehat{T}_{\cdot, \cdot}^m) \theta_n^m\|_{\ell_2}^2 \mathbf{1}_{\Omega_{m,k \wedge n}}) \leq 4k^{-1} v_T D^2 \|t_n^{-1} \mathbf{1}_n^m\|_{v_n}^2 \|\theta_n^m\|_{\ell_2}^2 + \mathbb{E}_T^k (\|\dot{\eta}_{\cdot, \cdot}^m \theta_n^m\|_{\ell_2}^2 \mathbf{1}_{\Omega_{m,k}^c})$.
- with $\Omega_{m,k \wedge n}$ and $\mathcal{U}_{m,k}$ as in (08.18) and (08.19), respectively.

§08|02.41 **Proof of Corollary** §08|02.40. Given in the lecture. \square

§08|02.42 **Proposition (Upper bound).** Under Assumptions §08|02.02 and §08|02.38 let $\theta_n := T_{\cdot, \cdot}^\dagger g_n \in \ell_2^{a,r}$ and

$\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t, d, D}$ (or $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t, d}^{\geq}$ with $D = 3d^2$) for all $n, m \in \mathbb{N}$ setting similar to (08.07)

$$\begin{aligned} \mathbf{R}_{n \wedge k}^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) &:= [(\mathbf{a}\mathbf{v})_{(m)}^2 \vee (n \wedge k)^{-1} \|\mathbf{t}^{-1} \mathbf{1}^m\|_{\mathbf{v}}^2], \\ m_{n \wedge k}^* &:= \arg \min \{ \mathbf{R}_{n \wedge k}^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) : m \in \mathbb{N} \} \quad \text{and} \\ \mathbf{R}_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) &:= \mathbf{R}_{n \wedge k}^{m_{n \wedge k}^*}(\mathbf{a}, \mathbf{t}, \mathbf{v}) = \min \{ \mathbf{R}_{n \wedge k}^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) : m \in \mathbb{N} \} \end{aligned} \quad (08.27)$$

for all $m \in \mathbb{N}$ the (generalised) tGE $\widehat{\theta}_{\cdot, \cdot}^m = \widehat{\mathbb{T}}_{\cdot, \cdot}^{m|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbf{1}^m \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k}$ -a.s. satisfies

$$\begin{aligned} \mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot, \cdot}^m - \theta\|_{\mathbf{v}}^2) &\leq 2D^2 (C_{(t/v)}^{-2} d^2 r^2 + 2\nu_{\theta|\mathbb{T}} + 4\nu_{\mathbb{T}} \alpha_1^2 D^2 d^2 r^2) \mathbf{R}_{n \wedge k}^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) \\ &\quad + \nu_{\theta|\mathbb{T}} \mathbb{E}_{\mathbb{T}}^k (\mathcal{U}_{m, k}^c) + 2\mathbb{E}_{\mathbb{T}}^k (\|\widehat{\mathbf{n}}_{\cdot, \cdot}^m \theta^m\|_{\ell_2}^2 \mathbf{1}_{\mathcal{U}_{m, k}^c}^c) + \|\theta\|_{\mathbf{v}}^2 \mathbb{E}_{\mathbb{T}}^k (\Omega_{m, k \wedge n}^c) \end{aligned} \quad (08.28)$$

with $\Omega_{m, k \wedge n}$ and $\mathcal{U}_{m, k}$ as in (08.18) and (08.19), respectively.

§08102.43 **Proof of Proposition** §08102.42. Given in the lecture. \square

§08102.44 **Corollary.** Under the assumptions of **Proposition** §08102.42 for $k, n \in \mathbb{N}$ with $\mathbf{R}_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \leq 1/(4D^2)$ the (generalised) tGE $\widehat{\theta}_{\cdot, \cdot}^{m_{n \wedge k}^*} = \widehat{\mathbb{T}}_{\cdot, \cdot}^{m_{n \wedge k}^*|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbf{1}^{m_{n \wedge k}^*} \subseteq \ell_2(\mathbf{v}^?) \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k}$ -a.s. with $\Omega_{m_{n \wedge k}^*, k \wedge n}$ as in (08.18) and dimension $m_{n \wedge k}^*$ as in (08.27) satisfies

$$\begin{aligned} \mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot, \cdot}^{m_{n \wedge k}^*} - \theta\|_{\mathbf{v}}^2) &\leq 2D^2 (C_{(t/v)}^{-2} d^2 r^2 + 2\nu_{\theta|\mathbb{T}} + 4\nu_{\mathbb{T}} \alpha_1^2 D^2 d^2 r^2) \mathbf{R}_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \\ &\quad + 2^{2l} K_{\mathbb{T}}^{2l} D^{2(l-1)} \{ D^2 (\nu_{\theta|\mathbb{T}} + (\mathbf{a}\mathbf{v})_{(0)}^2 r^2) k^{-1} m_{n \wedge k}^* \mathbf{t}_{m_{n \wedge k}^*}^{-2} + \alpha_1^2 D^2 d^2 r^2 / 4 \} \\ &\quad \times (m_{n \wedge k}^*)^2 (k^{-1} (m_{n \wedge k}^*)^3 \mathbf{t}_{m_{n \wedge k}^*}^{-2})^{l-1}. \end{aligned} \quad (08.29)$$

and if in addition

$$(m_{n \wedge k}^*)^2 (k^{-1} (m_{n \wedge k}^*)^3 \mathbf{t}_{m_{n \wedge k}^*}^{-2})^{l-1} \leq \mathbf{R}_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \leq 1/(4D^2) \quad (08.30)$$

then we have

$$\mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot, \cdot}^{m_{n \wedge k}^*} - \theta\|_{\mathbf{v}}^2) \leq 2^{2l+2} K_{\mathbb{T}}^{2l} D^{2l} (\nu_{\theta|\mathbb{T}} + (\mathbf{a}\mathbf{v})_{(0)}^2 r^2 + (C_{(t/v)}^{-2} + \alpha_1^2) d^2 r^2) \mathbf{R}_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}). \quad (08.31)$$

§08102.45 **Proof of Corollary** §08102.44. Given in the lecture. \square

§08102.46 **Remark.** Arguing similarly as in **Remark** §07101.21 we note that $\|\mathbf{t}^{-1} \mathbf{1}^m\|_{\mathbf{v}}^2 \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and hence $\mathbf{R}_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) = o(1)$ as $n \wedge k \rightarrow \infty$, whenever $(\mathbf{a}\mathbf{v})_{(m)} = o(1)$ as $m \rightarrow \infty$, i.e. $(\mathbf{a}\mathbf{v})_{\cdot} \in (\mathbb{R}_{>0})_{\neq 0}^{\mathbb{N}}$. If there is in addition $C \in \mathbb{R}_{\geq 1}$ such that $K_{\mathbb{T}}^{2l} \leq C$ and $\nu_{\theta|\mathbb{T}} \leq C$ for all $\theta_{\cdot} := \mathbb{T}_{\cdot, \cdot}^{\dagger} g_{\cdot} \in \ell_2^{\alpha, r}$ and $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t, d, D}$ then from the bound (08.31) **Corollary** §08102.44 follows immediately

$$\begin{aligned} \sup \{ \mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot, \cdot}^{m_{n \wedge k}^*} - \theta\|_{\mathbf{v}}^2) : \mathbb{T}_{\cdot, \cdot} \in \mathbb{T}_{t, d, D}, \theta_{\cdot} \in \ell_2^{\alpha, r} \} &\leq \mathbf{R}_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \\ &\quad \times 2^{2l+2} C D^{2l} (C + (\mathbf{a}\mathbf{v})_{(0)}^2 r^2 + (C_{(t/v)}^{-2} + \alpha_1^2) d^2 r^2). \end{aligned}$$

Note that the dimension $m_{n \wedge k}^* := m_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ does not depend on the unknown parameter of interest θ but on the classes $\ell_2^{\alpha, r}$ and $\mathbb{T}_{t, d, D}$ only, and thus also the statistic $\widehat{\theta}_{\cdot, \cdot}^{m_{n \wedge k}^*}$. In other words, if the regularity of θ and $\mathbb{T}_{\cdot, \cdot}$ is known in advance, then the thresholded GE $\widehat{\theta}_{\cdot, \cdot}^{m_{n \wedge k}^*}$ is a feasible estimator. \square

§08102.47 **Corollary** (GniSM with noisy operator §08102.08 continued). Consider independent noisy versions $(\widehat{g}_{\cdot}, \widehat{\mathbb{T}}_{\cdot, \cdot}) = (g_{\cdot} + n^{-1/2} \dot{B}_{\cdot}, \mathbb{T}_{\cdot, \cdot} + k^{-1/2} \dot{W}_{\cdot}) \sim N_{\theta|\mathbb{T}}^{n \otimes k} = N_{\theta|\mathbb{T}}^n \otimes N_{\mathbb{T}}^k$ as in **Model** §08102.08, where $\dot{B}_{\cdot} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{W}_{\cdot} \sim N_{(0,1)}^{\otimes \mathbb{N}^2}$ are independent, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}$ and $\theta_{\cdot} \in \ell_2$, and hence $g_{\cdot} = \mathbb{T}_{\cdot, \cdot} \theta_{\cdot} \in$

$\text{dom}(\mathbb{T}_{\cdot,\cdot}^\dagger) \subseteq \ell_2$. Given Assumption §08102.38 for each $k, n \in \mathbb{N}$ fulfilling (08.30) the (generalised) tGE $\widehat{\theta}_{\cdot,\cdot}^{m_{n,k}^*} = \widehat{\mathbb{T}}_{\cdot,\cdot}^{m_{n,k}^*|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbb{1}_{\cdot,\cdot}^{m_{n,k}^*} \subseteq \ell_2(\mathbf{v}^?)$ satisfies

$$\sup \left\{ \mathbb{N}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot,\cdot}^{m_{n,k}^*} - \theta\|_{\mathbf{v}}^2) : \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D}, \theta \in \ell_2^{a,r} \right\} \leq R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \\ \times 2^{2l+2} ((2l-1)!) D^{2l} (1 + (\mathbf{a}\mathbf{v})_{(0)}^2 \mathbf{r}^2 + (C_{(t/v)}^{-2} + \mathbf{a}_1^2) d^2 \mathbf{r}^2)$$

where $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ is the rate in a GniSM §08101.04 (see *Corollary* §08101.34).

§08102.48 **Proof of Corollary** §08102.47. Given in the lecture. □

§08102.49 **Corollary** (niSM with noisy operator §08102.10 continued). Consider independent noisy versions $(\widehat{g}_\cdot, \widehat{\mathfrak{s}}_\cdot) = (g_\cdot + n^{-1/2} \dot{\varepsilon}_\cdot, \mathbb{T}_{\cdot,\cdot} + k^{-1/2} \dot{\eta}_{\cdot,\cdot}) \sim \mathbb{P}_{\theta|\mathbb{T}|\sigma|\xi|\xi^{(2l)}}$ as in Model §08102.10, where $\dot{\varepsilon}_\cdot$ and $\dot{\eta}_{\cdot,\cdot}$ satisfies (iSM1) with $\mathbf{v}_\sigma = \|\sigma^2\|_{\ell_\infty} \vee 1$ and (niSMnO1)–(niSMnO2) with $K_{\xi^{(2l)}}^{2l} := 1 \vee \|\xi_{\cdot,\cdot}^{(2l)}\|_{\ell_\infty(\mathbb{N}^?)}$, respectively, $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}$ and $\theta \in \ell_2$, and hence $g_\cdot = \mathbb{T}_{\cdot,\cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot,\cdot}^\dagger) \subseteq \ell_2$. Given Assumption §08102.38 for each $k, n \in \mathbb{N}$ fulfilling (08.30) the (generalised) tGE $\widehat{\theta}_{\cdot,\cdot}^{m_{n,k}^*} = \widehat{\mathbb{T}}_{\cdot,\cdot}^{m_{n,k}^*|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbb{1}_{\cdot,\cdot}^{m_{n,k}^*} \subseteq \ell_2(\mathbf{v}^?)$ satisfies

$$\sup \left\{ \mathbb{P}_{\theta|\mathbb{T}|\sigma|\xi|\xi^{(2l)}}^{n \otimes k} (\|\widehat{\theta}_{\cdot,\cdot}^{m_{n,k}^*} - \theta\|_{\mathbf{v}}^2) : \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D}, \theta \in \ell_2^{a,r} \right\} \leq R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \\ \times 2^{2l+2} K_{\xi^{(2l)}}^{2l} D^{2l} (\mathbf{v}_\sigma + (\mathbf{a}\mathbf{v})_{(0)}^2 \mathbf{r}^2 + (C_{(t/v)}^{-2} + \mathbf{a}_1^2) d^2 \mathbf{r}^2)$$

where $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ is the rate in a niSM §08101.06 (see *Corollary* §08101.36).

§08102.50 **Proof of Corollary** §08102.49. Given in the lecture. □

§08102.51 **Corollary** (nieMM with noisy operator §08102.13 continued). Consider independent noisy versions $(\widehat{g}_\cdot, \widehat{\mathbb{T}}_{\cdot,\cdot}) = (g_\cdot + n^{-1/2} \dot{\varepsilon}_\cdot, \mathbb{T}_{\cdot,\cdot} + k^{-1/2} \dot{\eta}_{\cdot,\cdot})$ defined on $(\mathcal{Z}^{n+k}, \mathcal{X}^{\otimes(n+k)}, \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k})$ as in Model §08102.13, where $\psi_\cdot \in \mathcal{M}(\mathcal{X} \otimes 2^{\mathbb{N}})$ and $\varphi_{\cdot,\cdot} \in \mathcal{M}(\mathcal{X} \otimes 2^{\mathbb{N}^2})$ satisfies (nieMM1)–(nieMM2) for $\mathbf{v}_{\theta|\mathbb{T}|\psi} \in \mathbb{R}_{\geq 1}$ and (nieMMnO1)–(nieMMnO3) for $K_{\mathbb{T}|\varphi} \in \mathbb{R}_{\geq 1}$, respectively, $\mathbb{T}_{\cdot,\cdot} \in \mathbb{T}$ and $\theta \in \ell_2$, and hence $g_\cdot = \mathbb{T}_{\cdot,\cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot,\cdot}^\dagger) \subseteq \ell_2$. Given Assumption §08102.38 for each $k, n \in \mathbb{N}$ fulfilling (08.30) the (generalised) tGE $\widehat{\theta}_{\cdot,\cdot}^{m_{n,k}^*} = \widehat{\mathbb{T}}_{\cdot,\cdot}^{m_{n,k}^*|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbb{1}_{\cdot,\cdot}^{m_{n,k}^*} \subseteq \ell_2(\mathbf{v}^?)$ satisfies

$$\sup \left\{ \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k} (\|\widehat{\theta}_{\cdot,\cdot}^{m_{n,k}^*} - \theta\|_{\mathbf{v}}^2) : \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D}, \theta \in \ell_2^{a,r} \right\} \leq R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \\ \times C_{2l} \sup \left\{ K_{\mathbb{T}|\varphi}^{2l} : \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D} \right\} D^{2l} (\sup \left\{ \mathbf{v}_{\theta|\mathbb{T}|\psi} : \mathbb{T}_{\cdot,\cdot} \in \mathbb{T}_{t,d,D}, \theta \in \ell_2^{a,r} \right\} + (\mathbf{a}\mathbf{v})_{(0)}^2 \mathbf{r}^2 + (C_{(t/v)}^{-2} + \mathbf{a}_1^2) d^2 \mathbf{r}^2)$$

where $C_{2l} \in \mathbb{R}_{\geq 1}$ is a constant depending on $l \in \mathbb{N}$ only and $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ is the rate in a nieMM §08101.08 (see *Corollary* §08101.38).

§08102.52 **Proof of Corollary** §08102.51. Given in the lecture. □

§08102.53 **Illustration**. We distinguish again the two cases **(p)** and **(np)** given in *Illustration* §08101.40 where in case **(p)** the bound in *Corollary* §08102.44 is parametric, that is, $(n \wedge k) R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) = O(1)$, in case **(np)** the bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} (n \wedge k) R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) = \infty$. In case **(np)** we consider similar to **(o-m)**, **(o-s)** and **(s-m)** in *Illustration* §07101.44 the following three specifications:

Table 07 [§08]

Order of the rate $R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ as $n \wedge k \rightarrow \infty$

$(j \in \mathbb{N})$ $v_j^2 = j^{2v}$	$(\mathbf{a} \in \mathbb{R}_{>0})$ ($\mathbf{t} \in \mathbb{R}_{>0}$) α_j^2 \mathfrak{t}_j^2	(squared bias) $(\mathbf{a}\mathbf{v})_{(m)}^2$	(variance) $\ \mathfrak{t}_\cdot^{-1} \mathbf{1}_m\ _{\mathbf{v}}^2$	$m_{n \wedge k}^*$	$R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$
(o-m) $v \in (-1/2 - t, a)$	j^{-2a} j^{-2t}	$m^{-2(a-v)}$	$m^{2v+2t+1}$	$(n \wedge k)^{\frac{1}{2a+2t+1}}$	$(n \wedge k)^{-\frac{2(a-v)}{2a+2t+1}}$
$v + t = -1/2$	j^{-2a} j^{-2t}	$m^{-2a-2t-1}$	$\log m$	$\left(\frac{n \wedge k}{\log n \wedge k}\right)^{\frac{1}{2a+2t+1}}$	$\frac{\log n \wedge k}{n \wedge k}$
(o-s) $a - v \in \mathbb{R}_{>0}$	j^{-2a} $e^{-j^{2t}}$	$m^{-2(a-v)}$	$m^{(1-2(t-v))_+} e^{m^{2t}}$	$(\log n \wedge k)^{\frac{1}{2t}}$	$(\log n \wedge k)^{-\frac{a-v}{t}}$
(s-m) $v + t + 1/2 \in \mathbb{R}_{>0}$	$e^{-j^{2a}}$ j^{-2t}	$m^{2v} e^{-m^{2a}}$	$m^{2v+2t+1}$	$(\log n \wedge k)^{\frac{1}{2a}}$	$\frac{(\log n \wedge k)^{\frac{2t+2v+1}{2a}}}{n \wedge k}$
$v + t = -1/2$	$e^{-j^{2a}}$ j^{-2t}	$m^{2v} e^{-m^{2a}}$	$\log m$	$(\log n \wedge k)^{\frac{1}{2a}}$	$\frac{\log \log n \wedge k}{n \wedge k}$

We note that in case **(o-m)** and **(s-m)** for $v + t < -1/2$ the rate $R_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ is parametric. The tGE attains the rate $R_{n \wedge k}^* := R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ due to **Corollary** §08|02.44 under the additional condition

$$(k^{-1}(m_{n \wedge k}^*)^3 \mathfrak{t}_{m_{n \wedge k}^*}^{-2})^{l-1} \leq (m_{n \wedge k}^*)^{-2} R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}). \quad (08.32)$$

Since $(m_{n \wedge k}^*)^{-2} R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) = o(1)$ also $k^{-1}(m_{n \wedge k}^*)^3 \mathfrak{t}_{m_{n \wedge k}^*}^{-2} = o(1)$ is necessary as $n \wedge k \rightarrow \infty$. The next table depicts the order of both terms in case **(o-m)**, **(o-s)** and **(s-m)**.

Table 08 [§08]

Order as $n \wedge k \rightarrow \infty$

	(o-m) $v \in (-1/2 - t, a)$	(o-s) $a - v \in \mathbb{R}_{>0}$	(s-m) $v + t + 1/2 \in \mathbb{R}_{>0}$
$(m_{n \wedge k}^*)^{-2} R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$	$(n \wedge k)^{-\frac{2(a-v)+2}{2a+2t+1}}$	$(\log n \wedge k)^{-\frac{2a-2v+2}{2t}}$	$\frac{(\log n \wedge k)^{\frac{2t+2v-1}{2a}}}{n \wedge k}$
$(n \wedge k)^{-1} (m_{n \wedge k}^*)^3 \mathfrak{t}_{m_{n \wedge k}^*}^{-2}$	$(n \wedge k)^{-\frac{2a-2}{2a+2t+1}}$	$(n \wedge k)^{-c}$	$\frac{(\log n \wedge k)^{\frac{2t+3}{2a}}}{n \wedge k}$

In case **(o-s)** a value $l \geq 2$ and **(s-m)** a value $l \geq 3$ is sufficient to ensure (08.32) as $n \wedge k \rightarrow \infty$. In case **(o-m)** assuming $a > 1$ we have $k^{-1}(m_{n \wedge k}^*)^3 \mathfrak{t}_{m_{n \wedge k}^*}^{-2} = o(1)$ as $n \wedge k \rightarrow \infty$. In this situation we have (08.32) if $2(a-1)(l-1) > 2(a-v) + 2$ or in equal $l > (2a-v)/(a-1)$. \square

§08|02|03 Local and maximal local ϕ -risk

We measure the accuracy of the (generalised) tGE $\hat{\theta}^m := \widehat{\mathbb{T}}_{\cdot, \cdot}^{m|(n \wedge k)|\dagger} \hat{g}$ of the (generalised) Galerkin solution $\theta^m = \mathbb{T}_{\cdot, \cdot}^{m|\dagger} g \in \ell_2 \mathbf{1}_m^m$ with $g = \mathbb{T}_{\cdot, \cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^\dagger)$ by the mean of its local ϕ -error introduced in §05|01|02 and §05|02|02, i.e. its ϕ -risk.

§08|02.54 **Reminder.** If $\phi \in \mathbb{R}_{\setminus 0}^{\mathbb{N}}$ then we have $\phi^2 \mathbf{1}_m^m \in \ell_2$ and $\ell_2 \mathbf{1}_m^m \subseteq \text{dom}(\phi_{\mathbb{N}})$. Consequently, for each $\theta \in \text{dom}(\phi_{\mathbb{N}})$ the (generalised) Galerkin solution $\theta^m = \mathbb{T}_{\cdot, \cdot}^{m|\dagger} g \in \ell_2 \mathbf{1}_m^m$ satisfies $\theta^m \in \text{dom}(\phi_{\mathbb{N}})$ too. If in addition $C_T := \sup \{ \|\mathbb{M}_{\mathbb{T}_m^\perp} \mathbb{T}_{\cdot, \cdot}^*(\mathbb{T}_{\cdot, \cdot}^{m|\dagger})^* \phi\|_{\ell_2} : m \in \mathbb{N} \} \in \mathbb{R}_{\geq 0}$ then $|\phi_{\mathbb{N}}(\theta^m - \theta)| \leq (1 + C_T) \|\mathbf{1}_m^m \perp \theta\|_{\ell_2}$ which implies $\sup \{ |\phi_{\mathbb{N}}(\theta^j - \theta)| : j \in \mathbb{N}_{\geq m} \} = o(1)$ as $m \rightarrow \infty$ (**Property** §05|01.31 and **Property** §05|02.12). \square

§08|02.55 **Comment.** Under Assumption §08|02.02 we have $\dot{\mathbf{e}} \mathbf{1}_m^m \in \ell_\infty \mathbb{P}_{\theta|\mathbb{T}}^n$ -a.s. and $\widehat{\mathbb{T}}_{\cdot, \cdot}^m \in \mathbb{L}(\ell_2)$ with $\text{ran}(\widehat{\mathbb{T}}_{\cdot, \cdot}^m) \subseteq \ell_2 \mathbf{1}_m^m \mathbb{P}_T^k$ -a.s. for each $m \in \mathbb{N}$. Consequently, $\text{ran}(\widehat{\mathbb{T}}_{\cdot, \cdot}^{m|(k \wedge n)|\dagger}) \subseteq \ell_2 \mathbf{1}_m^m \mathbb{P}_T^k$ -a.s., and $\widehat{\mathbb{T}}_{\cdot, \cdot}^{m|(k \wedge n)|\dagger} \dot{\mathbf{e}} \in \ell_2 \mathbf{1}_m^m \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k}$ -a.s., and hence

$$\hat{\theta}^m = \widehat{\mathbb{T}}_{\cdot, \cdot}^{m|(k \wedge n)|\dagger} \hat{g} = n^{-1/2} \widehat{\mathbb{T}}_{\cdot, \cdot}^{m|(k \wedge n)|\dagger} \dot{\mathbf{e}} + \widehat{\mathbb{T}}_{\cdot, \cdot}^{m|(k \wedge n)|\dagger} g \in \ell_2 \mathbf{1}_m^m \subseteq \text{dom}(\phi_{\mathbb{N}}) \quad \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k}\text{-a.s.} \quad \square$$

§08|02|03|01 Local ϕ -risk

§08|02.56 **Assumption.** Let $\phi \in \mathbb{R}_{\setminus 0}^{\mathbb{N}}$ and $\theta \in \text{dom}(\phi_{\mathbb{N}})$ be satisfied. \square

§08|02.57 **Definition.** Under Assumptions §08|02.02 and §08|02.56 the *local ϕ -risk* of a (generalised) tGE $\widehat{\theta}^m = \widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(n \wedge k)|\dagger} \widehat{g} \in \ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi_{\mathbb{N}}) \mathbb{P}_{\theta|\mathbb{T}}^n$ -a.s. satisfies

$$\mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}^m - \theta)|^2) = \mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(k \wedge n)|\dagger}(\widehat{g} - g))|^2) + \mathbb{E}_{\mathbb{T}}^k (|\phi_{\mathbb{N}}(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(k \wedge n)|\dagger} g - \theta)|^2) \quad (08.33)$$

with $\mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(k \wedge n)|\dagger}(\widehat{g} - g))|^2) = n^{-1} \mathbb{E}_{\mathbb{T}}^k \|(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(k \wedge n)|\dagger})^* \phi^m\|_{\mathbb{E}_{\theta|\mathbb{T}}}^2$ (see **Property** §08|01.45). \square

§08|02.58 **Property.** Under Assumption §08|02.02 we have

$$\begin{aligned} \mathbb{E}_{\mathbb{T}}^k (|\phi_{\mathbb{N}}(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(k \wedge n)|\dagger} g - \theta)|^2) &= \mathbb{E}_{\mathbb{T}}^k (|\phi_{\mathbb{N}}(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|\dagger}(\mathbb{T}_{\bullet, \bullet}^m - \widehat{\mathbb{T}}_{\bullet, \bullet}^m)\theta^m + (\theta^m - \theta))|^2 \mathbb{1}_{\Omega_{m,k \wedge n}}) + |\phi_{\mathbb{N}}(\theta)|^2 \mathbb{E}_{\mathbb{T}}^k (\Omega_{m,k \wedge n}^c) \\ &\leq 2 \mathbb{E}_{\mathbb{T}}^k (|\phi_{\mathbb{N}}(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|\dagger}(\mathbb{T}_{\bullet, \bullet}^m - \widehat{\mathbb{T}}_{\bullet, \bullet}^m)\theta^m)|^2 \mathbb{1}_{\Omega_{m,k \wedge n}}) + 2 |\phi_{\mathbb{N}}(\theta^m - \theta)|^2 + |\phi_{\mathbb{N}}(\theta)|^2 \mathbb{E}_{\mathbb{T}}^k (\Omega_{m,k \wedge n}^c) \end{aligned}$$

(since $\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|\dagger} \widehat{\mathbb{T}}_{\bullet, \bullet}^m \mathbb{1}_{\Omega_{m,k \wedge n}} = \widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(k \wedge n)|\dagger} \widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(k \wedge n)} = M_{\mathbb{1}^m} \mathbb{1}_{\Omega_{m,k \wedge n}}$). \square

§08|02.59 **Notation (Reminder).** Let $A \in \mathbb{L}(\ell_2)$ be a *Hilbert-Schmidt operator*, $A \in \mathbb{HS}(\ell_2)$ for short, where $\|A\|_{\mathbb{HS}}^2 := \text{tr}(A^*A) = \text{tr}(AA^*) \in \mathbb{R}_{\geq 0}$. If $\Gamma \in \mathbb{L}(\ell_2)$ then $\text{tr}(A^*\Gamma A) \leq \|\Gamma\|_{\mathbb{L}(\ell_2)} \text{tr}(A^*A) = \|\Gamma\|_{\mathbb{L}(\ell_2)} \|A\|_{\mathbb{HS}}^2$. For arbitrary $A \in \mathbb{L}(\ell_2)$ we have $M_v A^m = M_v^m A^m \in \mathbb{HS}(\ell_2)$. \square

§08|02.60 **Notation.** For each $m \in \mathbb{N}$ and $\mathbb{T}_{\bullet, \bullet} \in \mathbb{L}^{\mathbb{R}}(\ell_2)$ we consider the observable event and its complement

$$\Omega_{m,k \wedge n} := \{ \|(\widehat{\mathbb{T}}_{\bullet, \bullet}^m)^{-1} [\phi]_{\mathbb{1}^m} \|^2 \leq k \wedge n \} \quad \text{and} \quad \Omega_{m,k \wedge n}^c := \{ \|(\widehat{\mathbb{T}}_{\bullet, \bullet}^m)^{-1} [\phi]_{\mathbb{1}^m} \|^2 > k \wedge n \}. \quad (08.34)$$

On the event $\Omega_{m,k \wedge n}$ the random matrix $[\widehat{\mathbb{T}}_{\bullet, \bullet}^m] \in \mathbb{R}^{(m,m)}$ is regular with inverse $[\widehat{\mathbb{T}}_{\bullet, \bullet}^m]^{-1} \in \mathbb{R}^{(m,m)}$. Moreover, setting $A_{\bullet, \bullet}^m := \mathfrak{r}_{\bullet, \bullet}^m \mathbb{T}_{\bullet, \bullet}^{m|\dagger}$ we introduce an unobserved event and its complement

$$\mathcal{U}_{m,k} := \{ 4m \|A_{\bullet, \bullet}^m\|_{\mathbb{L}(\ell_2)}^2 \leq k \} \quad \text{and} \quad \mathcal{U}_{m,k}^c := \{ 4m \|A_{\bullet, \bullet}^m\|_{\mathbb{L}(\ell_2)}^2 > k \}. \quad (08.35)$$

Note that $\mathbb{1}_{\mathcal{U}_{m,k}} = \mathbb{1}_{\{4m \|A_{\bullet, \bullet}^m\|_{\mathbb{L}(\ell_2)}^2 \leq k\}}$ denotes an unobserved elementary random variable. \square

§08|02.61 **Lemma.** Under Assumptions §08|02.02 and §08|02.56 for all $m, k, n \in \mathbb{N}$ we have

- (i) if $4 \|(\mathbb{T}_{\bullet, \bullet}^{m|\dagger})^* \phi^m\|_{\ell_2}^2 \leq k \wedge n$ then $\mathcal{U}_{m,k} \subseteq \Omega_{m,k \wedge n}$,
 - (ii) $\mathbb{E}_{\mathbb{T}}^k (\|(\widehat{\mathbb{T}}_{\bullet, \bullet}^m)^{-1} [\phi]_{\mathbb{1}^m} \|^2_{\mathbb{E}_{\theta|\mathbb{T}}} \mathbb{1}_{\Omega_{m,k \wedge n}}) \leq \mathfrak{v}_{\theta|\mathbb{T}} (4 \|(\mathbb{T}_{\bullet, \bullet}^{m|\dagger})^* \phi^m\|_{\ell_2}^2 + (k \wedge n) \mathbb{E}_{\mathbb{T}}^k (\mathcal{U}_{m,k}^c))$, and
 - (iii) $\mathbb{E}_{\mathbb{T}}^k (|\phi_{\mathbb{N}}(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|\dagger}(\mathbb{T}_{\bullet, \bullet}^m - \widehat{\mathbb{T}}_{\bullet, \bullet}^m)\theta^m)|^2 \mathbb{1}_{\Omega_{m,k \wedge n}}) \leq 4k^{-1} \mathfrak{v}_{\mathbb{T}} \|(\mathbb{T}_{\bullet, \bullet}^{m|\dagger})^* \phi^m\|_{\ell_2}^2 \|\theta^m\|_{\ell_2}^2 + \mathbb{E}_{\mathbb{T}}^k (\|\mathfrak{r}_{\bullet, \bullet}^m \theta^m\|_{\ell_2}^2 \mathbb{1}_{\mathcal{U}_{m,k}^c})$.
- with $\Omega_{m,k \wedge n}$ and $\mathcal{U}_{m,k}$ as in (08.34) and (08.35), respectively.

§08|02.62 **Proof of Lemma** §08|02.61. Given in the lecture. \square

§08|02.63 **Proposition (Upper bound).** Under Assumptions §08|02.02 and §08|02.56 for all $n, m \in \mathbb{N}$ with (generalised) Galerkin solution $\theta^m = \mathbb{T}_{\bullet, \bullet}^{m|\dagger} g \in \ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi_{\mathbb{N}})$ setting similar to (08.11)

$$\begin{aligned} R_{n \wedge k}^m(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) &:= |\phi_{\mathbb{N}}(\theta^m - \theta)|^2 + (n \wedge k)^{-1} \|(\mathbb{T}_{\bullet, \bullet}^{m|\dagger})^* \phi^m\|_{\ell_2}^2, \\ m_{n \wedge k}^{\circ} &:= \arg \min \{ R_{n \wedge k}^m(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) : m \in \mathbb{N} \} \quad \text{and} \\ R_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) &:= R_{n \wedge k}^{m_{n \wedge k}^{\circ}}(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) = \min \{ R_{n \wedge k}^m(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) : m \in \mathbb{N} \} \quad (08.36) \end{aligned}$$

for all $m \in \mathbb{N}$ the (generalised) tGE $\widehat{\theta}^m = \widehat{\mathbb{T}}_{\cdot, \cdot}^{m|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbf{1}^m \subseteq \text{dom}(\phi_{\mathbb{N}}) \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k}$ -a.s. satisfies

$$\begin{aligned} \mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}^m - \theta)|^2) &\leq (4\mathfrak{v}_{\theta|\mathbb{T}} + 8\mathfrak{v}_{\mathbb{T}} \|\theta^m\|_{\ell_2}^2) \mathbb{R}_{n \wedge k}^m(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) \\ &\quad + \mathfrak{v}_{\theta|\mathbb{T}} \mathbb{E}_{\mathbb{T}}^k(\mathbb{U}_{m,k}^c) + 2\mathbb{E}_{\mathbb{T}}^k(\|\mathfrak{n}_{\cdot, \cdot}^m \theta^m\|_{\ell_2}^2 \mathbf{1}_{\mathbb{U}_{m,k}^c}) + |\phi_{\mathbb{N}}(\theta)|^2 \mathbb{E}_{\mathbb{T}}^k(\Omega_{m,k \wedge n}^c) \end{aligned} \quad (08.37)$$

with $\Omega_{m,k \wedge n}$ and $\mathbb{U}_{m,k}$ as in (08.34) and (08.35), respectively.

§08102.64 **Proof of Proposition** §08102.63. Given in the lecture. \square

§08102.65 **Corollary.** Under the assumptions of **Proposition** §08102.63 the (infeasible, generalised) tGE $\widehat{\theta}_{n \wedge k}^{m \circ} = \widehat{\mathbb{T}}_{\cdot, \cdot}^{m \circ |k \wedge n| \dagger} \widehat{g} \in \ell_2 \mathbf{1}^{m \circ} \subseteq \text{dom}(\phi_{\mathbb{N}}) \mathbb{P}_{\theta|\mathbb{T}}^{n \otimes k}$ -a.s. with $\Omega_{m \circ, k \wedge n}^{\circ}$ as in (08.34) and (infeasible) dimension $m_{n \wedge k}^{\circ}$ as in (08.36) for each $k, n \in \mathbb{N}$ with $\mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) \leq 1/4$ satisfies

$$\begin{aligned} \mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}_{n \wedge k}^{m \circ} - \theta)|^2) &\leq (4\mathfrak{v}_{\theta|\mathbb{T}} + 8\mathfrak{v}_{\mathbb{T}} \|\theta^{m \circ}\|_{\ell_2}^2) \mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) \\ &\quad + 2^{2l} \mathbb{K}_{\mathbb{T}}^{2l} \{ (\mathfrak{v}_{\theta|\mathbb{T}} + |\phi_{\mathbb{N}}(\theta)|^2) k^{-1} m_{n \wedge k}^{\circ} \|\mathbb{T}_{\cdot, \cdot}^{m \circ} \|_{\mathbb{L}(\ell_2)}^2 + \|\theta^{m \circ}\|_{\mathbb{L}(\ell_2)}^2 / 4 \} \\ &\quad \times (m_{n \wedge k}^{\circ})^2 (k^{-1} (m_{n \wedge k}^{\circ})^3 \|\mathbb{T}_{\cdot, \cdot}^{m \circ} \|_{\mathbb{L}(\ell_2)}^2)^{l-1} \end{aligned} \quad (08.38)$$

and if in addition

$$(m_{n \wedge k}^{\circ})^2 (k^{-1} (m_{n \wedge k}^{\circ})^3 \|\mathbb{T}_{\cdot, \cdot}^{m \circ} \|_{\mathbb{L}(\ell_2)}^2)^{l-1} \leq \mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) \leq 1/4 \quad (08.39)$$

then we have

$$\begin{aligned} \mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}_{n \wedge k}^{m \circ} - \theta)|^2) &\leq \mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) \\ &\quad \times \{ (4 + 2^{2l} \mathbb{K}_{\mathbb{T}}^{2l} (m_{n \wedge k}^{\circ})^{-2}) \mathfrak{v}_{\theta|\mathbb{T}} + (8\mathfrak{v}_{\mathbb{T}} + 2^{2l-2} \mathbb{K}_{\mathbb{T}}^{2l}) \|\theta^{m \circ}\|_{\ell_2}^2 + 2^{2l} \mathbb{K}_{\mathbb{T}}^{2l} (m_{n \wedge k}^{\circ})^{-2} |\phi_{\mathbb{N}}(\theta)|^2 \} \\ &\leq 2^{2l+2} \mathbb{K}_{\mathbb{T}}^{2l} (\mathfrak{v}_{\theta|\mathbb{T}} + \|\theta^{m \circ}\|_{\ell_2}^2 + |\phi_{\mathbb{N}}(\theta)|^2) \mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) \end{aligned} \quad (08.40)$$

§08102.66 **Proof of Corollary** §08102.65. Given in the lecture. \square

§08102.67 **Remark.** Consider $m_{n \wedge k}^{\circ} = \arg \min \{ \mathbb{R}_{n \wedge k}^m(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) : m \in \mathbb{N} \}$ and $\mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) = \mathbb{R}_{n \wedge k}^{m_{n \wedge k}^{\circ}}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi)$ as in (08.36). Arguing similarly as in **Remark** §07101.21 we note that $\|(\mathbb{T}_{\cdot, \cdot}^{m \circ})^* \phi^m\|_{\ell_2}^2 \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and hence $\mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi) = o(1)$ as $n \wedge k \rightarrow \infty$, whenever $|\phi_{\mathbb{N}}(\theta^m - \theta)| = o(1)$ as $m \rightarrow \infty$ (c.f. **Remark** §05101.05). In this situation if $\sup \{ \|\theta^m\|_{\ell_2}^2 : m \in \mathbb{N} \} \leq \mathbb{K}_{\theta|\mathbb{T}}^2 \in \mathbb{R}_{\geq 0}$ then from (08.40) in **Corollary** §08102.65 follows

$$\mathbb{E}_{\theta|\mathbb{T}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}_{n \wedge k}^{m \circ} - \theta)|^2) \leq 2^{2l+2} \mathbb{K}_{\mathbb{T}}^{2l} (\mathfrak{v}_{\theta|\mathbb{T}} + \mathbb{K}_{\theta|\mathbb{T}}^2 + |\phi_{\mathbb{N}}(\theta)|^2) \mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi).$$

However, the dimension $m_{n \wedge k}^{\circ} = m_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi)$ as defined in (08.11) depends on the unknown parameter of interest θ and the nuisance parameter $\mathbb{T}_{\cdot, \cdot}$, and thus also the statistic $\widehat{\theta}_{n \wedge k}^{m \circ}$. In other words $\widehat{\theta}_{n \wedge k}^{m \circ}$ is not a feasible estimator. \square

§08102.68 **Corollary** (GniSM with noisy operator §08102.08 continued). Consider independent noisy versions $(\widehat{g}, \widehat{\mathbb{T}}_{\cdot, \cdot}) = (g + n^{-1/2} \dot{\mathbb{B}}_{\cdot, \cdot}, \mathbb{T}_{\cdot, \cdot} + k^{-1/2} \dot{\mathbb{W}}_{\cdot, \cdot}) \sim \mathbb{N}_{\theta|\mathbb{T}}^{n \otimes k} = \mathbb{N}_{\theta|\mathbb{T}}^n \otimes \mathbb{N}_{\mathbb{T}}^k$ as in Model §08102.08, where $\dot{\mathbb{B}}_{\cdot, \cdot} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{\mathbb{W}}_{\cdot, \cdot} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}^2}$ are independent, $\mathbb{T}_{\cdot, \cdot} \in \mathbb{T}$ and $\theta \in \ell_2$, and hence $g = \mathbb{T}_{\cdot, \cdot} \theta \in \text{dom}(\mathbb{T}_{\cdot, \cdot}^{\dagger}) \subseteq \ell_2$. Given Assumption §08102.18 for each $k, n \in \mathbb{N}$ fulfilling (08.23) the (infeasible, generalised) tGE $\widehat{\theta}_{n \wedge k}^{m \circ} = \widehat{\mathbb{T}}_{\cdot, \cdot}^{m \circ |k \wedge n| \dagger} \widehat{g} \in \ell_2 \mathbf{1}^{m \circ} \subseteq \text{dom}(\phi_{\mathbb{N}})$ satisfies

$$\mathbb{N}_{\theta|\mathbb{T}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}_{n \wedge k}^{m \circ} - \theta)|^2) \leq 2^{2l+2} ((2l-1)!!) (1 + \|\theta^{m \circ}\|_{\ell_2}^2 + |\phi_{\mathbb{N}}(\theta)|^2) \mathbb{R}_{n \wedge k}^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi)$$

where $\mathbb{R}_n^{\circ}(\theta, \mathbb{T}_{\cdot, \cdot}, \phi)$ is the oracle rate in a GniSM §08101.04 (see **Corollary** §08101.51).

§08102.69 **Proof of Corollary §08102.68.** Given in the lecture. \square

§08102.70 **Corollary** (niSM with noisy operator §08102.10 continued). *Consider independent noisy versions $(\widehat{g}_\cdot, \widehat{\mathfrak{s}}_\cdot) = (g_\cdot + n^{-1/2}\widehat{\epsilon}_\cdot, T_{\cdot, \cdot} + k^{-1/2}\widehat{\eta}_{\cdot, \cdot}) \sim \mathbb{P}_{\theta|T|\sigma|\xi|\zeta^{(2)}}^{n \otimes k}$ as in Model §08102.10, where $\widehat{\epsilon}_\cdot$ and $\widehat{\eta}_{\cdot, \cdot}$ satisfies (iSM1) with $\mathbb{V}_\sigma = \|\sigma_\cdot^2\|_{\ell_\infty} \vee 1$ and (niSMnO1)–(niSMnO2) with $K_{\xi^{(2)}}^{2l} := 1 \vee \|\xi_{\cdot, \cdot}^{(2l)}\|_{\ell_\infty(\mathbb{N}^2)}$, respectively, $T_{\cdot, \cdot} \in \mathbb{T}$ and $\theta \in \ell_2$, and hence $g_\cdot = T_{\cdot, \cdot}\theta \in \text{dom}(T_{\cdot, \cdot}^\dagger) \subseteq \ell_2$. Given Assumption §08102.56 for each $k, n \in \mathbb{N}$ fulfilling (08.39) the (infeasible, generalised) tGE $\widehat{\theta}_{n \wedge k}^{m_{n \wedge k}^\circ} = \widehat{T}_{\cdot, \cdot}^{m_{n \wedge k}^\circ | (k \wedge n) | \dagger} \widehat{g}_\cdot \in \ell_2 \mathbf{1}_{m_{n \wedge k}^\circ} \subseteq \text{dom}(\phi_{\mathbb{N}})$ satisfies*

$$\mathbb{E}_{\theta|T|\sigma|\xi|\zeta^{(2)}}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}_{n \wedge k}^{m_{n \wedge k}^\circ} - \theta)|^2) \leq 2^{2l+2} K_{\xi^{(2)}}^{2l} (\mathbb{V}_\sigma + \|\theta_{n \wedge k}^{m_{n \wedge k}^\circ}\|_{\ell_2}^2 + |\phi_{\mathbb{N}}(\theta)|^2) R_{n \wedge k}^\circ(\theta, T_{\cdot, \cdot}, \phi)$$

where $R_n^\circ(\theta, T_{\cdot, \cdot}, \phi)$ is the oracle rate in a niSM §08101.06 (see Corollary §08101.53).

§08102.71 **Proof of Corollary §08102.70.** Given in the lecture. \square

§08102.72 **Corollary** (nieMM with noisy operator §08102.13 continued). *Consider independent noisy versions $(\widehat{g}_\cdot, \widehat{T}_{\cdot, \cdot}) = (g_\cdot + n^{-1/2}\widehat{\epsilon}_\cdot, T_{\cdot, \cdot} + k^{-1/2}\widehat{\eta}_{\cdot, \cdot})$ defined on $(\mathcal{Z}^{n+k}, \mathcal{X}^{\otimes(n+k)}, \mathbb{P}_{\theta|T}^{n \otimes k})$ as in Model §08102.13, where $\psi \in \mathcal{M}(\mathcal{X} \otimes 2^{\mathbb{N}})$ and $\varphi_{\cdot, \cdot} \in \mathcal{M}(\mathcal{X} \otimes 2^{\mathbb{N}^2})$ satisfies (nieMM1)–(nieMM2) for $\mathbb{V}_{\theta|T|\psi} \in \mathbb{R}_{\geq 1}$ and (nieMMnO1)–(nieMMnO3) for $K_{T|\varphi} \in \mathbb{R}_{\geq 1}$, respectively, $T_{\cdot, \cdot} \in \mathbb{T}$ and $\theta \in \ell_2$, and hence $g_\cdot = T_{\cdot, \cdot}\theta \in \text{dom}(T_{\cdot, \cdot}^\dagger) \subseteq \ell_2$. Given Assumption §08102.56 for each $k, n \in \mathbb{N}$ fulfilling (08.39) the (infeasible, generalised) tGE $\widehat{\theta}_{n \wedge k}^{m_{n \wedge k}^\circ} = \widehat{T}_{\cdot, \cdot}^{m_{n \wedge k}^\circ | (k \wedge n) | \dagger} \widehat{g}_\cdot \in \ell_2 \mathbf{1}_{m_{n \wedge k}^\circ} \subseteq \text{dom}(\phi_{\mathbb{N}})$ satisfies*

$$\mathbb{E}_{\theta|T}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}_{n \wedge k}^{m_{n \wedge k}^\circ} - \theta)|^2) \leq C_{2l} K_{T|\varphi}^{2l} (\mathbb{V}_{\theta|T|\psi} + \|\theta_{n \wedge k}^{m_{n \wedge k}^\circ}\|_{\ell_2}^2 + |\phi_{\mathbb{N}}(\theta)|^2) R_{n \wedge k}^\circ(\theta, T_{\cdot, \cdot}, \phi)$$

where $C_{2l} \in \mathbb{R}_{\geq 1}$ is a constant depending on $l \in \mathbb{N}$ only and $R_n^\circ(\theta, T_{\cdot, \cdot}, \phi)$ is the oracle rate in a nieMM §08101.08 (see Corollary §08101.55).

§08102.73 **Proof of Corollary §08102.72.** Given in the lecture. \square

§08102.74 **Illustration.** We distinguish as in Illustration §08101.57 the two cases **(p)** and **(np)**, where **(p)** is implied by $\theta \mathbf{1}_{K^{1+}} = 0$. In case **(p)** the oracle bound is parametric, that is, $nR_n^\circ(\theta, T_{\cdot, \cdot}, \phi) = O(1)$, in case **(np)** the oracle bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^\circ(\theta, T_{\cdot, \cdot}, \phi) = \infty$. In case **(np)** consider similar to **(o-m)**, **(o-s)** and **(s-m)** in Illustration §08101.57 the following specifications:

Table 09 [§08]

Order of the rate $R_{n \wedge k}^\circ(\theta, T_{\cdot, \cdot}, \phi)$ as $n \wedge k \rightarrow \infty$

	(squared bias)	(variance)	$m_{n \wedge k}^\circ$	$R_{n \wedge k}^\circ(\theta, T_{\cdot, \cdot}, \phi)$	
$(m \in \mathbb{N})$ $(\phi_m = m^{v-1/2})$	$ \phi_{\mathbb{N}}(\widehat{\theta}_m - \theta) ^2$ ($a \in \mathbb{R}_{>0}$)	$\ (T_{\cdot, \cdot}^{m \dagger})^* \phi_m\ _{\ell_2}^2$ ($t \in \mathbb{R}_{>0}$)			
(o-m)	$v \in (-t, a)$ $v = -t$	$m^{-2(a-v)}$ $m^{-2(a+t)}$	m^{2t+2v}	$(n \wedge k)^{-\frac{1}{2a+2t}}$ $(\frac{n \wedge k}{\log n \wedge k})^{\frac{1}{2(a+t)}}$	$(n \wedge k)^{-\frac{a-v}{a+t}}$ $\frac{\log n \wedge k}{n \wedge k}$
(o-s)	$a - v \in \mathbb{R}_{>0}$	$m^{-2(a-v)}$	$m^{2(v-t)+e^{m^a}}$	$(\log n \wedge k)^{\frac{1}{2t}}$	$(\log n \wedge k)^{-\frac{a-v}{t}}$
(s-m)	$v + t \in \mathbb{R}_{>0}$ $v = -t$	$m^{(1-4a+2v)+e^{-m^{2a}}}$ $m^{(1-4a-2t)+e^{-m^{2a}}}$	m^{2t+v}	$(\log n \wedge k)^{\frac{1}{2a}}$ $(\log n \wedge k)^{\frac{1}{2a}}$	$\frac{(\log n \wedge k)^{\frac{t+v}{a}}}{n \wedge k}$ $\frac{\log \log n \wedge k}{n \wedge k}$

We note that in case **(o-m)** and **(s-m)** for $v < -t$ the rate $R_{n \wedge k}^\circ(\theta, T_{\cdot, \cdot}, \phi)$ is parametric. The tGE attains the rate $R_{n \wedge k}^\circ := R_{n \wedge k}^\circ(\theta, T_{\cdot, \cdot}, \phi)$ due to Corollary §08102.65 under the additional condition

$$(k^{-1}(m_{n \wedge k}^\circ)^3 \|\mathbf{T}_{\cdot, \cdot}^{m_{n \wedge k}^\circ \dagger}\|_{\mathbb{L}(\ell_2)}^2)^{l-1} \leq (m_{n \wedge k}^\circ)^{-2} R_{n \wedge k}^\circ(\theta, T_{\cdot, \cdot}, \phi). \quad (08.41)$$

Since $(m_{n \wedge k}^\circ)^{-2} \mathbb{R}_{n \wedge k}^\circ(\theta, \mathbb{T}_{\bullet, \bullet}, \phi) = o(1)$ also $k^{-1}(m_{n \wedge k}^\circ)^3 \|\mathbb{T}_{\bullet, \bullet}^{m_{n \wedge k}^\circ \dagger}\|_{\mathbb{L}(\ell_2)}^2 = o(1)$ is necessary as $n \wedge k \rightarrow \infty$. The next table depicts the order of both terms in case **(o-m)**, **(o-s)** and **(s-m)**.

Table 10 [§08]

Order as $n \wedge k \rightarrow \infty$

	(o-m) $v \in (-t, a)$	(o-s) $a - v \in \mathbb{R}_{>0}$	(s-m) $v + t \in \mathbb{R}_{>0}$
$(m_{n \wedge k}^\circ)^{-2} \mathbb{R}_{n \wedge k}^\circ(\theta, \mathbb{T}_{\bullet, \bullet}, \phi)$	$(n \wedge k)^{-\frac{2(a-v)+2}{2a+2t}}$	$(\log n \wedge k)^{-\frac{2a-2v+2}{2t}}$	$\frac{(\log n \wedge k)^{\frac{2t+2v-2}{2a}}}{n \wedge k}$
$(n \wedge k)^{-1}(m_{n \wedge k}^\circ)^3 \ \mathbb{T}_{\bullet, \bullet}^{m_{n \wedge k}^\circ \dagger}\ _{\mathbb{L}(\ell_2)}^2$	$(n \wedge k)^{-\frac{2a-3}{2a+2t}}$	$(n \wedge k)^{-c}$	$\frac{(\log n \wedge k)^{\frac{2t+3}{2a}}}{n \wedge k}$

In case **(o-s)** a value $l \geq 2$ and **(s-m)** a value $l \geq 3$ is sufficient to ensure (08.41) as $n \wedge k \rightarrow \infty$. In case **(o-m)** assuming $a > 3/2$ we have $k^{-1}(m_{n \wedge k}^\circ)^3 \|\mathbb{T}_{\bullet, \bullet}^{m_{n \wedge k}^\circ \dagger}\|_{\mathbb{L}(\ell_2)}^2 = o(1)$ as $n \wedge k \rightarrow \infty$. In this situation we have (08.41) if $(2a-3)(l-1) > 2(a-v)+2$ or in equal $l > (4a-2v-1)/(2a-3)$. \square

§08|02|03|02 Maximal local ϕ -risk

§08|02.75 **Assumption.** Consider weights $\mathfrak{t}, \mathfrak{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and $\phi \in \mathbb{R}_{\searrow 0}^{\mathbb{N}}$ such that $(\mathfrak{a}\phi)_{\bullet} := \mathfrak{a}\phi \in \ell_2$ and $(\mathfrak{a}\mathfrak{t})_{\bullet} := \mathfrak{a}\mathfrak{t} \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$. \square

§08|02.76 **Comment.** Assuming $\mathfrak{t}, \mathfrak{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and hence $(\mathfrak{a}\mathfrak{t})_{\bullet}^2 \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ is rather weak. If in addition $\liminf_{j \rightarrow \infty} (\mathfrak{a}\mathfrak{t})_j^2 \geq c \in \mathbb{R}_{>0}$ is satisfied, and hence $(\mathfrak{a}\mathfrak{t})_{\bullet}^2, \mathfrak{a}_{\bullet}^2, \mathfrak{t}_{\bullet}^2 \notin (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$, then $\mathfrak{a}_{\bullet}^2 \notin (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ and the assumption $(\mathfrak{a}\phi)_{\bullet} \in \ell_2$ implies $\phi \in \ell_2$, which together with $\mathfrak{t}_{\bullet}^2 \notin (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ implies $(\phi/\mathfrak{t})_{\bullet} \in \ell_2$, and thus the rate $\mathbb{R}_n^*(\mathfrak{a}, \mathfrak{t}, \phi)$ is parametric (Illustration §08|01.72). Since we are interested in the case of a non-parametric rate, the additional assumption $(\mathfrak{a}\mathfrak{t})_{\bullet}^2 \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ imposes a rather weak condition satisfied also in Illustration §08|01.72. \square

§08|02.77 **Reminder.** Under Assumption §08|02.75 we have $\ell_2^{\mathfrak{a}} = \text{dom}(M_{\mathfrak{a}^{-1}}) = \ell_2 \mathfrak{a} \subseteq \ell_2$ and the three measures $\nu_{\mathfrak{a}}, \mathfrak{a}^{-2} \nu_{\mathfrak{a}}$ and $|\phi| \nu_{\mathfrak{a}}$ dominate mutually each other, i.e. they share the same null sets (see Property §04|01.02). We consider $\ell_2^{\mathfrak{a}}$ endowed with $\|\cdot\|_{\mathfrak{a}^{-1}} = \|M_{\mathfrak{a}^{-1}} \cdot\|_{\ell_2}$ and given a constant $r \in \mathbb{R}_{>0}$ the ellipsoid $\ell_2^{\mathfrak{a}, r} := \{\mathfrak{a} \in \ell_2^{\mathfrak{a}} : \|\mathfrak{a}\|_{\mathfrak{a}^{-1}} \leq r\} \subseteq \ell_2^{\mathfrak{a}}$. Since $(\mathfrak{a}\phi)_{\bullet} \in \ell_2$ we have $\ell_2^{\mathfrak{a}} \subseteq \text{dom}(\phi \nu_{\mathfrak{a}})$ (Property §04|02.23). Consequently, if Assumption §08|02.75 and $\theta \in \ell_2^{\mathfrak{a}, r}$ are satisfied, then Assumption §08|02.56 is also fulfilled. Moreover, from $(\mathfrak{a}\phi)_{\bullet} \in \ell_2$ follows $\|\mathfrak{a} \mathbb{1}^{m \perp \perp}\|_{\phi} = \|(\mathfrak{a}\phi)_{\bullet} \mathbb{1}^{m \perp \perp}\|_{\ell_2} = o(1)$ as $m \rightarrow \infty$. For $s \in [0, 1]$ from $(\mathfrak{a}\mathfrak{t})_{\bullet}^s = \mathfrak{a} \mathfrak{t}^s \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ follows $(\mathfrak{a}\mathfrak{t}^s)_{\bullet} = ((\mathfrak{a}\mathfrak{t}^s)_{(m)}) := (\mathfrak{a}\mathfrak{t}^s)_{m+1} = \|(\mathfrak{a}\mathfrak{t}^s)_{\bullet} \mathbb{1}^{m \perp \perp}\|_{\ell_{\infty}}\|_{m \in \mathbb{N}} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$. Since $\phi, \mathfrak{t} \in \mathbb{R}_{\searrow 0}^{\mathbb{N}}$ under Assumption §08|02.75, we have $\ell_2 \mathbb{1}^m \subseteq \text{dom}(\phi \nu_{\mathfrak{a}})$ and $\|\mathfrak{t}^{-1} \mathbb{1}^m\|_{\phi} = \|(\phi/\mathfrak{t})_{\bullet} \mathbb{1}^m\|_{\ell_2} \in \mathbb{R}_{>0}$ for each $m \in \mathbb{N}$. Under the Assumptions §08|00.02 and §08|02.75 considering the generalised link condition $\mathbb{T}_{\bullet, \bullet} \in \mathbb{T}_{t, d, D}$ with band $D \in \mathbb{R}_{\geq 1}$ and $d \in [1, D]$ as in Definition §05|02.05 we have $\sup_{m \in \mathbb{N}} \{ \|([\mathbb{T}_{\bullet, \bullet}]_m^{-1})^* [M_{\mathfrak{t}}]_m \|_{\text{spec}} \} \leq D$, and hence

$$\begin{aligned} \|(\mathbb{T}_{\bullet, \bullet}^{m \dagger})^* \phi^m\|_{\ell_2} &= \|([\mathbb{T}_{\bullet, \bullet}]_m^{-1})^* [\phi]_m\| = \|([\mathbb{T}_{\bullet, \bullet}]_m^{-1})^* [M_{\mathfrak{t}}]_m [M_{\mathfrak{t}}]_m^{-1} [\phi]_m\| \\ &\leq \|([\mathbb{T}_{\bullet, \bullet}]_m^{-1})^* [M_{\mathfrak{t}}]_m \|_{\text{spec}} \| [M_{\mathfrak{t}}]_m^{-1} [\phi]_m \| \leq D \|\mathfrak{t}^{-1} \mathbb{1}^m\|_{\phi} \end{aligned} \quad (08.42)$$

using $\|[M_{\mathfrak{t}}]_m^{-1} [\phi]_m\| = \|\mathfrak{t}^{-1} \mathbb{1}^m\|_{\phi}$. Moreover, for each $m \in \mathbb{N}$ the generalised Galerkin solution $\theta^m := \mathbb{T}_{\bullet, \bullet}^{m \dagger} g \in \ell_2 \mathbb{1}^m$ of $\theta = \mathbb{T}_{\bullet, \bullet}^{\dagger} g \in \ell_2^{\mathfrak{a}, r}$ satisfies (Lemma §05|02.14)

$$|\phi \nu_{\mathfrak{a}}(\theta^m - \theta)|^2 \leq Dd(Dd + 1)r^2 (\|\mathfrak{a} \mathbb{1}^{m \perp \perp}\|_{\phi}^2 + (\mathfrak{a}\mathfrak{t}^s)_{(m)} \|\mathfrak{t}^{-1} \mathbb{1}^m\|_{\phi}^2). \quad (08.43)$$

Under Assumptions §08|00.02 and §08|02.75 the link condition $\mathbb{T}_{\bullet, \bullet} \in \mathbb{T}_{t, d}^{\geq}$ with band $d \in \mathbb{R}_{\geq 1}$ as in Definition §05|01.08 implies $\sup_{m \in \mathbb{N}} \{ \|([\mathbb{T}_{\bullet, \bullet}]_m^{-1})^* [M_{\mathfrak{t}}]_m \|_{\text{spec}} \} \leq 3d^2$ (Lemma §05|01.22), and hence

for each $m \in \mathbb{N}$ we have (08.42) with $D = 3d^2$ and the Galerkin solution $\theta^m := \mathbb{T}_{\bullet, \bullet}^{m|\dagger} g \in \ell_2 \mathbb{1}^m$ of $\theta = \mathbb{T}_{\bullet, \bullet}^\dagger g \in \ell_2^{\alpha, r}$ satisfies (08.43) with $D = 3d^2$ (**Lemma** §05101.34). \square

§08102.78 **Corollary.** Under Assumptions §08102.02 and §08102.75 let $\theta_\bullet := \mathbb{T}_{\bullet, \bullet}^\dagger g_\bullet \in \ell_2^{\alpha, r}$ and $\mathbb{T}_{\bullet, \bullet} \in \mathbb{T}_{t, d, D}$ (or $\mathbb{T}_{\bullet, \bullet} \in \mathbb{T}_{t, d}^{\geq}$ with $D = 3d^2$), for all $m, k, n \in \mathbb{N}$ we have

- (i) if $4D^2 \|\mathbb{t}^{-1} \mathbb{1}^m\|_\phi^2 \leq k \wedge n$ then $\mathcal{U}_{m, k} \subseteq \Omega_{m, k \wedge n}$,
 - (ii) $\mathbb{E}_T^k (\|(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|\dagger})^* \phi_\bullet^m\|_{\mathbb{T}_{\theta|T}}^2 \mathbb{1}_{\Omega_{m, k \wedge n}}) \leq \mathfrak{v}_{\theta|T} (4D^2 \|\mathbb{t}^{-1} \mathbb{1}^m\|_\phi^2 + (k \wedge n) \mathbb{E}_T^k (\mathcal{U}_{m, k}^c))$, and
 - (iii) $\mathbb{E}_T^k (|\phi_{\mathbb{N}}(\widehat{\mathbb{T}}_{\bullet, \bullet}^{m|\dagger}(\mathbb{T}_{\bullet, \bullet}^m - \widehat{\mathbb{T}}_{\bullet, \bullet}^m)\theta^m)|^2 \mathbb{1}_{\Omega_{m, k \wedge n}}) \leq 4k^{-1} \mathfrak{v}_T D^2 \|\mathbb{t}^{-1} \mathbb{1}^m\|_\phi^2 \|\theta^m\|_{\ell_2}^2 + \mathbb{E}_T^k (\|\mathfrak{n}_{\bullet, \bullet}^m \theta^m\|_{\ell_2}^2 \mathbb{1}_{\mathcal{U}_{m, k}^c})$.
- with $\Omega_{m, k \wedge n}$ and $\mathcal{U}_{m, k}$ as in (08.34) and (08.35), respectively.

§08102.79 **Proof of Corollary** §08102.78. Given in the lecture. \square

§08102.80 **Proposition (Upper bound).** Under Assumptions §08102.02 and §08102.75 let $\theta := \mathbb{T}_{\bullet, \bullet}^\dagger g \in \ell_2^{\alpha, r}$ and $\mathbb{T}_{\bullet, \bullet} \in \mathbb{T}_{t, d, D}$ (or $\mathbb{T}_{\bullet, \bullet} \in \mathbb{T}_{t, d}^{\geq}$ with $D = 3d^2$) for all $n, m \in \mathbb{N}$ setting similar to (08.15)

$$\begin{aligned} R_{n \wedge k}^m(\mathbf{a}, \mathbf{t}, \phi) &:= \|\mathbf{a} \mathbb{1}^{m|\perp}\|_\phi^2 + (n \wedge k)^{-1} \|\mathbb{t}^{-1} \mathbb{1}^m\|_\phi^2, \\ m_{n \wedge k}^* &:= \arg \min \{R_{n \wedge k}^m(\mathbf{a}, \mathbf{t}, \phi) : m \in \mathbb{N}\} \quad \text{and} \\ R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) &:= R_{n \wedge k}^{m_{n \wedge k}^*}(\mathbf{a}, \mathbf{t}, \phi) = \min \{R_{n \wedge k}^m(\mathbf{a}, \mathbf{t}, \phi) : m \in \mathbb{N}\} \end{aligned} \quad (08.44)$$

for all $m \in \mathbb{N}$ the (generalised) tGE $\widehat{\theta}^m = \widehat{\mathbb{T}}_{\bullet, \bullet}^{m|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbb{1}^m$ $\mathbb{E}_{\theta|T}^{n \otimes k}$ -a.s. satisfies

$$\begin{aligned} \mathbb{E}_{\theta|T}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}^m - \theta)|^2) &\leq 4D^2 ((1 \vee (\mathbf{a}\mathbf{t})_{(m)}^2) (n \wedge k)) d^2 r^2 + \mathfrak{v}_{\theta|T} + 2\mathfrak{v}_T \mathbf{a}_1^2 D^2 d^2 r^2) R_{n \wedge k}^m(\mathbf{a}, \mathbf{t}, \phi) \\ &\quad + \mathfrak{v}_{\theta|T} \mathbb{E}_T^k (\mathcal{U}_{m, k}^c) + 2\mathbb{E}_T^k (\|\mathfrak{n}_{\bullet, \bullet}^m \theta^m\|_{\ell_2}^2 \mathbb{1}_{\mathcal{U}_{m, k}^c}) + |\phi_{\mathbb{N}}(\theta)|^2 \mathbb{E}_T^k (\Omega_{m, k \wedge n}^c) \end{aligned} \quad (08.45)$$

with $\Omega_{m, k \wedge n}$ and $\mathcal{U}_{m, k}$ as in (08.34) and (08.35), respectively.

§08102.81 **Proof of Proposition** §08102.80. Given in the lecture. \square

§08102.82 **Corollary.** Under the assumptions of **Proposition** §08102.80 for $k, n \in \mathbb{N}$ with $R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) \leq 1/(4D^2)$ the (generalised) tGE $\widehat{\theta}^{m_{n \wedge k}^*} = \widehat{\mathbb{T}}_{\bullet, \bullet}^{m_{n \wedge k}^*|(k \wedge n)|\dagger} \widehat{g} \in \ell_2 \mathbb{1}^{m_{n \wedge k}^*} \subseteq \text{dom}(\phi_{\mathbb{N}})$ $\mathbb{E}_{\theta|T}^{n \otimes k}$ -a.s. with $\Omega_{m_{n \wedge k}^*, k \wedge n}$ as in (08.34) and dimension $m_{n \wedge k}^*$ as in (08.44) satisfies

$$\begin{aligned} \mathbb{E}_{\theta|T}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}^{m_{n \wedge k}^*} - \theta)|^2) &\leq 4D^2 (d^2 r^2 + \mathfrak{v}_{\theta|T} + 2\mathfrak{v}_T \mathbf{a}_1^2 D^2 d^2 r^2) R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) \\ &\quad + 2^{2l} K_T^{2l} D^{2(l-1)} \{D^2 (\mathfrak{v}_{\theta|T} + \|\mathbf{a}_\bullet\|_\phi^2 r^2) k^{-1} m_{n \wedge k}^* \mathbb{t}_{m_{n \wedge k}^*}^{-2} + \mathbf{a}_1^2 D^2 d^2 r^2 / 4\} \\ &\quad \times (m_{n \wedge k}^*)^2 (k^{-1} (m_{n \wedge k}^*)^3 \mathbb{t}_{m_{n \wedge k}^*}^{-2})^{l-1}. \end{aligned} \quad (08.46)$$

and if in addition

$$(m_{n \wedge k}^*)^2 (k^{-1} (m_{n \wedge k}^*)^3 \mathbb{t}_{m_{n \wedge k}^*}^{-2})^{l-1} \leq R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) \leq 1/(4D^2) \quad (08.47)$$

then we have

$$\mathbb{E}_{\theta|T}^{n \otimes k} (|\phi_{\mathbb{N}}(\widehat{\theta}^{m_{n \wedge k}^*} - \theta)|^2) \leq 2^{2l+2} K_T^{2l} D^{2l} (\mathfrak{v}_{\theta|T} + \|\mathbf{a}_\bullet\|_\phi^2 r^2 + (1 \vee \mathbf{a}_1^2) d^2 r^2) R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi). \quad (08.48)$$

§08102.83 **Proof of Corollary** §08102.82. Given in the lecture. \square

§08102.84 **Remark.** Arguing similarly as in **Remark** §07101.56 we note that $\|\mathbf{t}^{-1}\mathbf{1}^m\|_\phi \in \mathbb{R}_{\geq 0}$ for all $m \in \mathbb{N}$ and $(\|\phi\mathbf{1}^{m+1}\|_\alpha^2 = o(1))$ as $m \rightarrow \infty$ (since $(\alpha\phi) \in \ell_2$), and hence $R_n^*(\mathbf{a}, \mathbf{t}, \phi) = o(1)$ as $n \rightarrow \infty$. If there is in addition $C \in \mathbb{R}_{\geq 1}$ such that $K_T^{2l} \leq C$ and $\mathbb{v}_{\theta|T} \leq C$ for all $\theta := T_{\cdot,\cdot}^\dagger g \in \ell_2^{\mathbf{a},r}$ and $T_{\cdot,\cdot} \in \mathbb{T}_{\mathbf{t},d,D}$ then from the bound (08.48) **Corollary** §08102.82 follows immediately

$$\sup \{ \mathbb{P}_{\theta|T}^{n \otimes k} (|\phi \nu_N(\widehat{\theta}^{m_{n\wedge k}^*} - \theta)|^2) : T_{\cdot,\cdot} \in \mathbb{T}_{\mathbf{t},d,D}, \theta \in \ell_2^{\mathbf{a},r} \} \leq R_{n\wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) \times 2^{2l+2} C D^{2l} (C + \|\mathbf{a}\|_\phi^2 r^2 + (1 \vee \alpha_1^2) d^2 r^2).$$

Note that the dimension $m_{n\wedge k}^* := m_{n\wedge k}^*(\mathbf{a}, \mathbf{t}, \phi)$ as defined in (08.44) does not depend on the unknown parameter of interest θ but on the classes $\ell_2^{\mathbf{a},r}$ and $\mathbb{T}_{\mathbf{t},d,D}$ only, and thus also the statistic $\widehat{\theta}^{m_{n\wedge k}^*}$. In other words, if the regularity of θ and $T_{\cdot,\cdot}$ is known in advance, then the thresholded GE $\widehat{\theta}^{m_{n\wedge k}^*}$ is a feasible estimator. \square

§08102.85 **Corollary** (GniSM with noisy operator §08102.08 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{T}_{\cdot,\cdot}) = (g + n^{-1/2}\dot{B}, T_{\cdot,\cdot} + k^{-1/2}\dot{W}_{\cdot,\cdot}) \sim N_{\theta|T}^{n \otimes k} = N_{\theta|T}^n \otimes N_T^k$ as in Model §08102.08, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{W}_{\cdot,\cdot} \sim N_{(0,1)}^{\otimes \mathbb{N}^2}$ are independent, $T_{\cdot,\cdot} \in \mathbb{T}$ and $\theta \in \ell_2$, and hence $g = T_{\cdot,\cdot}\theta \in \text{dom}(T_{\cdot,\cdot}^\dagger) \subseteq \ell_2$. Given Assumption §08102.75 for each $k, n \in \mathbb{N}$ fulfilling (08.47) the (generalised) tGE $\widehat{\theta}^{m_{n\wedge k}^*} = \widehat{T}_{\cdot,\cdot}^{m_{n\wedge k}^* | (k \wedge n) | \dagger} \widehat{g} \in \ell_2 \mathbf{1}_{m_{n\wedge k}^*} \subseteq \text{dom}(\phi \nu_N)$ satisfies*

$$\sup \{ N_{\theta|T}^{n \otimes k} (\|\widehat{\theta}^{m_{n\wedge k}^*} - \theta\|_\nu^2) : T_{\cdot,\cdot} \in \mathbb{T}_{\mathbf{t},d,D}, \theta \in \ell_2^{\mathbf{a},r} \} \leq R_{n\wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) \times 2^{2l+2} ((2l-1)!) D^{2l} (1 + \|\mathbf{a}\|_\phi^2 r^2 + (1 \vee \alpha_1^2) d^2 r^2)$$

where $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ is the rate in a GniSM §08101.04 (see **Corollary** §08101.66).

§08102.86 **Proof of Corollary** §08102.85. Given in the lecture. \square

§08102.87 **Corollary** (niSM with noisy operator §08102.10 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{\mathfrak{s}}_{\cdot,\cdot}) = (g + n^{-1/2}\dot{\epsilon}, T_{\cdot,\cdot} + k^{-1/2}\dot{\eta}_{\cdot,\cdot}) \sim P_{\theta|T|\sigma|\xi|\xi^{(2l)}}^{n \otimes k}$ as in Model §08102.10, where $\dot{\epsilon}$ and $\dot{\eta}_{\cdot,\cdot}$ satisfies (iSM1) with $\mathbb{v}_\sigma = \|\sigma^2\|_{\ell_\infty} \vee 1$ and (niSMnO1)–(niSMnO2) with $K_{\xi^{(2l)}}^{2l} := 1 \vee \|\xi_{\cdot,\cdot}^{(2l)}\|_{\ell_\infty(\mathbb{N}^2)}$, respectively, $T_{\cdot,\cdot} \in \mathbb{T}$ and $\theta \in \ell_2$, and hence $g = T_{\cdot,\cdot}\theta \in \text{dom}(T_{\cdot,\cdot}^\dagger) \subseteq \ell_2$. Given Assumption §08102.75 for each $k, n \in \mathbb{N}$ fulfilling (08.47) the (generalised) tGE $\widehat{\theta}^{m_{n\wedge k}^*} = \widehat{T}_{\cdot,\cdot}^{m_{n\wedge k}^* | (k \wedge n) | \dagger} \widehat{g} \in \ell_2 \mathbf{1}_{m_{n\wedge k}^*} \subseteq \text{dom}(\phi \nu_N)$ satisfies*

$$\sup \{ P_{\theta|T|\sigma|\xi|\xi^{(2l)}}^{n \otimes k} (\|\widehat{\theta}^{m_{n\wedge k}^*} - \theta\|_\nu^2) : T_{\cdot,\cdot} \in \mathbb{T}_{\mathbf{t},d,D}, \theta \in \ell_2^{\mathbf{a},r} \} \leq R_{n\wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) \times 2^{2l+2} K_{\xi^{(2l)}}^{2l} D^{2l} (\mathbb{v}_\sigma + \|\mathbf{a}\|_\phi^2 r^2 + (1 \vee \alpha_1^2) d^2 r^2)$$

where $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ is the rate in a niSM §08101.06 (see **Corollary** §08101.68).

§08102.88 **Proof of Corollary** §08102.87. Given in the lecture. \square

§08102.89 **Corollary** (nieMM with noisy operator §08102.13 continued). *Consider independent noisy versions $(\widehat{g}, \widehat{T}_{\cdot,\cdot}) = (g + n^{-1/2}\dot{\epsilon}, T_{\cdot,\cdot} + k^{-1/2}\dot{\eta}_{\cdot,\cdot})$ defined on $(\mathcal{Z}^{n+k}, \mathcal{Z}^{\otimes(n+k)}, \mathbb{P}_{\theta|T}^{n \otimes k})$ as in Model §08102.13, where $\psi \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}})$ and $\varphi_{\cdot,\cdot} \in \mathcal{M}(\mathcal{Z} \otimes 2^{\mathbb{N}^2})$ satisfies (nieMM1)–(nieMM2) for $\mathbb{v}_{\theta|T|\psi} \in \mathbb{R}_{\geq 1}$ and (nieMMnO1)–(nieMMnO3) for $K_{T|\varphi} \in \mathbb{R}_{\geq 1}$, respectively, $T_{\cdot,\cdot} \in \mathbb{T}$ and $\theta \in \ell_2$, and hence $g = T_{\cdot,\cdot}\theta \in \text{dom}(T_{\cdot,\cdot}^\dagger) \subseteq \ell_2$. Given Assumption §08102.75 for each $k, n \in \mathbb{N}$ fulfilling (08.47) the (generalised) tGE $\widehat{\theta}^{m_{n\wedge k}^*} = \widehat{T}_{\cdot,\cdot}^{m_{n\wedge k}^* | (k \wedge n) | \dagger} \widehat{g} \in \ell_2 \mathbf{1}_{m_{n\wedge k}^*} \subseteq \text{dom}(\phi \nu_N)$ satisfies*

$$\sup \{ \mathbb{P}_{\theta|T}^{n \otimes k} (\|\widehat{\theta}^{m_{n\wedge k}^*} - \theta\|_\nu^2) : T_{\cdot,\cdot} \in \mathbb{T}_{\mathbf{t},d,D}, \theta \in \ell_2^{\mathbf{a},r} \} \leq R_{n\wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) \times C_{2l} \sup \{ K_{T|\varphi}^{2l} : T_{\cdot,\cdot} \in \mathbb{T}_{\mathbf{t},d,D} \} D^{2l} (\sup \{ \mathbb{v}_{\theta|T|\psi} : T_{\cdot,\cdot} \in \mathbb{T}_{\mathbf{t},d,D}, \theta \in \ell_2^{\mathbf{a},r} \} + \|\mathbf{a}\|_\phi^2 r^2 + (1 \vee \alpha_1^2) d^2 r^2)$$

where $C_{2l} \in \mathbb{R}_{>1}$ is a constant depending on $l \in \mathbb{N}$ only and $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ is the rate in a nieMM §08101.08 (see **Corollary** §08101.70).

§08102.90 **Proof** of **Corollary** §08102.89. Given in the lecture. □

§08102.91 **Illustration**. We distinguish as in **Illustration** §08101.72 the two cases **(p)** and **(np)**. Interestingly, in case **(p)** the bound is parametric, that is, $nR_n^*(\mathbf{a}, \mathbf{t}, \phi) = O(1)$, in case **(np)** the bound is nonparametric, i.e. $\lim_{n \rightarrow \infty} nR_n^*(\mathbf{a}, \mathbf{t}, \phi) = \infty$. In case **(np)** consider similar to **(o-m)**, **(o-s)** and **(s-m)** in **Illustration** §08101.72 the following specifications:

Table 11 [§08]

Order of the rate $R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi)$ as $n \wedge k \rightarrow \infty$

$j \in \mathbb{N}$	$(\mathbf{a} \in \mathbb{R}_{>0})$	$(\mathbf{t} \in \mathbb{R}_{>0})$	(squared bias)	(variance)	$m_{n \wedge k}^*$	$R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi)$	
$\phi_j^2 = j^{2v-1}$	\mathbf{a}_j^2	\mathbf{t}_j^2	$\ \mathbf{a} \cdot \mathbb{1}_m^\perp\ _\phi^2$	$\ \mathbf{t}^{-1} \mathbb{1}_m\ _\phi^2$			
(o-m)	$v \in (-t, a)$	j^{-2a}	j^{-2t}	$m^{-2(a-v)}$	m^{2v+2t}	$(n \wedge k)^{-\frac{1}{2a+2t}}$	$(n \wedge k)^{-\frac{a-v}{a+t}}$
	$v = -t$	j^{-2a}	j^{-2t}	$m^{-2(a+t)}$	$\log m$	$\left(\frac{n \wedge k}{\log n \wedge k}\right)^{\frac{1}{2(a+t)}}$	$\frac{\log n \wedge k}{n \wedge k}$
(o-s)	$a - v \in \mathbb{R}_{>0}$	j^{-2a}	$e^{-j^{2t}}$	$m^{-2(a-v)}$	$m^{2(v-t)+e^{m^{2t}}}$	$(\log n \wedge k)^{\frac{1}{2t}}$	$(\log n \wedge k)^{-\frac{a-v}{t}}$
(s-m)	$v + t \in \mathbb{R}_{>0}$	$e^{-j^{2a}}$	j^{-2t}	$e^{-m^{2a}}$	m^{2v+2t}	$(\log n \wedge k)^{\frac{1}{2a}}$	$\frac{(\log n \wedge k)^{\frac{1+v}{a}}}{n \wedge k}$
	$v = -t$	$e^{-j^{2a}}$	j^{-2t}	$e^{-m^{2a}}$	$\log m$	$(\log n \wedge k)^{\frac{1}{2a}}$	$\frac{\log \log n \wedge k}{n \wedge k}$

We note that in case **(o-m)** and **(s-m)** for $v < -t$ the rate $R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi)$ is parametric. The tGE attains the rate $R_{n \wedge k}^* := R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi)$ due to **Corollary** §08102.82 under the additional condition

$$(k^{-1}(m_{n \wedge k}^*)^3 \mathbf{t}_{m_{n \wedge k}^*}^{-2})^{l-1} \leq (m_{n \wedge k}^*)^{-2} R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi). \tag{08.49}$$

Since $(m_{n \wedge k}^*)^{-2} R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \phi) = o(1)$ also $k^{-1}(m_{n \wedge k}^*)^3 \mathbf{t}_{m_{n \wedge k}^*}^{-2} = o(1)$ is necessary as $n \wedge k \rightarrow \infty$. The next table depicts the order of both terms in case **(o-m)**, **(o-s)** and **(s-m)**.

Table 12 [§08]

Order as $n \wedge k \rightarrow \infty$

	(o-m)	(o-s)	(s-m)
	$v \in (-t, a)$	$a - v \in \mathbb{R}_{>0}$	$v + t \in \mathbb{R}_{>0}$
$(m_{n \wedge k}^o)^{-2} R_{n \wedge k}^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$	$(n \wedge k)^{-\frac{2(a-v)+2}{2a+2t}}$	$(\log n \wedge k)^{-\frac{2a-2v+2}{2t}}$	$\frac{(\log n \wedge k)^{\frac{2t+2v-2}{2a}}}{n \wedge k}$
$(n \wedge k)^{-1} (m_{n \wedge k}^*)^3 \mathbf{t}_{m_{n \wedge k}^*}^{-2}$	$(n \wedge k)^{-\frac{2a-3}{2a+2t}}$	$(n \wedge k)^{-c}$	$\frac{(\log n \wedge k)^{\frac{2t+3}{2a}}}{n \wedge k}$

In case **(o-s)** a value $l \geq 2$ and **(s-m)** a value $l \geq 3$ is sufficient to ensure (08.49) as $n \wedge k \rightarrow \infty$. In case **(o-m)** assuming $a > 3/2$ we have $k^{-1}(m_{n \wedge k}^o)^3 \mathbf{t}_{m_{n \wedge k}^o}^{-2} = o(1)$ as $n \wedge k \rightarrow \infty$. In this situation we have (08.49) if $(2a - 3)(l - 1) > 2(a - v) + 2$ or in equal $l > (4a - 2v - 1)/(2a - 3)$. □

§09 Spectral regularisation estimator

§09100.01 **Notation**. Consider the measure space $(\mathcal{J}, \mathcal{J}, \nu)$ and the Hilbert space $\mathbb{J} = \mathbb{L}_2(\nu)$ as in **Notation** §01101.01. We suppose that $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ are surjective partial isometries, hence $VV^* = \text{id}_{\mathbb{J}} = UU^*$. As in **Definition** §03100.08 we denote for $A := VTU^* \in \mathbb{L}(\mathbb{J})$ its Moore-Penrose inverse by $A^\dagger : \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$. □

§09100.02 **Assumption**. For $\mathbb{J} = \mathbb{L}_2(\nu)$ let $U \in \mathbb{L}(\mathbb{H}, \mathbb{J})$ and $V \in \mathbb{L}(\mathbb{G}, \mathbb{J})$ be surjective partial isometries fixed and presumed to be *known* in advance, let $T \in \mathbb{L}(\mathbb{H}, \mathbb{G})$, hence $A = VTU^* \in \mathbb{L}(\mathbb{J})$ with Moore-Penrose inverse $A^\dagger : \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ and let $g \in \text{dom}(A^\dagger)$, and hence $\theta = A^\dagger g \in \mathbb{J}$. □

§09100.03 **Reminder.** Under Assumption §09100.02 we consider $\theta \in \mathbb{J}$ and $A \in \mathbb{L}(\mathbb{J})$ and hence $g = A\theta \in \text{ran}(A) \subseteq \text{dom}(A^\dagger)$. Let $\{r_\alpha: \alpha \in (0, 1)\}$ be a collection of real-valued Borel-measurable functions defined on $[0, \|T_\cdot\|_{\mathbb{L}(\mathbb{J})}^2]$ satisfying (see §06102.01)

(sR1) for all $\alpha \in (0, 1)$ there exists $C_\alpha \in \mathbb{R}_{>0}$ such that $|r_\alpha(x)| \leq C_\alpha$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$,

(sR2) for all $x \in (0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$ holds $|1 - xr_\alpha(x)| = o(1)$ as $\alpha \rightarrow 0$, and

(sR3) there is $K \in \mathbb{R}_{>0}$ such that $|xr_\alpha(x)| \leq K$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$ and $\alpha \in (0, 1)$,

Then the collection $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ of operators is called *spectral regularisation* of $A^\dagger: \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$. \square

§09|01 Statistical inverse problem

§09101.01 **Assumption.** Consider a random function $\tilde{g} \in \mathcal{M}(\mathcal{A}, \mathcal{B}_1)$ on a measurable space (Ω, \mathcal{A}) with values in \mathbb{J} (Definition §01101.17). Let Assumption §08100.02 be satisfied where $A \in \mathbb{L}(\mathbb{J})$ is known in advance. For $\theta \in \mathbb{J}$, hence image $g = A\theta \in \mathbb{J}$, and probability measure $\mathbb{P}_{\theta|A} \in \mathcal{W}(\mathcal{A})$ on (Ω, \mathcal{A}) the random function \tilde{g} has a finite second moment (i.e. $\mathbb{E}_{\theta|A}(\|\tilde{g}\|_{\mathbb{J}}^2) \in \mathbb{R}_{>0}$). \square

§09101.02 **Definition.** Under Assumption §09101.01 for $\theta \in \mathbb{J}$, $A \in \mathbb{L}(\mathbb{J})$, and a continuous spectral regularisation $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ of A^\dagger as in Definition §06102.01 we call $\tilde{\theta}^\alpha = R_\alpha \tilde{g}$ *spectral regularisation estimator (sRE)* of θ . \square

§09101.03 **Comment.** Since $g = A\theta \in \text{dom}(A^\dagger)$ and hence $\theta = A^\dagger g$ the spectral regularised approximation $\theta^\alpha := R_\alpha g = r_\alpha(A^*A)A^*g \in \mathbb{J}$ converges to θ as $\alpha \rightarrow 0$, i.e. the approximation error $\|\theta^\alpha - \theta\|_{\mathbb{J}}$ converges to zero as $\alpha \rightarrow 0$ (compare Proposition §06102.02). \square

§09|01|01 Global risk

§09101.04 **Lemma (\mathbb{J} -consistency).** Let $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ be a continuous spectral regularisation of A^\dagger as in Definition §06102.01. Assume Definition §06102.01 (sR1) and (sR2), and in addition replace (sR3) by

(sR3a) for all $s \in [0, 1]$ there exists $K_s \in \mathbb{R}_{>0}$ such that $x^s |r_\alpha(x)| \leq C_s \alpha^{s-1}$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$ and $\alpha \in (0, 1)$.

Under Assumption §09101.01 a sRE $\tilde{\theta}^\alpha = R_\alpha \tilde{g}$ of $\theta = A^\dagger g \in \mathbb{J}$ satisfies for all $\alpha \in (0, 1)$

$$\mathbb{E}_{\theta|A}(\|\tilde{\theta}^\alpha - \theta\|_{\mathbb{J}}^2) \leq 2K_{1/2}^2 \alpha^{-1} \mathbb{E}_{\theta|A}(\|\tilde{g} - g\|_{\mathbb{J}}^2) + 2\|\theta^\alpha - \theta\|_{\mathbb{J}}^2 \quad (09.01)$$

If \tilde{g} is a \mathbb{J} -consistent estimator of g , that is $\mathbb{E}_{\theta|A}^n(\|\tilde{g} - g\|_{\mathbb{J}}^2) = o(1)$ as $n \rightarrow \infty$, then

$$\mathbb{E}_{\theta|A}^n(\|\tilde{\theta}^{\alpha_n} - \theta\|_{\mathbb{J}}^2) = o(1) \quad \text{as } n \rightarrow \infty$$

for any sequence $(\alpha_n)_{n \in \mathbb{N}}$ such that $\alpha_n = o(1)$ and $\alpha_n^{-1} \mathbb{E}_{\theta|A}^n(\|\tilde{g} - g\|_{\mathbb{J}}^2) = o(1)$ as $n \rightarrow \infty$.

§09101.05 **Proof of Lemma §09101.04.** Given in the lecture. \square

§09101.06 **Reminder.** Given $A \in \mathbb{L}(\mathbb{J})$ let $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ be a spectral regularisation of $A^\dagger: \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ as in Definition §06102.01. Assume Definition §06102.01 (sR1), and (sR3), and in addition replace (sR2) by

(sR2a) there are $s_0 \in \mathbb{R}_{>1}$, $C_s \in \mathbb{R}_{>0}$ for all $s \in [0, s_0]$ such that $x^s |1 - xr_\alpha(x)| \leq C_s \alpha^s$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}^2]$ and $\alpha \in (0, 1)$.

For $\theta \in \mathbb{J}$, $g = A\theta \in \text{dom}(A^\dagger)$, and $\alpha \in (0, 1)$ consider $\theta^\alpha = R_\alpha g = r_\alpha(A^*A)A^*g \in \mathbb{J}$. If $\theta = A^\dagger g \in \mathbb{J}$ fulfills a source condition as in [Definition §06102.05](#), that is, there are $s \in [0, 2s_s]$ and $h_s \in \mathbb{J}$ such that $\theta = (A^*A)^{s/2}h_s$ or in equal $\theta \in \text{ran}((A^*A)^{s/2})$, then we have

$$\|\theta^\alpha - \theta\|_{\mathbb{J}} \leq C_{s/2} \alpha^{s/2} \|h_s\|_{\mathbb{J}} \quad \forall \alpha \in (0, 1) \quad (09.02)$$

due to [Proposition §06102.06](#). □

§09101.07 **Corollary.** *Let the assumptions of [Lemma §09101.04](#) be satisfied and in addition let (sR2) be replaced by (sR2a). If $\theta = A^\dagger g \in \mathbb{J}$ fulfills a source condition as in [Definition §06102.05](#), that is, there are $s \in [0, 2s_s]$ and $h_s \in \mathbb{J}$ such that $\theta = (A^*A)^{s/2}h_s$, then the sRE $\tilde{\theta}^{\alpha_n} = R_{\alpha_n} \tilde{g}$ of $\theta = A^\dagger g \in \mathbb{J}$ with $\alpha_n := (\mathbb{E}_{\theta|A}(\|\tilde{g} - g\|_{\mathbb{J}}^2))^{1/(1+s)}$ fulfills*

$$\mathbb{E}_{\theta|A}(\|\tilde{\theta}^{\alpha_n} - \theta\|_{\mathbb{J}}^2) \leq 2(K_{1/2}^2 + C_{s/2}^2 \|h_s\|_{\mathbb{J}}^2) (\mathbb{E}_{\theta|A}(\|\tilde{g} - g\|_{\mathbb{J}}^2))^{s/(1+s)}. \quad (09.03)$$

§09101.08 **Proof of [Corollary §09101.07](#).** Given in the lecture. □

§09101.09 **Reminder.** Given $A \in \mathbb{L}^{\geq}(\mathbb{J})$, i.e., A is positive definite, we eventually consider as in [Notation §06102.19](#) a spectral regularisation $\{R_\alpha := r_\alpha(A) \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ of A^\dagger for a given collection $\{r_\alpha: \alpha \in (0, 1)\}$ of real-valued Borel-measurable functions defined on $[0, \|A\|_{\mathbb{L}(\mathbb{J})}]$ satisfying (sR1') for all $\alpha \in (0, 1)$ there exists $C_\alpha \in \mathbb{R}_{\geq 0}$ such that $|r_\alpha(x)| \leq C_\alpha$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}]$, (sR2'a) there are $s_s \in [1, \infty)$ and $C_s \in \mathbb{R}_{> 0}$ for all $s \in [0, s_s]$ such that $x^s |1 - xr_\alpha(x)| \leq C_s \alpha^s$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}]$ and $\alpha \in (0, 1)$, (sR3') there is $K \in \mathbb{R}_{> 0}$ such that $|xr_\alpha(x)| \leq K$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}]$ and $\alpha \in (0, 1)$.

We consider the spectral regularised approximation $\theta^\alpha = R_\alpha g = r_\alpha(A)g \in \mathbb{J}$ of $\theta := A^\dagger g \in \mathbb{J}$ for $g \in \text{dom}(A^\dagger)$. Under Assumption §09101.01 we call $\tilde{\theta}^\alpha = R_\alpha \tilde{g}$ *spectral regularisation estimator (sRE)* of θ . □

§09101.10 **Lemma (\mathbb{J} -consistency).** *Given $A \in \mathbb{L}^{\geq}(\mathbb{J})$ let $\{R_\alpha := r_\alpha(A) \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ be a continuous spectral regularisation of A^\dagger as in [Notation §06102.19](#). Assume [Notation §06102.19](#) (sR1') and (sR2'a), and in addition replace (sR3') by*

(sR3'a) *for all $s \in [0, 1]$ there exists $K_s \in \mathbb{R}_{> 0}$ such that $x^s |r_\alpha(x)| \leq K_s \alpha^{s-1}$ for all $x \in [0, \|A\|_{\mathbb{L}(\mathbb{J})}]$ and $\alpha \in (0, 1)$.*

Under Assumption §09101.01 a sRE $\tilde{\theta}^{\alpha_n} = R_{\alpha_n} \tilde{g}$ of $\theta = A^\dagger g \in \mathbb{J}$ satisfies for all $\alpha \in (0, 1)$

$$\mathbb{E}_{\theta|A}(\|\tilde{\theta}^{\alpha_n} - \theta\|_{\mathbb{J}}^2) \leq 2K_0^2 \alpha^{-2} \mathbb{E}_{\theta|A}(\|\tilde{g} - g\|_{\mathbb{J}}^2) + 2\|\theta^\alpha - \theta\|_{\mathbb{J}}^2 \quad (09.04)$$

If \tilde{g} is a \mathbb{J} -consistent estimator of g , that is $\mathbb{E}_{\theta|A}^n(\|\tilde{g} - g\|_{\mathbb{J}}^2) = o(1)$ as $n \rightarrow \infty$, then

$$\mathbb{E}_{\theta|A}^n(\|\tilde{\theta}^{\alpha_n} - \theta\|_{\mathbb{J}}^2) = o(1) \quad \text{as } n \rightarrow \infty$$

for any sequence $(\alpha_n)_{n \in \mathbb{N}}$ such that $\alpha_n = o(1)$ and $\alpha_n^{-2} \mathbb{E}_{\theta|A}^n(\|\tilde{g} - g\|_{\mathbb{J}}^2) = o(1)$ as $n \rightarrow \infty$.

§09101.11 **Proof of [Lemma §09101.10](#).** Given in the lecture. □

§09101.12 **Reminder.** Given $A \in \mathbb{L}^{\geq}(\mathbb{J})$ let $\{R_\alpha := r_\alpha(A) \in \mathbb{L}(\mathbb{J}): \alpha \in (0, 1)\}$ be a spectral regularisation of $A^\dagger: \mathbb{J} \supseteq \text{dom}(A^\dagger) \rightarrow \mathbb{J}$ as in [Notation §06102.19](#). Assume [Notation §06102.19](#) (sR1'), (sR2'a), and (sR3'). For $\theta \in \mathbb{J}$, $g = A\theta \in \text{dom}(A^\dagger)$, and $\alpha \in (0, 1)$ consider $\theta^\alpha = R_\alpha g = r_\alpha(A)g \in \mathbb{J}$. If

$\theta = A^\dagger g \in \mathbb{J}$ fulfills a source condition as in [Definition §06102.05](#), that is, there are $s \in [0, 2s_s]$ and $h_s \in \mathbb{J}$ such that $\theta = A^s h_s$ or in equal $\theta \in \text{ran}(A^s)$, then we have

$$\|\theta^\alpha - \theta\|_{\mathbb{J}} \leq C_s \alpha^s \|h_s\|_{\mathbb{J}} \quad \forall \alpha \in (0, 1) \quad (09.05)$$

due to [Proposition §06102.20](#). □

§09101.13 **Corollary.** *Let the assumptions of [Lemma §09101.10](#) be satisfied. If $\theta = A^\dagger g \in \mathbb{J}$ fulfills a source condition as in [Definition §06102.05](#), that is, there are $s \in [0, 2s_s]$ and $h_s \in \mathbb{J}$ such that $\theta = (A)^s h_s$, then the sRE $\tilde{\theta}^{\alpha_s} = R_{\alpha_s} \tilde{g}$ of $\theta = A^\dagger g \in \mathbb{J}$ with $\alpha_s := (\mathbb{P}_{\theta|A}(\|\tilde{g} - g\|_{\mathbb{J}}^2))^{1/(2+2s)}$ fulfills*

$$\mathbb{P}_{\theta|A}(\|\tilde{\theta}^{\alpha_s} - \theta\|_{\mathbb{J}}^2) \leq 2(K_0^2 + C_s^2 \|h_s\|_{\mathbb{J}}^2) (\mathbb{P}_{\theta|A}(\|\tilde{g} - g\|_{\mathbb{J}}^2))^{s/(1+s)}. \quad (09.06)$$

§09101.14 **Proof of [Corollary §09101.13](#).** Given in the lecture. □

§09|01|02 Maximal global v-risk

§09101.15 **Assumption.** Consider the separable Hilbert space $\mathbb{J} = \mathbb{L}_2(\mathcal{J}, \mathcal{F}, \nu)$ with σ -algebra \mathcal{F} over \mathcal{J} containing all elementary events $\{j\}$, $j \in \mathcal{J}$, and all events $\llbracket m \rrbracket := [-m, m] \cap \mathcal{J}$, $m \in \mathbb{N}$, and with σ -finite measure $\nu \in \mathcal{M}_\sigma(\mathcal{F})$ such that $\nu(\llbracket m \rrbracket) \in \mathbb{R}_{>0}$, for all $m \in \mathbb{N}$. Let Assumption §09100.02 be satisfied where $A \in \mathbb{L}(\mathbb{J})$ is known in advance. For $\theta \in \mathbb{J}$, and hence image $g = A\theta \in \mathbb{J}$, let $\mathbb{P}_{\theta|A} \in \mathcal{W}(\mathcal{A})$ be a probability measure on (Ω, \mathcal{A}) . Consider a stochastic process $\dot{\epsilon} = (\dot{\epsilon}_j)_{j \in \mathcal{J}}$ on (Ω, \mathcal{A}) satisfying Assumption §01101.04 (i.e. $\dot{\epsilon} \in \mathcal{M}(\mathcal{A} \otimes \mathcal{F})$) which for each $\theta \in \mathbb{J}$ and $A \in \mathbb{L}(\mathbb{J})$ in addition fulfills

$$\text{(SIPg1)} \quad \dot{\epsilon}_j \in \mathcal{L}_1(\mathbb{P}_{\theta|A}) := \mathcal{L}_1(\Omega, \mathcal{A}, \mathbb{P}_{\theta|A}) \text{ for all } j \in \mathcal{J} \text{ and } \mathbb{P}_{\theta|A}(\dot{\epsilon}) = (\mathbb{P}_{\theta|A}(\dot{\epsilon}_j))_{j \in \mathcal{J}} = 0,$$

$$\text{(SIPg2)} \quad \mathbb{V}^{\theta|A} := \mathbb{P}_{\theta|A}(\dot{\epsilon}^2) := (\mathbb{V}_j^{\theta|A} := \mathbb{P}_{\theta|A}(\dot{\epsilon}_j^2))_{j \in \mathcal{J}} \in \mathbb{L}_\infty(\nu) \text{ and}$$

$$\text{(SIPg3)} \quad \dot{\epsilon} \cdot \mathbb{1}^m \in \mathbb{L}_\infty(\nu) \text{ } \mathbb{P}_{\theta|A}\text{-a.s., for each } m \in \mathbb{N}.$$

Given a sample size $n \in \mathbb{N}$ the observable noisy image with mean $g = A\theta \in \mathbb{J}$ takes the form $\hat{g} = g + n^{-1/2} \dot{\epsilon}$. We denote by $\mathbb{P}_{\theta|A}^n$ the distribution of \hat{g} . □

§09101.16 **Comment.** Under Assumption §09101.15 we have $\dot{\epsilon} \cdot \mathbb{1}^m \in \mathbb{J}$ $\mathbb{P}_{\theta|A}$ -a.s.. Since $g \in \mathbb{J}$, and hence $g^m = g \cdot \mathbb{1}^m \in \mathbb{J}$ ([Property §04103.09](#)), it follows

$$\hat{g}^m = \hat{g} \cdot \mathbb{1}^m = n^{-1/2} \dot{\epsilon} \cdot \mathbb{1}^m + g^m \in \mathbb{J} \quad \mathbb{P}_{\theta|A}^n\text{-a.s..} \quad (09.07)$$

If $\mathcal{J} \subseteq \mathbb{Z}$ (at most countable) and $\nu_{\mathcal{J}}$ is the counting measure over the index set \mathcal{J} then Assumption §01101.04 and (SIPg2) $\mathbb{V}^{\theta|A} = \mathbb{P}_{\theta|A}(\dot{\epsilon}^2) \in \mathbb{L}_\infty(\nu_{\mathcal{J}})$ implies the additional assumption (SIPg3) $\dot{\epsilon} \cdot \mathbb{1}^m \in \mathbb{L}_\infty(\nu_{\mathcal{J}})$ $\mathbb{P}_{\theta|A}^n$ -a.s.. However, the last implication does generally not hold, if $\mathcal{J} \in \{\mathbb{R}, \mathbb{R}_{>0}\}$ for example. □

§09101.17 **Assumption.** Consider $\mathfrak{v} \in \mathcal{M}_{>0, \nu}(\mathcal{F}) \cap \mathbb{L}_\infty(\nu)$, and for $t \in \mathbb{R}_{>0}$, $a \in (0, t]$ set $\mathfrak{t} := \mathfrak{v}^t$ and $\mathfrak{a} := \mathfrak{v}^a$ where $\mathfrak{t}, \mathfrak{a} \in \mathcal{M}_{>0, \nu}(\mathcal{F}) \cap \mathbb{L}_\infty(\nu)$ and hence $\nu(\mathcal{N}_{\mathfrak{v}}) = \nu(\mathcal{N}_{\mathfrak{t}}) = \nu(\mathcal{N}_{\mathfrak{a}}) = 0$. □

§09101.18 **Reminder.** Under Assumption §09101.17 we have $\mathbb{J}^a = \mathbb{L}_2(\nu) = \text{dom}(M_{\cdot^{-1}}) = \mathfrak{J}^a = \mathbb{L}_2(\mathfrak{a}^{-2}\nu)$ and the measures ν , $\mathfrak{v}^2\nu$, $\mathfrak{t}^2\nu$ and $\mathfrak{a}^{-2}\nu$ dominate mutually each other (see [Property §04101.02](#)). Consequently, $\mathbb{J}^a \subseteq \mathbb{J} = \mathbb{L}_2(\nu)$ and $\mathbb{J}^a \subseteq \mathbb{L}_2(\mathfrak{v}^2\nu)$ ([Property §04102.11](#)) since $(a\mathfrak{v}) = \mathfrak{v}^{1+a} \in \mathbb{L}_\infty(\nu)$. We assume in the following that $\theta \in \mathbb{J}$ satisfies an abstract smoothness condition ([Definition §04102.12](#)), i.e., there is $r \in \mathbb{R}_{>0}$ such that $\theta \in \mathbb{J}^{a,r} = \{h \in \mathbb{J}^a : \|h\|_{\mathfrak{a}^{-1}} \leq r\} \subseteq \mathbb{J}^a \subseteq \mathbb{J}$. Under Assumption §06102.11 by [Corollary §05101.14](#) (see [Comment §05101.16](#)) if $A \in \mathbb{T}_{t,d}$ (or in equal $(A^*A)^{1/2} \in \mathbb{T}_{t,d}^{\geq}$) then (i) for any $\theta \in \mathbb{J}^a$ we have $\theta = (A^*A)^{a/(2t)} h$ with $\|h\|_{\mathbb{J}} \leq d^{a/t} \|\theta\|_{\mathfrak{a}^{-1}}$, and conversely

(ii) for any $\theta = (A^*A)^{a/(2t)} h_*$ with $h_* \in \mathbb{L}_2(\nu)$ we obtain $\theta \in \mathbb{J}^a$ with $\|\theta\|_{\mathbb{J}^{a-1}} \leq d^{a/t} \|h_*\|_{\mathbb{J}}$. In particular since $(\text{ta})_* = \mathbf{v}^{t+a} \in \mathcal{M}_{>0,\nu}(\mathcal{J}) \cap \mathbb{L}_\infty(\nu)$ if $A \in \mathbb{T}_{t,d}$ and $\theta \in \mathbb{J}^{a,r}$, then due to **Corollary** §06102.13 we have $g_* = A\theta \in \mathbb{J}^{(ta),dr}$. \square

§09101.19 **Notation (Reminder)**. For sequences $a_n, b_n \in (\mathbb{K})^{\mathbb{N}}$ taking its values in $\mathbb{K} \in \{\mathbb{R}, \mathbb{R}_{>0}, \mathbb{Q}, \mathbb{Z}, \dots\}$ we write $a_* \in (\mathbb{K})_{>}^{\mathbb{N}}$ and $b_* \in (\mathbb{K})_{\leq}^{\mathbb{N}}$ if a_* and b_* , respectively, is monotonically *non-decreasing* and *non-increasing*. If in addition $a_n \rightarrow \infty$ and $b_n \rightarrow 0$ as $n \rightarrow \infty$, then we write $a_* \in (\mathbb{K})_{\uparrow\infty}^{\mathbb{N}}$ and $b_* \in (\mathbb{K})_{\downarrow 0}^{\mathbb{N}}$ for short. For $w_* \in \mathbb{L}_\infty(\nu)$ we set $w_{(0)} := \|w_*\|_{\mathbb{L}_\infty(\nu)}$ and $w_{(\bullet)} = (w_{(j)} := \|w_* \mathbf{1}_j^{\perp}\|_{\mathbb{L}_\infty(\nu)})_{j \in \mathbb{N}}$, where by construction $w_{(\bullet)} \in (\mathbb{R}_{>0})_{\leq}^{\mathbb{N}}$. \square

§09101.20 **Corollary**. Under Assumptions §09101.15 and §09101.17 setting for $n, m \in \mathbb{N}$

$$\begin{aligned} R_n^m((\text{at})_*) &:= [(\text{at})_{(m)} \vee n^{-1}m], \quad m_n^* := \arg \min \{R_n^m((\text{at})_*) : m \in \mathbb{N}\} \\ \text{and } R_n^*(\text{at})_* &:= R_n^{m_n^*}((\text{at})_*) = \min \{R_n^m((\text{at})_*) : m \in \mathbb{N}\} \end{aligned} \quad (09.08)$$

and $\|\mathbf{v}_*^{\theta|A}\|_{\mathbb{L}_\infty(\nu)} \leq \mathbf{v}_{\theta|A} \in \mathbb{R}_{>0}$, for $A \in \mathbb{T}_{t,d}$ and for all $\theta \in \mathbb{J}^{a,r}$, hence $g_* = A\theta \in \text{dom}(A^\dagger) \subseteq \mathbb{J}$, the orthogonal projection estimator (OPE) $\widehat{g}_*^m := \widehat{g}_* \mathbf{1}_*^m$ fulfills

$$\mathbb{E}_{\theta|A}^n (\|\widehat{g}_*^m - g_*\|_{\mathbb{J}}^2) \leq (\mathbf{v}_{\theta|A} + d^2 r^2) R_n^m((\text{at})_*) \quad \forall m, n \in \mathbb{N}$$

and hence $\mathbb{E}_{\theta|A}^n (\|\widehat{g}_*^{m_n^*} - g_*\|_{\mathbb{J}}^2) \leq (\mathbf{v}_{\theta|A} + d^2 r^2) R_n^*(\text{at})_*$.

§09101.21 **Proof of Corollary** §09101.20. Given in the lecture. \square

Consider the OPE $\widehat{g}_*^m := \widehat{g}_* \mathbf{1}_*^m$ for the orthogonal projection $g_*^m = g_* \mathbf{1}_*^m \in \mathbb{J} \mathbf{1}_*^m$ of $g_* = A\theta \in \mathbb{J}$. Given a continuous spectral regularisation $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}) : \alpha \in (0, 1)\}$ of A^\dagger as in **Definition** §06102.01 We measure the accuracy of the sRE $\widehat{\theta}_*^{\alpha,m} = R_\alpha \widehat{g}_*^m$ of $\theta_* = A^\dagger g_* \in \mathbb{J}$ by the mean of its global \mathbf{v} -error introduced in §04103101, i.e. its \mathbf{v} -risk.

§09101.22 **Proposition**. Under Assumptions §09101.15 and §09101.17 with $\|\mathbf{v}_*^{\theta|A}\|_{\mathbb{L}_\infty(\nu)} \leq \mathbf{v}_{\theta|A} \in \mathbb{R}_{>0}$ let $\{R_\alpha := r_\alpha(A^*A)A^* \in \mathbb{L}(\mathbb{J}) : \alpha \in (0, 1)\}$ be a continuous spectral regularisation of A^\dagger as in **Definition** §06102.01 and in addition replace (sR2) and (sR3) by (sR2a) and (sR3a), respectively. Consider for $m \in \mathbb{N}$ and $\alpha \in (0, 1)$ the sRE $\widehat{\theta}_*^{\alpha,m} = R_\alpha \widehat{g}_*^m$. If $\theta_* \in \mathbb{J}^{a,r}$ and $A \in \mathbb{T}_{t,d}$ then for all $m \in \mathbb{N}$ and $\alpha \in (0, 1)$ we have

$$\begin{aligned} \mathbb{E}_{\theta|A}^n (\|\widehat{\theta}_*^{\alpha,m} - \theta_*\|_{\mathbf{v}^q}^2) &\leq [\alpha^{(a+q)/t} \vee \alpha^{(q-t)/t} R_n^m(\mathbf{v}^{a+t})] \\ &\quad \times 2d^{2|q|/t} \{C_{(q+a)/(2t)}^2 d^{2(a+|q|)/t} r^2 + K_{(q+t)/(2t)}^2 (\mathbf{v}_{\theta|A} + d^2 r^2)\}. \end{aligned} \quad (09.09)$$

§09101.23 **Proof of Proof** §09101.23. Given in the lecture. \square

§09101.24 **Corollary**. Under the assumptions of **Proposition** §09101.22 the SRE $\widehat{\theta}_*^{\alpha_n, m_n^*} := R_{\alpha_n} \widehat{g}_*^{m_n^*}$ with m_n^* and $R_n^*(\mathbf{v}^{p+a})$ as in (09.08) (**Corollary** §09101.20 using $(\text{at})_* = \mathbf{v}^{a+t}$) and $\alpha_n^* := (R_n^*(\mathbf{v}^{a+t}))^{t/(a+t)}$ for all n satisfies

$$\begin{aligned} \mathbb{E}_{\theta|A}^n (\|\widehat{\theta}_*^{\alpha_n, m_n^*} - \theta_*\|_{\mathbf{v}^q}^2) &\leq (R_n^*(\mathbf{v}^{a+t}))^{(a+q)/(a+t)} \\ &\quad \times 2d^{2|q|/t} \{C_{(q+a)/(2t)}^2 d^{2(a+|q|)/t} r^2 + K_{(q+t)/(2t)}^2 (\mathbf{v}_{\theta|A} + d^2 r^2)\}. \end{aligned} \quad (09.10)$$

§09101.25 **Proof of Corollary** §09101.24. Given in the lecture. \square

Chapter 4

Minimax optimal estimation

We present a general approach to derive lower bounds and thus in combination with the upper bounds *Chapter 3* establish minimax optimality.

Overview

§10	Minimax theory: a general approach	123
§11	Deriving a lower bound	127
§11 01	Lower bound based on two hypothesis	128
§11 02	Lower bound based on m hypothesis	133

§10 Minimax theory: a general approach

Suppose that the function of interest θ belongs to a class $\Theta \subseteq \mathbb{H}$. For each noise level $n \in \mathbb{N}$ let $\mathbb{P}_\Theta^n := (\mathbb{P}_\theta^n)_{\theta \in \Theta}$ denote a family of probability measures and let \mathbb{E}_θ^n be the expectation with respect to the measure \mathbb{P}_θ^n in \mathbb{P}_Θ^n . Moreover, we assume that the probability measure associated with an observable quantity belongs to \mathbb{P}_Θ^n .

§10|00.01 **GdSM (§01|03.06 continued).** Given $\ell_2 = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ consider a Gaussian direct sequence model (GdSM) as in §01|03.06. Here the observable stochastic process $\hat{\theta} = \theta + n^{-1/2}\dot{B} \sim N_\theta^n$ is a noisy version of $\theta \in \Theta \subseteq \ell_2$ and $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$. Consequently, $\hat{\theta}$ admits a N_θ^n -distribution belonging to the family $N_\Theta^n := (N_\theta^n)_{\theta \in \Theta}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, N_\Theta^n)$ where $\Theta \subseteq \ell_2$. □

Assume furthermore, that an estimator $\tilde{\theta}$ of θ based on observable quantities is available which takes its values in \mathbb{H} but does not necessarily belong to Θ . We shall measure the accuracy of any estimator $\tilde{\theta}$ of θ by its distance $\mathfrak{d}_{\text{ist}}(\tilde{\theta}, \theta)$ where $\mathfrak{d}_{\text{ist}}(\cdot, \cdot)$ is a certain semi metric to be specified below. Moreover, we call the quantity $\mathbb{E}_\theta^n(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta))$ risk of the estimator $\tilde{\theta}$ of θ .

§10|00.02 **Definition.** Given an estimator $\tilde{\theta}$ of a function of interest θ belonging to a class of solutions Θ based on observable quantities with probability measure $\mathbb{P}_\theta^n \in \mathbb{P}_\Theta^n$ we call

$$\mathcal{R}_n[\tilde{\theta} | \Theta] := \sup \{ \mathbb{E}_\theta^n(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta \}$$

its *maximal risk* over Θ . □

§10|00.03 **Remark.** An advantage of taking a maximal risk instead of a risk is that the former does not depend on the unknown function θ . Imagine we would have taken a constant estimator, say $\tilde{\theta} = h$, of θ . This would be the perfect estimator if by chance $\theta = h$, but in all other cases this estimator is likely to perform poorly. Therefore it is reasonable to consider the supremum over the whole class of possible functions in order to get consolidated findings. However, considering the maximal risk may be a very pessimistic point of view. □

§10100.04 **Definition.** Consider a maximal risk $\mathcal{R}_n[\bullet|\Theta]$ over a family \mathbb{P}_Θ^n of probability measures. Let $\widehat{\theta}$ be an estimator of $\theta \in \Theta$, $C \in \mathbb{R}_{>0}$ and for each $n \in \mathbb{N}$ let $R_n^* \in \mathbb{R}_{\geq 0}$ satisfy
(lower) R_n^* is a *lower bound* up to the constant C^{-1} of the maximal risk over Θ , that is

$$\inf_{\widehat{\theta}} \mathcal{R}_n[\widehat{\theta}|\Theta] \geq C^{-1} R_n^*$$

where the infimum is taken over all possible estimators of θ ;

(upper) R_n^* is an *upper bound* up to the constant C of the maximal risk over Θ , that is

$$\mathcal{R}_n[\widehat{\theta}|\Theta] \leq C R_n^*$$

Then we call R_n^* *minimax-bound* and the estimator $\widehat{\theta}$ *minimax-optimal* (up to the constant C). As a consequence, up to the constant C^2 the estimator $\widehat{\theta}$ attains the lower maximal risk bound that is, $\mathcal{R}_n[\widehat{\theta}|\Theta] \leq C^2 \inf_{\widehat{\theta}} \mathcal{R}_n[\widehat{\theta}|\Theta]$. \square

§10100.05 **Remark.** We call a minimax-bound $(R_n^*)_{n \in \mathbb{N}}$ a *minimax-optimal rate* (of convergence) if in addition $R_n^* = o(1)$ as $n \rightarrow \infty$. It is worth noting that a minimax-optimal rate is not unique since every other rate that is equivalent of order is also minimax-optimal. \square

§10100.06 **dSM (§01103.05 continued).** Given $\ell_2 = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ consider a Direct sequence model (dSM) as in §01103.05. Here the observable stochastic process $\widehat{\theta}_j = \theta_j + n^{-1/2} \dot{\epsilon}_j$ is a noisy version of $\theta_j \in \Theta \subseteq \ell_2$ and $\dot{\epsilon}_j \sim \otimes_{j \in \mathbb{N}} \mathbb{P}^{\dot{\epsilon}_j}$, where

(SM:ub) for $\sigma_j \in \Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$ and $\mathbb{P}_{(0,1)} \in \mathcal{W}_2(\mathcal{B})$ we have $\mathbb{P}^{\dot{\epsilon}_j} = \mathbb{P}_{(0,\sigma_j)}$ for all $j \in \mathbb{N}$,

Under (SM:ub) $\widehat{\theta}_j$ admits a $\mathbb{P}_{\theta|\sigma}^n$ -distribution belonging to the family $\mathbb{P}_{\Theta \times \Sigma}^n := (\mathbb{P}_{\theta|\sigma}^n)_{\theta \in \Theta, \sigma \in \Sigma}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, \mathbb{P}_{\Theta \times \Sigma}^n)$ where $\Theta \subseteq \ell_2$ and $\Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$. \square

More generally, given a class of solutions Θ , a class of nuisances parameters Ξ and a noise level $n \in \mathbb{N}$ let $\mathbb{P}_{\Theta \times \Xi}^n := (\mathbb{P}_{\theta|\xi}^n)_{\theta \in \Theta, \xi \in \Xi}$ denote a family of probability measures. Moreover, we assume that the probability measure associated with an observable quantity belongs to $\mathbb{P}_{\Theta \times \Xi}^n$. Note that dismissing in Model §10100.06 compared to Model §10100.01 the assumption of a known sequence of variances σ_j^2 the class of nuisances parameters Ξ equals Σ .

§10100.07 **Definition.** Given an estimator $\widehat{\theta}$ of a function of interest θ belonging to a class of solutions Θ based on observable quantities with probability measure $\mathbb{P}_{\theta|\xi}^n \in \mathbb{P}_{\Theta \times \Xi}^n$ we call

$$\mathcal{R}_n[\widehat{\theta}|\Theta, \Xi] := \sup \left\{ \mathbb{E}_{\theta|\xi}^n (\mathfrak{d}_{\text{ist}}^2(\widehat{\theta}, \theta)) : \theta \in \Theta, \xi \in \Xi \right\}$$

its *maximal risk* over $\Theta \times \Xi$. \square

§10100.08 **diSM (§01104.08 continued).** Given $\ell_2 = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ and $\ell_\infty = \mathbb{L}_\infty(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ consider a Diagonal inverse sequence model (diSM) as in §01104.08 where $\mathfrak{s}_j \in \ell_\infty$ is *known* in advance. Here the observable stochastic process $\widehat{g}_j = g_j + n^{-1/2} \dot{\epsilon}_j$ is a noisy version of $g_j = \mathfrak{s}_j \theta_j \in \ell_2$ with $\theta_j = \mathfrak{s}_j^\dagger g_j \in \Theta \subseteq \ell_2$ and $\dot{\epsilon}_j \sim \otimes_{j \in \mathbb{N}} \mathbb{P}^{\dot{\epsilon}_j}$, where $\dot{\epsilon}_j$ satisfies (SM:ub) in Model §10100.06 for $\Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$. Under (SM:ub) \widehat{g}_j admits a $\mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n$ -distribution belonging to the family $\mathbb{P}_{\Theta \times \{\mathfrak{s}_j\} \times \Sigma}^n := (\mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n)_{\theta \in \Theta, \mathfrak{s} \in \Sigma}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, \mathbb{P}_{\Theta \times \{\mathfrak{s}_j\} \times \Sigma}^n)$ where $\Theta \subseteq \ell_2$ and $\Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$. \square

Given some transformation \mathbb{T} defined on \mathbb{H} let the probability measure associated with an observable quantity belong to a family of probability measures $\mathbb{P}_{\Theta \times \{\mathbb{T}\} \times \Xi}^n := (\mathbb{P}_{\theta|\mathbb{T}|\xi}^n)_{\theta \in \Theta, \xi \in \Xi}$.

§10100.09 **Definition.** Given an estimator $\tilde{\theta}$ of a function of interest θ belonging to a class of solutions Θ based on an observable quantity with probability measure $\mathbb{P}_{g,\xi}^n \in \mathbb{P}_{\Theta \times \{T\} \times \Xi}^n$ we call

$$\mathcal{R}_n[\tilde{\theta} | \Theta, \{T\}, \Xi] := \sup \{ \mathbb{E}_{\theta|T|\xi}^n (\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta, \xi \in \Xi \}$$

its *maximal risk* over $\Theta \times \{T\} \times \Xi$. \square

§10100.10 **diSM with noisy operator (§02104.05 continued).** Given $\ell_2 = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ and $\ell_{\infty} = \mathbb{L}_{\infty}(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ consider a Diagonal inverse sequence model (diSM) with noisy operator as in §02104.05 where $\mathfrak{s} \in \mathcal{S} \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ is *not known* anymore. Here the observable stochastic process $\widehat{\mathfrak{s}} = \mathfrak{s} + k^{-1/2} \dot{\eta}$ and $\widehat{g} = g + n^{-1/2} \dot{\epsilon}$ is a noisy version of $\mathfrak{s} \in \mathcal{S} \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ and $g = \mathfrak{s} \cdot \theta \in \text{dom}(M_{\mathfrak{s}}) \subseteq \ell_2$ with $\theta \in \Theta \subseteq \ell_2$, respectively, where $\dot{\epsilon} \sim \otimes_{j \in \mathbb{N}} \mathbb{P}^{\dot{\epsilon}_j}$ and $\dot{\eta} \sim \otimes_{j \in \mathbb{N}} \mathbb{P}^{\dot{\eta}_j}$ are *independent*. In addition, let $\dot{\epsilon}$ satisfy (SM:ub) in Model §10100.06 for $\sigma \in \Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ and let $\dot{\eta}$ fulfill

(SMnO:ub) for $\xi \in \Xi \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ we have $\mathbb{P}^{\dot{\eta}} \in \mathcal{W}_4(\mathcal{B})$ with $\xi_j^4 = \mathbb{P}(\dot{\eta}_j^4)$ and $0 = \mathbb{P}(\dot{\eta}_j)$, $j \in \mathbb{N}$.

Under (SM:ub) \widehat{g} admits a $\mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n$ -distribution belonging to the family $\mathbb{P}_{\Theta \times \mathcal{S} \times \Sigma}^n := (\mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n)_{\theta \in \Theta, \mathfrak{s} \in \mathcal{S}, \sigma \in \Sigma}$ and under (SMnO:ub) $\widehat{\mathfrak{s}}$ admits a $\mathbb{P}_{\mathfrak{s}|\xi}^k$ -distribution belonging to the family $\mathbb{P}_{\mathcal{S} \times \Xi}^k := (\mathbb{P}_{\mathfrak{s}|\xi}^k)_{\mathfrak{s} \in \mathcal{S}, \xi \in \Xi}$. Consequently, $(\widehat{g}, \widehat{\mathfrak{s}})$ admits a joint $\mathbb{P}_{\theta|\mathfrak{s}|\sigma|\xi}^{n \otimes k} = \mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n \otimes \mathbb{P}_{\mathfrak{s}|\xi}^k$ distribution belonging to the family $\mathbb{P}_{\Theta \times \mathcal{S} \times \Sigma \times \Xi}^{n \otimes k} := (\mathbb{P}_{\theta|\mathfrak{s}|\sigma|\xi}^{n \otimes k})_{\theta \in \Theta, \mathfrak{s} \in \mathcal{S}, \sigma \in \Sigma, \xi \in \Xi}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}^2}, \mathcal{B}^{\otimes \mathbb{N}^2}, \mathbb{P}_{\Theta \times \mathcal{S} \times \Sigma \times \Xi}^{n \otimes k})$ where $\Sigma, \Xi \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$, $\mathcal{S} \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ and $\Theta \subseteq \ell_2$. \square

Finally, given a class of solutions Θ , a class of operators \mathbb{T} , a class of nuisance parameters Ξ and noise levels $n, k \in \mathbb{N}$ let $\mathbb{P}_{\Theta \times \mathbb{T} \times \Xi}^{n,k} := (\mathbb{P}_{\theta|T|\xi}^{n,k})_{\theta \in \Theta, T \in \mathbb{T}, \xi \in \Xi}$ denote a family of joint probability measures.

§10100.11 **Definition.** Given an estimator $\tilde{\theta}$ of a function of interest θ belonging to a class of solutions Θ based on observable quantities with joint probability measure $\mathbb{P}_{\theta|T|\xi}^{n,k} \in \mathbb{P}_{\Theta \times \mathbb{T} \times \Xi}^{n,k}$ we call

$$\mathcal{R}_{n,k}[\tilde{\theta} | \Theta, \mathbb{T}, \Xi] := \sup \{ \mathbb{E}_{\theta|T|\xi}^{n,k} (\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta, T \in \mathbb{T}, \xi \in \Xi \}$$

its *maximal risk* over $\Theta \times \mathbb{T} \times \Xi$. \square

§10100.12 **Remark.** Taking the supremum over the class of operators allows us to quantify the additional complexity due to the estimation of the operator. Moreover, if there exist an estimator $\widehat{\theta}$, a constant $C \in \mathbb{R}_{>0}$ and for each $n, k \in \mathbb{N}$ there is $R_{n,k}^* \in \mathbb{R}_{>0}$ such that

(lower) $R_{n,k}^*$ is a *lower bound* up to the constant C^{-1} of the maximal risk over $\Theta \times \mathbb{T} \times \Xi$, that is

$$\inf_{\tilde{\theta}} \mathcal{R}_{n,k}[\tilde{\theta} | \Theta, \mathbb{T}, \Xi] \geq C^{-1} R_{n,k}^*$$

where the infimum is taken over all possible estimators of θ ;

(upper) $R_{n,k}^*$ is an *upper bound* up to the constant C of the maximal risk over $\Theta \times \mathbb{T} \times \Xi$, that is

$$\mathcal{R}_{n,k}[\widehat{\theta} | \Theta, \mathbb{T}, \Xi] \leq C R_{n,k}^*,$$

then we call $R_{n,k}^*$ *minimax-bound* and the estimator $\widehat{\theta}$ *minimax-optimal* (up to the constant C). As a consequence, up to the constant C^2 the estimator $\widehat{\theta}$ attains the lower maximal risk bound that is, $\mathcal{R}_{n,k}[\widehat{\theta} | \Theta, \mathbb{T}, \Xi] \leq C^2 \inf_{\tilde{\theta}} \mathcal{R}_{n,k}[\tilde{\theta} | \Theta, \mathbb{T}, \Xi]$. Typically, we assume first that the nuisance parameter ξ is known *a priori*, and hence $\mathbb{P}_{\Theta, T, \{\xi\}}^{n,k}$ is a class of probability measures associated with the observable quantities. In this situation, we consider the maximal risk $\{ \mathbb{P}_{\theta|T|\xi}^{n,k} (\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta, T \in \mathbb{T} \}$ and we seek a bound $R_{n,k}^*$ up to a constant which depends possibly on the nuisance parameter ξ . However, if the bound $R_{n,k}^*$ is a valid lower and upper bound up to a constant uniformly for all nuisance parameters $\xi \in \Xi$, then it is, obviously, also a bound of the maximal risk $\mathcal{R}_{n,k}[\widehat{\theta} | \Theta, \mathbb{T}, \Xi]$. \square

Considering the Hilbert space $\ell_2 = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ and a surjective partial isometry $U \in \mathbb{L}(\mathbb{H}, \ell)$, which is *fixed* and presumed to be *known* in advance, we study *statistical inverse problems* where observable quantities admit a probability measure $\mathbb{P}_{\theta|\mathbb{T}|\xi}^{n,k} \in \mathbb{P}_{\Theta \times \mathbb{T} \times \Xi}^{n,k}$ for some class Θ, \mathbb{T} and Ξ of solutions, operators and nuisance parameters, respectively. We consider the following global and local measures of accuracy.

§10100.13 **Notation (Reminder).** For sequences $a_n, b_n \in (\mathbb{K})^{\mathbb{N}}$ taking its values in $\mathbb{K} \in \{\mathbb{R}, \mathbb{R}_{>0}, \mathbb{Z}, \dots\}$ we write $a_n \in (\mathbb{K})_{\nearrow}^{\mathbb{N}}$ and $b_n \in (\mathbb{K})_{\searrow}^{\mathbb{N}}$ if a_n and b_n , respectively, is monotonically *non-decreasing* and *non-increasing*. If in addition $a_n \rightarrow \infty$ and $b_n \rightarrow 0$ as $n \rightarrow \infty$, then we write $a_n \in (\mathbb{K})_{\nearrow\infty}^{\mathbb{N}}$ and $b_n \in (\mathbb{K})_{\searrow 0}^{\mathbb{N}}$ for short. For $w_n \in \ell_{\infty} = \mathbb{L}_{\infty}(\nu_{\mathbb{N}})$ we set $w_{[0]} := 0$, $w_{[j]} = (w_{[j]}) := \|w_n \mathbb{1}_n^j\|_{\ell_{\infty}}\big)_{j \in \mathbb{N}}$, $w_{(0)} := \|w_n\|_{\ell_{\infty}}$, and $w_{(\bullet)} = (w_{(j)}) := \|w_n \mathbb{1}_n^{j+1}\|_{\ell_{\infty}}\big)_{j \in \mathbb{N}}$, where by construction $w_{[j]} \in (\mathbb{R}_{\geq 0})_{\nearrow}^{\mathbb{N}}$ and $w_{(\bullet)} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$. \square

§10100.14 **Assumption (Maximal global \mathbf{v} -risk).** Consider weights $\mathbf{t}, \mathbf{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and $\mathbf{v} \in \mathbb{R}_{>0}^{\mathbb{N}}$ such that $(\mathbf{av})_{\bullet} = \mathbf{a} \cdot \mathbf{v} \in \ell_{\infty}$, and $(\mathbf{av})_{(\bullet)} \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ and there exists $C_{(\mathbf{av})} \in (0, 1]$ such that for all $m \in \mathbb{N}$

$$(\mathbf{av})_{(m-1)}^2 \geq \min \{(\mathbf{av})_j^2 : j \in \llbracket m \rrbracket\} \geq C_{(\mathbf{av})} (\mathbf{av})_{(m-1)}^2$$

or in equal $1 \geq C_{(\mathbf{av})} \|(\mathbf{av})_{\bullet}^{-2} \mathbb{1}_{\bullet}^m\|_{\ell_{\infty}} (\mathbf{av})_{(m-1)}^2$. \square

§10100.15 **Reminder (Maximal global \mathbf{v} -risk).** Under Assumption §10100.14 we introduce $\ell_2^{\mathbf{a}} = \text{dom}(M_{\mathbf{a}^{-2}}) = \ell_2 \mathbf{a}_{\bullet} = \ell_2(\mathbf{a}_{\bullet}^{-2})$ endowed with $\|\cdot\|_{\mathbf{a}^{-1}} := \|\cdot\|_{\ell_2(\mathbf{a}_{\bullet}^{-2})}$ and the ellipsoid $\ell_2^{\mathbf{a},r} := \{h \in \ell_2^{\mathbf{a}} : \|h\|_{\mathbf{a}^{-1}}^2 \leq r^2\} \subseteq \ell_2^{\mathbf{a}}$, where the measures $\nu_{\mathbb{N}}, \mathbf{v}_{\bullet}^2 \nu_{\mathbb{N}}$ and $\mathbf{a}_{\bullet}^{-2} \nu_{\mathbb{N}}$ dominate mutually each other. Under Assumption §10100.14 we consider the following global measure of accuracy. Introduce $\ell_2(\mathbf{v}_{\bullet}^2) = \mathbb{L}_2(\mathbf{v}_{\bullet}^2 \nu_{\mathbb{N}}) = \text{dom}(M_{\mathbf{v}}) = \ell_2 \mathbf{v}_{\bullet}^{-1} \subseteq \ell_2$ and $\|\cdot\|_{\mathbf{v}} = \|M_{\mathbf{v}} \cdot\|_{\ell_2}$, where $\ell_2^{\mathbf{a},r} \subseteq \ell_2(\mathbf{v}_{\bullet}^2)$ (**Property** §04102.11). For $\theta = U\theta \in \ell_2^{\mathbf{a},r}$ we call $\mathfrak{d}_{\text{ist}}(\tilde{\theta}, \theta) = \|\tilde{\theta} - \theta\|_{\mathbf{v}}$ *global \mathbf{v} -error*, $\mathbb{P}_{\theta|\mathbb{T}|\xi}^{n,k}(\|\tilde{\theta} - \theta\|_{\mathbf{v}}^2)$ *global \mathbf{v} -risk* and

$$\mathcal{R}_{n,k}^{\mathbf{v}}[\tilde{\theta} | \ell_2^{\mathbf{a},r}, \mathbb{T}, \Xi] := \sup \{ \mathbb{P}_{\theta|\mathbb{T}|\xi}^{n,k}(\|\tilde{\theta} - \theta\|_{\mathbf{v}}^2) : \theta \in \mathbb{J}^{\mathbf{a},r}, \mathbb{T} \in \mathbb{T}, \xi \in \Xi \}$$

maximal \mathbf{v} -risk over $\ell_2^{\mathbf{a},r} \times \mathbb{T} \times \Xi$. Note that $(\mathbf{av})_{(\bullet)}^2 \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ by definition, hence $(\mathbf{av})_{(\bullet)}^2 \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ is satisfied if and only if $(\mathbf{av})_{(m)}^2 = o(1)$ as $m \rightarrow \infty$ (i.e. the maximal global approximation is consistent). Moreover if $(\mathbf{av})_{(\bullet)}^2 \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ then we have trivially $(\mathbf{av})_{(\bullet)}^2 \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$ and $\|(\mathbf{av})_{\bullet}^{-2} \mathbb{1}_{\bullet}^m\|_{\ell_{\infty}} = (\mathbf{av})_m^{-2} = (\mathbf{av})_{(m-1)}^{-2}$ for all $m \in \mathbb{N}$, i.e. Assumption §10100.14 is satisfied with $C_{(\mathbf{av})} = 1$. \square

§10100.16 **Assumption (Maximal local ϕ -risk).** Consider weights $\mathbf{t}, \mathbf{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and $\phi \in \mathbb{R}_{>0}^{\mathbb{N}}$ such that $(\mathbf{a}\phi)_{\bullet} := \mathbf{a} \cdot \phi \in \ell_2$ and $(\mathbf{at})_{\bullet} := \mathbf{a} \cdot \mathbf{t} \in (\mathbb{R}_{>0})_{\searrow 0}^{\mathbb{N}}$. \square

§10100.17 **Reminder (Maximal local ϕ -risk).** Under Assumption §10100.16 introduce $\ell_2^{\mathbf{a}} = \ell_2(\mathbf{a}_{\bullet}^{-2})$ endowed with $\|\cdot\|_{\mathbf{a}^{-1}} := \|\cdot\|_{\ell_2(\mathbf{a}_{\bullet}^{-2})}$ and the ellipsoid $\ell_2^{\mathbf{a},r} := \{h \in \ell_2^{\mathbf{a}} : \|h\|_{\mathbf{a}^{-1}}^2 \leq r^2\} \subseteq \ell_2^{\mathbf{a}}$, where the measures $\nu_{\mathbb{N}}, |\phi| \nu_{\mathbb{N}}$ and $\mathbf{a}_{\bullet}^{-2} \nu_{\mathbb{N}}$ dominate mutually each other. Under Assumption §10100.16 we consider the following local measure of accuracy. Under Assumption §10100.16 introduce $\text{dom}(\phi \nu_{\mathbb{N}}) := \{h \in \ell_2 : \phi h \in \ell_1 = \mathbb{L}_1(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})\}$ and the linear functional $\phi \nu_{\mathbb{N}} : \ell_2 \supseteq \text{dom}(\phi \nu_{\mathbb{N}}) \rightarrow \mathbb{R}$ with $h \mapsto \phi \nu_{\mathbb{N}}(h) := \nu_{\mathbb{N}}(\phi h)$ where $\ell_2^{\mathbf{a},r} \subseteq \text{dom}(\phi \nu_{\mathbb{N}})$ (**Property** §04102.23). For $\theta \in \ell_2^{\mathbf{a},r}$ we call $\mathfrak{d}_{\text{ist}}(\tilde{\theta}, \theta) = |\phi \nu_{\mathbb{N}}(\tilde{\theta} - \theta)|$ *local ϕ -error*, $\mathbb{P}_{\theta|\mathbb{T}|\xi}^{n,k}(|\phi \nu_{\mathbb{N}}(\tilde{\theta} - \theta)|^2)$ *local ϕ -risk* and

$$\mathcal{R}_{n,k}^{\phi}[\tilde{\theta} | \ell_2^{\mathbf{a},r}, \mathbb{T}, \Xi] := \sup \{ \mathbb{P}_{\theta|\mathbb{T}|\xi}^{n,k}(|\phi \nu_{\mathbb{N}}(\tilde{\theta} - \theta)|^2) : \theta \in \ell_2^{\mathbf{a},r}, \mathbb{T} \in \mathbb{T}, \xi \in \Xi \}$$

maximal ϕ -risk over $\ell_2^{\mathbf{a},r} \times \mathbb{T} \times \Xi$. Imposing by Assumption §10100.16 $\mathbf{t}, \mathbf{a} \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ and hence $(\mathbf{at})_{\bullet}^2 \in (\mathbb{R}_{>0})_{\searrow}^{\mathbb{N}}$ is rather weak. If in addition $\liminf_{j \rightarrow \infty} (\mathbf{at})_j^2 \geq c \in \mathbb{R}_{>0}$ is satisfied, and hence

$(\mathbf{a}\mathbf{t})_*, \mathbf{a}_*, \mathbf{t}_* \notin (\mathbb{R}_{>0})_{i_0}^{\mathbb{N}}$, then $\mathbf{a}_*^2 \notin (\mathbb{R}_{>0})_{i_0}^{\mathbb{N}}$ and the assumption $(\mathbf{a}\phi)_* \in \ell_2$ implies $\phi_* \in \ell_2$, which together with $\mathbf{t}_*^2 \notin (\mathbb{R}_{>0})_{i_0}^{\mathbb{N}}$ implies $(\phi/\mathbf{t})_* \in \ell_2$, and thus the rate $\mathcal{R}_n^*(\mathbf{a}, \mathbf{t}, \phi)$ is parametric (Illustration §07101.78 or Illustration §08101.72). Since we are interested in the case of a non-parametric rate, the additional assumption $(\mathbf{a}\mathbf{t})_*^2 \in (\mathbb{R}_{>0})_{i_0}^{\mathbb{N}}$ imposes a rather weak condition satisfied also in Illustration §07101.78 or Illustration §08101.72. \square

§10100.18 **Comment.** We formulate the results in terms of $\theta = U\theta \in \mathbb{J}$ rather than directly for $\theta \in \mathbb{H}$. Since U is known, considering the class $\mathbb{H}^{\mathbf{a},\mathbf{r}} := U^* \mathbb{J}^{\mathbf{a},\mathbf{r}} := \{U^*\theta : \theta \in \mathbb{J}^{\mathbf{a},\mathbf{r}}\}$ we obtain immediately also bounds over $\mathbb{H}^{\mathbf{a},\mathbf{r}}$ for the maximal global risk

$$\mathcal{R}_{n,k}^{\mathbf{v}}[\tilde{\theta} | U^* \mathbb{J}^{\mathbf{a},\mathbf{r}}, \mathbb{T}, \Xi] := \sup \{ \mathbb{E}_{\theta|\mathbb{T}|\xi}^{n,k} (\|U(\tilde{\theta} - \theta)\|_{\mathbf{v}}^2) : \theta \in \mathbb{H}^{\mathbf{a},\mathbf{r}}, \mathbb{T} \in \mathbb{T}, \xi \in \Xi \}$$

and maximal local risk

$$\mathcal{R}_{n,k}^{\phi}[\tilde{\theta} | U^* \mathbb{J}^{\mathbf{a},\mathbf{r}}, \mathbb{T}, \Xi] := \sup \{ \mathbb{E}_{\theta|\mathbb{T}|\xi}^{n,k} (|\phi \nu_{\mathbb{N}}(U(\tilde{\theta} - \theta))|^2) : \theta \in \mathbb{H}^{\mathbf{a},\mathbf{r}}, \mathbb{T} \in \mathbb{T}, \xi \in \Xi \}$$

which we do not explicitly state in the sequel. \square

§11 Deriving a lower bound: a general reduction scheme

For a detailed discussion of several other strategies to derive lower bounds we refer the reader, for example, to the text book by Tsybakov [2009].

§11100.01 **Definition.** Let \mathbb{P}_0 and \mathbb{P}_1 be two probability measures on a measurable space $(\mathcal{X}, \mathcal{X})$.

(a) The function

$$\text{KL}(\mathbb{P}_0|\mathbb{P}_1) = \begin{cases} \mathbb{E}_0 \left(\log \frac{d\mathbb{P}_0}{d\mathbb{P}_1} \right) = \int \log \left(\frac{d\mathbb{P}_0}{d\mathbb{P}_1} \right) d\mathbb{P}_0, & \text{if } \mathbb{P}_0 \ll \mathbb{P}_1, \\ +\infty, & \text{otherwise} \end{cases}$$

is called *Kullback-Leibler-divergence* of \mathbb{P}_0 with respect to \mathbb{P}_1 .

Let $\mu \in \mathcal{M}_{\sigma}(\mathcal{X})$ be a \mathbb{P}_0 and \mathbb{P}_1 dominating σ -finite measure (e.g. $\mathbb{P}_0, \mathbb{P}_1 \ll \mu = \mathbb{P}_0 + \mathbb{P}_1$). We write $d\mathbb{P}_0 := d\mathbb{P}_0/d\mu$ and $d\mathbb{P}_1 := d\mathbb{P}_1/d\mu$ for short.

(b) The *Hellinger distance* between \mathbb{P}_0 and \mathbb{P}_1 is defined by

$$H(\mathbb{P}_0, \mathbb{P}_1) := \left(\int |\sqrt{d\mathbb{P}_0} - \sqrt{d\mathbb{P}_1}|^2 \right)^{1/2} := \|\sqrt{d\mathbb{P}_0} - \sqrt{d\mathbb{P}_1}\|_{\mathbb{L}_2(\mu)}$$

(c) and the *Hellinger affinity* is given by

$$\rho(\mathbb{P}_0, \mathbb{P}_1) := \int \sqrt{d\mathbb{P}_0} \sqrt{d\mathbb{P}_1} := \langle \sqrt{d\mathbb{P}_0}, \sqrt{d\mathbb{P}_1} \rangle_{\mathbb{L}_2(\mu)},$$

where both do not depend on the choice of the dominating measure μ . \square

§11100.02 **Remark.** The Kullback-Leibler-divergence satisfies $\text{KL}(\mathbb{P}_0|\mathbb{P}_1) \geq 0$ as well as $\text{KL}(\mathbb{P}_0|\mathbb{P}_1) = 0$ if and only if $\mathbb{P}_0 = \mathbb{P}_1$, but $\text{KL}(\cdot|\cdot)$ is not symmetric. Moreover, for product measures holds $\text{KL}(\mathbb{P}_{0,1} \otimes \mathbb{P}_{0,2}|\mathbb{P}_{1,1} \otimes \mathbb{P}_{1,2}) = \text{KL}(\mathbb{P}_{0,1}|\mathbb{P}_{1,1}) + \text{KL}(\mathbb{P}_{0,2}|\mathbb{P}_{1,2})$ and $\rho(\mathbb{P}_{0,1} \otimes \mathbb{P}_{0,2}, \mathbb{P}_{1,1} \otimes \mathbb{P}_{1,2}) = \rho(\mathbb{P}_{0,1}, \mathbb{P}_{1,1})\rho(\mathbb{P}_{0,2}, \mathbb{P}_{1,2})$. \square

§11100.03 **Lemma.** (i) $0 \leq H^2(\mathbb{P}_0, \mathbb{P}_1) \leq 2$; (ii) $\rho(\mathbb{P}_0, \mathbb{P}_1) = 1 - \frac{1}{2}H^2(\mathbb{P}_0, \mathbb{P}_1)$; and (iii) $H^2(\mathbb{P}_0, \mathbb{P}_1) \leq \text{KL}(\mathbb{P}_0|\mathbb{P}_1)$.

§11100.04 **Proof of Lemma §11100.03.** Given in the lecture course *Statistik 2* (Lemma §13.12, p.54). \square

§1100.05 **Lemma.** For $a, b \in \ell_\infty$ and $n \in \mathbb{N}$ we have $\text{KL}(\mathbb{N}_a^n | \mathbb{N}_b^n) = \frac{n}{2} \|a - b\|_{\ell_2}^2$.

§1100.06 **Proof of Lemma §1100.05.** Given in the lecture course [Statistik 2](#) (Lemma §13.14, p.54). \square

§1100.07 **Assumption.** The distribution $\mathbb{P} \in \mathcal{W}(\mathcal{B})$ admits a Lebesgue-density $\mathbb{p} := d\mathbb{P}/d\lambda$ and there exist constants $C_\circ, x_\circ \in \mathbb{R}_{>0}$ such that

$$\forall x \in [-x_\circ, x_\circ] : \int \mathbb{p}(u) \log \left(\frac{\mathbb{p}(u)}{\mathbb{p}(u-x)} \right) \lambda(du) \leq C_\circ x^2.$$

\square

§1100.08 **Lemma.** Let $Y \sim \mathbb{P}^{\otimes \mathbb{N}}$ where $\mathbb{P} \in \mathcal{W}(\mathcal{B})$ fulfills Assumption §1100.07 with $C_\circ, x_\circ \in \mathbb{R}_{>0}$. For $a, b, \sigma \in \ell_\infty$ and $n \in \mathbb{N}$ consider $a + n^{-1/2} \sigma Y \sim \mathbb{P}_{a|\sigma}^n$ and $b + n^{-1/2} \sigma Y \sim \mathbb{P}_{b|\sigma}^n$. If $\|\sigma^{-2}\|_{\ell_\infty} =: \nu_\sigma \in \mathbb{R}_{>0}$ and $n^{1/2} \nu_\sigma^{1/2} \|a - b\|_{\ell_\infty} \leq x_\circ$ then we have $\text{KL}(\mathbb{P}_{a|\sigma}^n | \mathbb{P}_{b|\sigma}^n) \leq n \nu_\sigma C_\circ \|a - b\|_{\ell_2}^2$.

§1100.09 **Proof of Lemma §1100.08.** Given in the lecture. \square

§1100.10 **Comment.** For $\sigma \in \mathbb{R}_{>0}$ the normal distribution $\mathbb{N}_{(0, \sigma^2)} \in \mathcal{W}(\mathcal{B})$ satisfy Assumption §1100.07 with $C_\circ = 1/(2\sigma^2)$ and $x_\circ = \infty$ (see [Proof §1100.06](#)). \square

§1100.11 **Assumption.** The semi metric $\mathfrak{d}_{\text{ist}}(\cdot, \cdot)$ is *symmetric* and satisfies the *triangular inequality*. Moreover, for any estimator $\tilde{\theta}$ and parameter θ_0 and θ_1 such that $\mathfrak{d}_{\text{ist}}(\theta_0, \theta_1) \in \mathbb{R}_{>0}$ the quantities $\mathfrak{d}_{\text{ist}}(\tilde{\theta}, \theta_0)$ and $\mathfrak{d}_{\text{ist}}(\tilde{\theta}, \theta_1)$ are *measurable*. \square

§1100.12 **Lemma.** Let $(\mathcal{X}, \mathcal{X})$ be a measurable space, let θ_0 and θ_1 be parameters with $\mathfrak{d}_{\text{ist}}(\theta^0, \theta^1) \in \mathbb{R}_{>0}$, and let Assumption §1100.11 be satisfied.

(i) If $\mathbb{P}_0, \mathbb{P}_1 \in \mathcal{W}(\mathcal{X})$ are probability measures then for any estimator $\tilde{\theta}$ we have

$$\mathbb{P}_0(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta^0)) + \mathbb{P}_1(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta^1)) \geq \frac{1}{2} \mathfrak{d}_{\text{ist}}^2(\theta^0, \theta^1) \rho^2(\mathbb{P}_0, \mathbb{P}_1). \quad (11.01)$$

(ii) If $\mathbb{P}_0, \mathbb{P}_1 \in \mathcal{W}(\mathcal{X})$ satisfy $H(\mathbb{P}_0, \mathbb{P}_1) \leq 1$, then for any estimator $\tilde{\theta}$ we have

$$\mathbb{P}_0(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta^0)) + \mathbb{P}_1(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta^1)) \geq \frac{1}{8} \mathfrak{d}_{\text{ist}}^2(\theta^0, \theta^1). \quad (11.02)$$

(iii) For $n \in \mathbb{N}_{\geq 2}$ let $\mathbb{P}_0^n := \otimes_{j \in \llbracket n \rrbracket} \mathbb{P}_{0j} \in \mathcal{W}(\mathcal{X}^{\otimes n})$ and $\mathbb{P}_1^n := \otimes_{j \in \llbracket n \rrbracket} \mathbb{P}_{1j} \in \mathcal{W}(\mathcal{X}^{\otimes n})$ be product probability measures with marginals $\mathbb{P}_{0j}, \mathbb{P}_{1j} \in \mathcal{W}(\mathcal{X})$ fulfilling $H(\mathbb{P}_{0j}, \mathbb{P}_{1j}) \leq 2n^{-1}$ for each $j \in \llbracket n \rrbracket$. Then for any estimator $\tilde{\theta}$ we have

$$\mathbb{P}_0^n(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta^0)) + \mathbb{P}_1^n(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta^1)) \geq \frac{1}{32} \mathfrak{d}_{\text{ist}}^2(\theta^0, \theta^1). \quad (11.03)$$

§1100.13 **Proof of Lemma §1100.12.** Given in the lecture. \square

§11|01 Lower bound based on two hypothesis

§1101.01 **Lemma (Lower bound based on two hypothesis).** Given a noise level $n \in \mathbb{N}$ let $\mathbb{P}_\theta^n := (\mathbb{P}_\theta^n)_{\theta \in \Theta}$ be a family of probability measures. We measure the accuracy of an estimator $\tilde{\theta}$ by its maximal risk

$$\mathcal{R}_n[\tilde{\theta} | \Theta] := \sup \{ \mathbb{E}_\theta^n(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta \}.$$

- (i) If there are $\theta^0, \theta^1 \in \Theta$ with $\mathfrak{d}_{\text{ist}}(\theta^0, \theta^1) \in \mathbb{R}_{>0}$ and associated probability measures $\mathbb{P}_{\theta^0}^n$ and $\mathbb{P}_{\theta^1}^n$ such that Assumption §1100.11 and $H(\mathbb{P}_{\theta^0}^n, \mathbb{P}_{\theta^1}^n) \leq 1$ are satisfied then we have

$$\inf_{\tilde{\theta}} \mathcal{R}_n[\tilde{\theta} | \Theta] \geq \frac{1}{16} \mathfrak{d}_{\text{ist}}^2(\theta^0, \theta^1) \quad (11.04)$$

where the infimum is taken over all possible estimators.

- (ii) Let $n \in \mathbb{N}_{\geq 2}$ and for each $\theta \in \Theta$ let $\mathbb{P}_{\theta}^n = \otimes_{j \in [n]} \mathbb{P}_{\theta}^1$ be a product probability measure with identically \mathbb{P}_{θ}^1 -distributed marginals. If there are $\theta^0, \theta^1 \in \Theta$ with $\mathfrak{d}_{\text{ist}}(\theta^0, \theta^1) \in \mathbb{R}_{>0}$ and associated marginal probability measures $\mathbb{P}_{\theta^0}^1$ and $\mathbb{P}_{\theta^1}^1$ such that Assumption §1100.11 and $H(\mathbb{P}_{\theta^0}^1, \mathbb{P}_{\theta^1}^1) \leq 2n^{-1}$ are satisfied then we have

$$\inf_{\tilde{\theta}} \mathcal{R}_n[\tilde{\theta} | \Theta] \geq \frac{1}{64} \mathfrak{d}_{\text{ist}}^2(\theta^0, \theta^1) \quad (11.05)$$

where the infimum is taken over all possible estimators.

§1101.02 **Proof of Lemma §1101.01.** Given in the lecture. □

§1101.03 **Remark (Lower bound for a local ϕ -risk).** Assuming the bounded Hellinger distance as for example in Lemma §1101.01, Le Cam's general method (see Le Cam [1973]) and Pinsker's inequality allow us to derive a lower bound for a local ϕ -risk as in Reminder §1000.17. However, in this special setting a lower bound can be obtained elementarily from Lemma §1101.01, which in case (i) for any estimator $\tilde{\theta}$ states

$$\mathcal{R}_n^{\phi}[\tilde{\theta} | \Theta] := \sup \{ \mathbb{E}_{\theta^0}^n (|\phi_{\mathcal{N}}(\tilde{\theta} - \theta)|^2) : \theta \in \Theta \} \geq \frac{1}{16} |\phi_{\mathcal{N}}(\theta^0 - \theta^1)|^2.$$

If we consider furthermore candidates $\theta^0 := \theta^*$ and $\theta^1 := -\theta^*$ for some $\theta^* \in \Theta$ such that $-\theta^* \in \Theta$, then trivially $|\phi_{\mathcal{N}}(\theta^0 - \theta^1)|^2 = 4|\phi_{\mathcal{N}}(\theta^*)|^2$ which in turn under the conditions of Lemma §1101.01 (i) implies

$$\inf_{\tilde{\theta}} \mathcal{R}_n^{\phi}[\tilde{\theta} | \Theta] \geq \frac{1}{4} |\phi_{\mathcal{N}}(\theta^*)|^2. \quad (11.06)$$

Similarly, under the conditions of Lemma §1101.01 (ii) we get

$$\inf_{\tilde{\theta}} \mathcal{R}_n^{\phi}[\tilde{\theta} | \Theta] \geq \frac{1}{16} |\phi_{\mathcal{N}}(\theta^*)|^2. \quad (11.07)$$

Often a minimax-optimal lower bound can be found by constructing a candidate $\theta^* = U\theta^* \in \Theta$ that has the largest possible $|\phi_{\mathcal{N}}(\theta^*)|^2$ -value but $\mathbb{P}_{\theta^*}^n$ and $\mathbb{P}_{-\theta^*}^n$ are still statistically indistinguishable in the sense that $H(\mathbb{P}_{\theta^*}^n, \mathbb{P}_{-\theta^*}^n) \leq 1$ or $H(\mathbb{P}_{\theta^*}^1, \mathbb{P}_{-\theta^*}^1) \leq 2n^{-1}$. □

§1101.04 **Reminder (Maximal local ϕ -risk in diSM (§1000.08 continued)).** In Subsection §0701 we consider an orthogonal projection estimator (OPE) in a Diagonal inverse sequence model (diSM) as in Model §0104.08 (summarised in Model §1000.08). Here the observable noisy version \hat{g} satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, \mathbb{P}_{\Theta \times \{\mathfrak{s}\} \times \Sigma}^n := (\mathbb{P}_{|\mathfrak{s}| \sigma}^n)_{\theta \in \Theta, \alpha \in \Sigma})$ where $\mathfrak{s} \in \ell_{\infty}$ is known, $\Theta \subseteq \ell_2$ and $\Sigma \subseteq \mathbb{R}_{>0} \cap \ell_{\infty}$. Under Assumption §0701.64 (which is implied by Assumption §1000.16) in Corollary §0701.74 an upper bound for the maximal local ϕ -risk of an OPE is shown. More precisely, assuming a multiplication operator $M_{\mathfrak{s}} \in \mathbb{L}(\mathbb{J})$ (compare Notation §0104.01), which fulfills a link condition $M_{\mathfrak{s}} \in \mathbb{M}_{t,d}$ for $d \in \mathbb{R}_{\geq 1}$ (see Assumption §0403.04), the performance of the OPE $\hat{\theta}^m = \mathfrak{s}^{\dagger} \hat{g} \mathbf{1}^m \in \text{dom}(\phi_{\mathcal{N}})$ with dimension $m \in \mathbb{N}$ is measured by its maximal local ϕ -risk over the ellipsoid $\Theta = \ell_2^{\alpha,r}$ with $r \in \mathbb{R}_{>0}$, that is

$$\mathcal{R}_n^{\phi}[\hat{\theta}^m | \ell_2^{\alpha,r}, \{M_{\mathfrak{s}}\}, \{\sigma\}] := \sup \{ \mathbb{E}_{|\mathfrak{s}| \sigma}^n (|\phi_{\mathcal{N}}(\hat{\theta}^m - \theta)|^2) : \theta \in \ell_2^{\alpha,r} \} \quad \forall n, m \in \mathbb{N}.$$

For $n, m \in \mathbb{N}$ setting (as in (07.20))

$$\begin{aligned} R_n^m(\mathbf{a}, \mathbf{t}, \phi) &:= \|\mathbf{a} \mathbf{1}^{m \perp}\|_\phi^2 + n^{-1} \|\mathbf{t}^\dagger \mathbf{1}^m\|_\phi^2, \quad m_n^* := \arg \min \{R_n^m(\mathbf{a}, \mathbf{t}, \phi) : m \in \mathbb{N}\} \\ \text{and } R_n^*(\mathbf{a}, \mathbf{t}, \phi) &:= R_n^{m_n^*}(\mathbf{a}, \mathbf{t}, \phi) = \min \{R_n^m(\mathbf{a}, \mathbf{t}, \phi) : m \in \mathbb{N}\} \end{aligned} \quad (11.08)$$

the OPE $\widehat{\theta}^{m_n^*} = \mathfrak{s}^\dagger \widehat{g} \mathbf{1}^{m_n^*} \in \text{dom}(\phi \nu_n)$ with optimally chosen dimension $m_n^* = m_n^*(\mathbf{a}, \mathbf{t}, \phi)$ as in (11.08) fulfills

$$\mathcal{R}_n^\phi[\widehat{\theta}^{m_n^*} | \ell_2^{\mathbf{a}, \mathbf{r}}, \{M_s\}, \{\sigma\}] \leq (\mathbf{v}_\sigma \mathbf{d}^2 \vee \mathbf{r}^2) R_n^*(\mathbf{a}, \mathbf{t}, \phi) \quad \forall n \in \mathbb{N} \quad (11.09)$$

with $\|\sigma^2\|_{\ell_\infty} =: \mathbf{v}_\sigma \in \mathbb{R}_{>0}$. In the proof of the next proposition we make use of [Lemma §08101.61](#) which under [Assumption §07101.64](#) (implied by [Assumption §11100.07](#)) states that $(\mathbf{a}\mathbf{t})_{m_n^*}^2 > n^{-1} \geq (\mathbf{a}\mathbf{t})_{m_n^*+1}^2 = (\mathbf{a}\mathbf{t})_{(m_n^*)}^2$ for all $n \in \mathbb{N}$ with $(\mathbf{a}\mathbf{t})_2^2 > n^{-1}$, i.e. $n \in \mathbb{N}_{>(\mathbf{a}\mathbf{t})_2^2}$. \square

§11101.05 **diSM (§10100.08 continued)**. Consider $\widehat{g} = g + n^{-1/2} \dot{\epsilon} \sim P_{\theta|\mathfrak{s}\sigma}^n$ as in [Model §10100.08](#), where $\dot{\epsilon}$ satisfies [\(SM:ub\)](#) with $P_{(0,1)} \in \mathcal{W}(\mathcal{B})$ and $\sigma \in \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$. In addition [\(SM:lb\)](#) $P_{(0,1)} \in \mathcal{W}(\mathcal{B})$ fulfills [Assumption §11100.07](#) with $C_\epsilon, x_\epsilon \in \mathbb{R}_{>0}$ and $\sigma^{-2} \in \ell_\infty$. \square

§11101.06 **Corollary** (diSM §11101.05 continued). For $\mathfrak{s} \in \ell_\infty, \theta \in \ell_2$, hence $g = \mathfrak{s}\theta \in \text{dom}(M_s) \subseteq \ell_2$, consider $\widehat{g} = g + n^{-1/2} \dot{\epsilon} \sim P_{\theta|\mathfrak{s}\sigma}^n$ as in [Model §11101.05](#), where $\dot{\epsilon}$ fulfills [\(SM:ub\)](#) and [\(SM:lb\)](#) with $C_\epsilon, x_\epsilon \in \mathbb{R}_{>0}$ $\|\sigma^{-2}\|_{\ell_\infty} =: \mathbf{v}_\sigma \in \mathbb{R}_{>0}$. For each $\theta^* \in \ell_2$ with $2n^{1/2} \mathbf{v}_\sigma^{1/2} \|\mathfrak{s}\theta^*\|_{\ell_\infty} \leq x_\epsilon$ setting $\theta^0 := \theta^*$ and $\theta^1 := -\theta^*$ the distributions $P_{\theta^0|\mathfrak{s}\sigma}^n \in \mathcal{W}(\mathcal{B}^{\otimes \mathbb{N}})$, $\tau \in \{0, 1\}$, satisfy $H^2(P_{\theta^0|\mathfrak{s}\sigma}^n, P_{\theta^1|\mathfrak{s}\sigma}^n) \leq 4n \mathbf{v}_\sigma C_\epsilon \|\mathfrak{s}\theta^*\|_{\ell_2}^2$.

§11101.07 **Proof of Corollary §11101.06**. Given in the lecture. \square

§11101.08 **Proposition** (diSM §11101.05 continued). For $\mathfrak{s} \in \ell_\infty, \theta \in \ell_2$, hence $g = \mathfrak{s}\theta \in \text{dom}(M_s) \subseteq \ell_2$, consider $\widehat{g} = g + n^{-1/2} \dot{\epsilon} \sim P_{\theta|\mathfrak{s}\sigma}^n$ as in [Model §11101.05](#), where $\dot{\epsilon}$ fulfills [\(SM:ub\)](#) and [\(SM:lb\)](#) with $C_\epsilon, x_\epsilon \in \mathbb{R}_{>0}$ and $\|\sigma^{-2}\|_{\ell_\infty} =: \mathbf{v}_\sigma \in \mathbb{R}_{>0}$. Let [Assumption §10100.16](#) and in addition

$$|\mathfrak{s}| \leq \mathbf{d}\mathbf{t} \quad \nu_{\mathbb{N}}\text{-a.e.} \quad \text{for } \mathbf{d} \in \mathbb{R}_{\geq 1} \quad (11.10)$$

be satisfied. Then we have

$$\inf_{\tilde{g}} \mathcal{R}_n^\phi[\tilde{g} | \ell_2^{\mathbf{a}, \mathbf{r}}, \{M_s\}, \{\sigma\}] \geq R_n^*(\mathbf{a}, \mathbf{t}, \phi) \times \frac{1}{16} (4\mathbf{r}^2 \wedge \mathbf{v}_\sigma^{-1} \mathbf{d}^{-2} (C_\epsilon^{-1} \wedge x_\epsilon^2)) \quad \forall n \in \mathbb{N}_{>(\mathbf{a}\mathbf{t})_2^2} \quad (11.11)$$

where the infimum is taken over all possible estimators.

§11101.09 **Proof of Proposition §11101.08**. Given in the lecture. \square

§11101.10 **Comment**. By combining the lower bound in [Proposition §11101.08](#) and the upper bound in [Corollary §07101.74](#) for the maximal local ϕ -risk of an OPE in a diSM §11101.05 we have shown that $R_n^*(\mathbf{a}, \mathbf{t}, \phi)$ is a minimax-rate and the OPE with optimally chosen dimension parameter is minimax-optimal (up to a constant). \square

§11101.11 **GdiSM (§01104.09 continued)**. Consider a Gaussian diagonal inverse sequence model (GdiSM) as in §01104.09 where $\mathfrak{s} \in \ell_\infty$ is known in advance. Here the observable stochastic process $\widehat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta|\mathfrak{s}}^n$ is a noisy version of $g = \mathfrak{s}\theta \in \ell_2$ with $\theta = \mathfrak{s}^\dagger g \in \Theta \subseteq \ell_2$ and $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$. Consequently, \widehat{g} admits a $N_{\theta|\mathfrak{s}}^n$ -distribution belonging to the family $N_{\Theta \times \{\mathfrak{s}\}}^n := (N_{\theta|\mathfrak{s}}^n)_{\theta \in \Theta}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, N_{\Theta \times \{\mathfrak{s}\}}^n)$ where $\Theta \subseteq \ell_2$. Under [Assumption §07101.64](#) (which is implied by [Assumption §10100.16](#)) in [Corollary §07101.72](#) an upper bound for the maximal local ϕ -risk of an OPE is shown. More precisely,

assuming a multiplication operator $M_s \in \mathbb{L}(\mathbb{J})$ (compare **Notation** §01104.01), which fulfills a link condition $M_s \in \mathbb{M}_{t,d}$ for $d \in \mathbb{R}_{\geq 1}$ (see Assumption §04103.04), the performance of the OPE $\hat{\theta}^m = \mathfrak{s}^\dagger \hat{g} \mathbb{1}^m \in \text{dom}(\phi_{\mathbb{N}})$ with dimension $m \in \mathbb{N}$ is measured by its maximal local ϕ -risk over the ellipsoid $\Theta = \ell_2^{\alpha,r}$ with $r \in \mathbb{R}_{>0}$, that is

$$\mathcal{R}_n^\phi[\hat{\theta}^m | \ell_2^{\alpha,r}, \{M_s\}] := \sup \{ \mathbb{N}_{\theta|s}^n (|\phi_{\mathbb{N}}(\hat{\theta}^{m^*} - \theta)|^2) : \theta \in \ell_2^{\alpha,r} \} \quad \forall n, m \in \mathbb{N}.$$

The OPE $\hat{\theta}^{m^*} = \mathfrak{s}^\dagger \hat{g} \mathbb{1}^{m^*} \in \text{dom}(\phi_{\mathbb{N}})$ with optimally chosen dimension $m^* = m_n^*(\alpha, t, \phi)$ as in (11.08) fulfills $\mathcal{R}_n^\phi[\hat{\theta}^{m^*} | \ell_2^{\alpha,r}, \{M_s\}] \leq (d^2 \vee r^2) R_n^*(\alpha, t, \phi)$ for all $n \in \mathbb{N}$. \square

§11101.12 **Corollary** (GdiSM §11101.11 continued). For $\mathfrak{s} \in \ell_\infty$, $\theta_s \in \ell_2$, hence $g = \mathfrak{s} \cdot \theta_s \in \text{dom}(M_s) \subseteq \ell_2$, consider $\hat{g} = g + n^{-1/2} \dot{B} \sim \mathbb{N}_{\theta|s}^n$ as in Model §11101.11, where $\dot{B} \sim \mathbb{N}_{(0,1)}^{\otimes \mathbb{N}}$. Let Assumption §10100.16 and in addition (11.10) be satisfied. Then we have

$$\inf_{\hat{\theta}} \mathcal{R}_n^\phi[\hat{\theta}^m | \ell_2^{\alpha,r}, \{M_s\}] \geq R_n^*(\alpha, t, \phi) \times \frac{1}{8} (2r^2 \wedge d^{-2}) \quad \forall n \in \mathbb{N}_{>(at)^2} \quad (11.12)$$

where the infimum is taken over all possible estimators.

§11101.13 **Proof of Corollary** §11101.12. Given in the lecture. \square

§11101.14 **Comment**. By combining the lower bound in **Corollary** §11101.12 and the upper bound in **Corollary** §07101.72 for the maximal local ϕ -risk of an OPE in a GdiSM §11101.11 we have shown that $R_n^*(\alpha, t, \phi)$ is a minimax-rate and the OPE with optimally chosen dimension parameter is minimax-optimal (up to a constant). \square

§11101.15 **Remark**. Let $\mathbb{P}_{\Theta \times \Xi}^{n \otimes k} = (\mathbb{P}_{\theta|\xi}^{n \otimes k})_{\theta \in \Theta, \xi \in \Xi}$ be a family of product measures $\mathbb{P}_{\theta|\xi}^{n \otimes k} = \mathbb{P}_{\theta|\xi}^n \otimes \mathbb{P}_\xi^k$ depending on a function of interest $\theta \in \Theta$, a nuisance parameter $\xi \in \Xi$ and noise levels $n, k \in \mathbb{N}$. The **Lemma** §11101.01 allows us to bound from below the maximal risk for each nuisance parameter $\xi \in \Xi$ and noise level $n \in \mathbb{N}$. To be more precise, given a noise level $n \in \mathbb{N}$ for $\tau \in \{0, 1\}$ consider $\theta^\tau \in \Theta$ with associated product probability measure $\mathbb{P}_{\theta^\tau|\xi}^{n \otimes k} = \mathbb{P}_{\theta^\tau|\xi}^n \otimes \mathbb{P}_\xi^k$, then we have $\rho(\mathbb{P}_{\theta^\tau|\xi}^{n \otimes k}, \mathbb{P}_{\theta^1|\xi}^{n \otimes k}) = \rho(\mathbb{P}_{\theta^\tau|\xi}^n \otimes \mathbb{P}_\xi^k, \mathbb{P}_{\theta^1|\xi}^n \otimes \mathbb{P}_\xi^k) = \rho(\mathbb{P}_{\theta^\tau|\xi}^n, \mathbb{P}_{\theta^1|\xi}^n)$ due to the independence. Consequently, if $H(\mathbb{P}_{\theta^\tau|\xi}^n, \mathbb{P}_{\theta^1|\xi}^n) \leq 1$, then for any estimator $\tilde{\theta}$ we obtain

$$\mathcal{R}_{n,k}[\tilde{\theta} | \Theta, \{\xi\}] := \sup \{ \mathbb{E}_{\theta|\xi}^{n \otimes k} (\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta \} \geq \frac{1}{16} \mathfrak{d}_{\text{ist}}^2(\theta^0, \theta^1)$$

due to **Lemma** §11101.01. It is worth noting that we obtain the same lower bound when disposing of the family $\mathbb{P}_{\Theta \times \{\xi\}}^n = (\mathbb{P}_{\theta|\xi}^n)_{\theta \in \Theta}$ only, in other words assuming the nuisance parameter $\xi \in \Xi$ is known in advance. \square

§11101.16 **Corollary (Lower bound based on two hypothesis)**. Let $\mathbb{P}_{\Theta \times \Xi}^{n \otimes k} = (\mathbb{P}_{\theta|\xi}^{n \otimes k})_{\theta \in \Theta, \xi \in \Xi}$ be a family of product measures $\mathbb{P}_{\theta|\xi}^{n \otimes k} = \mathbb{P}_{\theta|\xi}^n \otimes \mathbb{P}_\xi^k$ depending on a function of interest $\theta \in \Theta$, a nuisance parameter $\xi \in \Xi$ and noise levels $n, k \in \mathbb{N}$. If for each $\tau \in \{0, 1\}$ there are $\theta^\tau \in \Theta$ and a nuisance parameter $\xi^\tau \in \Xi$ with associated product probability measure $\mathbb{P}_{\theta^\tau|\xi^\tau}^{n \otimes k} = \mathbb{P}_{\theta^\tau|\xi^\tau}^n \otimes \mathbb{P}_{\xi^\tau}^k$ such that $\mathbb{P}_{\theta^0|\xi^0}^n = \mathbb{P}_{\theta^1|\xi^1}^n$ and in addition $H(\mathbb{P}_{\xi^0}^k, \mathbb{P}_{\xi^1}^k) \leq 1$ then for any estimator $\tilde{\theta}$ we have

$$\mathcal{R}_{n,k}[\tilde{\theta} | \Theta, \Xi] := \sup \{ \mathbb{E}_{\theta|\xi}^{n \otimes k} (\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta, \xi \in \Xi \} \geq \frac{1}{16} \mathfrak{d}_{\text{ist}}^2(\theta^0, \theta^1). \quad (11.13)$$

§11101.17 **Proof of Corollary** §11101.16. Given in the lecture. \square

§11101.18 **Remark**. The last assertion allows us often to derive a lower bound depending on the classes Θ and Ξ and the noise level k but not on the noise level n . Roughly speaking this means that we cover the influence of the estimation of the nuisance parameter. Typically we combine this lower bound with the lower bound obtained in **Lemma** §11101.01 where the nuisance parameter is assumed to be known in advance. \square

§1101.19 **Reminder** (Maximal global \mathbf{v} -risk in diSM with noisy operator (§10100.10 continued)). For $\ell_2 = \mathbb{L}_2(\mathbb{N}, 2^{\mathbb{N}}, \nu_{\mathbb{N}})$ in **Subsection** §07102 we consider a thresholded orthogonal projection estimator (tOPE) in a Diagonal inverse sequence model (diSM) with noisy operator as in Model §02104.05 (summarised in Model §10100.10). Here the observable noisy versions $(\widehat{g}, \widehat{\mathfrak{s}})$ satisfy a statistical product experiment

$$(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, \mathbb{P}_{\Theta \times \mathcal{S} \times \Sigma \times \Xi}^{n \otimes k} := (\mathbb{P}_{\theta|\sigma|\xi}^{n \otimes k} := \mathbb{P}_{\theta|\sigma}^n \otimes \mathbb{P}_{\xi}^k)_{\theta \in \Theta, \mathfrak{s} \in \mathcal{S}, \sigma \in \Sigma, \xi \in \Xi})$$

where $\Sigma, \Xi \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$, $\mathcal{S} \subseteq \mathbb{R}_{\setminus 0}^{\mathbb{N}} \cap \ell_{\infty}$ and $\Theta \subseteq \ell_2$. Under Assumption §07102.32 (which is implied by Assumption §10100.14) in **Corollary** §07102.39 an upper bound for the maximal global \mathbf{v} -risk of a tOPE is shown. More precisely, assuming a multiplication operator $M_{\mathfrak{s}} \in \mathbb{L}(\ell_2)$ (compare **Notation** §02104.02), which fulfills a link condition $M_{\mathfrak{s}} \in \mathbb{M}_{t,d}$ for $d \in \mathbb{R}_{\geq 1}$ (see Assumption §04103.04), the performance of the tOPE $\widehat{\theta}^m := \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^m = \widehat{\mathfrak{s}}^{\dagger} \mathbb{1}_{\{\widehat{\mathfrak{s}}^2 \geq k^{-1}\}} \widehat{g}^m \mathbb{1}^m \in \ell_2(\mathbf{v}^2)$ (see **Definition** §07102.04) with dimension $m \in \mathbb{N}$ is measured by its maximal global \mathbf{v} -risk over the ellipsoid $\Theta = \ell_2^{\mathfrak{a},r}$ with $r \in \mathbb{R}_{>0}$ and the link condition $\mathcal{S} = \mathbb{M}_{t,d}$ with $d \in \mathbb{R}_{\geq 1}$, that is

$$\mathcal{R}_{n,k}^{\mathbf{v}}[\widehat{\theta}^m | \ell_2^{\mathfrak{a},r}, \mathbb{M}_{t,d}, \{\sigma\}, \{\xi\}] := \sup \{ \mathbb{P}_{\theta|\sigma|\xi}^{n \otimes k}(\|\widehat{\theta}^m - \theta\|_{\mathbf{v}}^2) : \theta \in \ell_2^{\mathfrak{a},r}, M_{\mathfrak{s}} \in \mathbb{M}_{t,d} \} \quad \forall n, k, m \in \mathbb{N}.$$

For $n, m \in \mathbb{N}$ setting (as in (07.37))

$$\begin{aligned} \mathbb{R}_n^m(\mathfrak{a}, \mathfrak{t}, \mathbf{v}) &:= [(\mathfrak{a}\mathbf{v})_{(m)}^2 \vee n^{-1} \|\mathfrak{t}^{\dagger} \mathbb{1}^m\|_{\mathbf{v}}^2], \quad m_n^* := \arg \min \{ \mathbb{R}_n^m(\mathfrak{a}, \mathfrak{t}, \mathbf{v}) : m \in \mathbb{N} \} \\ \text{and } \mathbb{R}_n^*(\mathfrak{a}, \mathfrak{t}, \mathbf{v}) &:= \mathbb{R}_n^{m_n^*}(\mathfrak{a}, \mathfrak{t}, \mathbf{v}) = \min \{ \mathbb{R}_n^m(\mathfrak{a}, \mathfrak{t}, \mathbf{v}) : m \in \mathbb{N} \} \end{aligned} \quad (11.14)$$

the OPE $\widehat{\theta}^{m_n^*} = \widehat{\mathfrak{s}}^{(k)\dagger} \widehat{g}^{m_n^*} \in \ell_2(\mathbf{v}^2)$ with optimally chosen dimension $m_n^* = m_n^*(\mathfrak{a}, \mathfrak{t}, \mathbf{v})$ as in (11.22) fulfills

$$\begin{aligned} \mathcal{R}_{n,k}^{\mathbf{v}}[\widehat{\theta}^{m_n^*} | \ell_2^{\mathfrak{a},r}, \mathbb{M}_{t,d}, \{\sigma\}, \{\xi\}] &\leq \mathbb{R}_n^*(\mathfrak{a}, \mathfrak{t}, \mathbf{v}) \vee \|(\mathfrak{a}\mathbf{v})^2 (1 \vee k \mathfrak{t}^2)^{-1}\|_{\ell_{\infty}} \\ &\quad \times (r^2 + 4K_{\xi}^2 K_{\sigma}^2 d^2 + 8K_{\xi}^4 r^2 d^2) \quad \forall n, k \in \mathbb{N} \end{aligned} \quad (11.15)$$

with $K_{\sigma} := \|\sigma\|_{\ell_{\infty}} \vee 1 \in \mathbb{R}_{\geq 1}$ and $K_{\xi} := \|\xi\|_{\ell_{\infty}} \vee 1 \in \mathbb{R}_{\geq 1}$. \square

§1101.20 **diSM with noisy operator** (§10100.10 continued). Consider $(\widehat{g} = g + n^{-1/2} \dot{\xi}, \widehat{\mathfrak{s}} = \mathfrak{s} + k^{-1/2} \dot{\eta}) \sim \mathbb{P}_{\theta|\sigma|\xi}^{n \otimes k} = \mathbb{P}_{\theta|\sigma}^n \otimes \mathbb{P}_{\xi}^k$ as in Model §10100.10, where $\dot{\eta} \sim \otimes_{j \in \mathbb{N}} \mathbb{P}^{\eta_j}$ fulfills **(SMnO:ub)** in Model §10100.10 with $\xi_j \in \Xi \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_{\infty}$ and hence $\mathbf{v}_j^{\eta} := \mathbb{P}(\dot{\eta}_j^2) \leq \xi_j^2$ for each $j \in \mathbb{N}$. In addition **(SMnO:lb)** there exists $\mathbb{P}_{(0,1)} \in \mathcal{W}(\mathcal{B})$ fulfilling Assumption §11100.07 with $C_{\eta_j}, x_{\eta_j} \in \mathbb{R}_{>0}$ such that $\mathbb{P}^{\eta_j} = \mathbb{P}_{(0,\mathbf{v}^{\eta_j})}$ for each $j \in \mathbb{N}$ and $(\mathbf{v}^{\eta_j})^{-1} \in \ell_{\infty}$. \square

§1101.21 **Corollary** (diSM with noisy operator §1101.20 continued). Consider $\widehat{\mathfrak{s}} = \mathfrak{s} + k^{-1/2} \dot{\eta} \sim \mathbb{P}_{\xi}^k$ as in Model §1101.20, where $\dot{\eta}$ fulfills **(SMnO:ub)** and **(SMnO:lb)** with $C_{\eta_j}, x_{\eta_j} \in \mathbb{R}_{>0}$ and $\|(\mathbf{v}^{\eta_j})^{-1}\|_{\ell_{\infty}} =: \mathbf{v}_{\eta_j} \in \mathbb{R}_{>0}$. For any $\mathfrak{s}_*^0, \mathfrak{s}_*^1 \in \ell_{\infty}$ with $k^{1/2} \mathbf{v}_{\eta_j}^{1/2} \|\mathfrak{s}_*^0 - \mathfrak{s}_*^1\|_{\ell_{\infty}} \leq x_{\eta_j}$ we have $\mathbb{H}^2(\mathbb{P}_{\sigma|\xi}^k, \mathbb{P}_{\sigma|\xi}^k) \leq k \mathbf{v}_{\eta_j} C_{\eta_j} \|\mathfrak{s}_*^0 - \mathfrak{s}_*^1\|_{\ell_2}^2$.

§1101.22 **Proof of Corollary** §1101.21. Given in the lecture. \square

§1101.23 **Proposition** (diSM with noisy operator §1101.20 continued). Consider $(\widehat{g} = g + n^{-1/2} \dot{\xi}, \widehat{\mathfrak{s}} = \mathfrak{s} + k^{-1/2} \dot{\eta}) \sim \mathbb{P}_{\theta|\sigma|\xi}^{n \otimes k} = \mathbb{P}_{\theta|\sigma}^n \otimes \mathbb{P}_{\xi}^k$ as in Model §10100.10, where $\dot{\eta}$ fulfills **(SMnO:ub)** and **(SMnO:lb)** with $C_{\eta_j}, x_{\eta_j} \in \mathbb{R}_{>0}$ and $\|(\mathbf{v}^{\eta_j})^{-1}\|_{\ell_{\infty}} =: \mathbf{v}_{\eta_j} \in \mathbb{R}_{>0}$. If Assumption §10100.16 is satisfied then we have

$$\begin{aligned} \inf_{\widehat{g}} \mathcal{R}_{n,k}^{\mathbf{v}}[\widehat{\theta} | \ell_2^{\mathfrak{a},r}, \mathbb{M}_{t,d}, \{\sigma\}, \{\xi\}] &\geq \|(\mathfrak{a}\mathbf{v})^2 (1 \vee k \mathfrak{t}^2)^{-1}\|_{\ell_{\infty}} \\ &\quad \times \frac{r^2}{16d^2} (\mathbf{v}_{\eta_j}^{-1} C_{\eta_j}^{-1} \wedge \mathbf{v}_{\eta_j}^{-1} x_{\eta_j}^2 \wedge 4(1 - d^{-1})^2) \quad \forall n, k \in \mathbb{N} \end{aligned} \quad (11.16)$$

where the infimum is taken over all possible estimators.

§1101.24 **Proof of Corollary §1101.26.** Given in the lecture. \square

§1101.25 **GdiSM with noisy operator (§0204.06 continued).** Consider a Gaussian diagonal inverse sequence model (GdiSM) with noisy operator as in §0204.06 where $\mathfrak{s}_\cdot \in \ell_\infty$ is *not known* anymore. Here the observable process $\widehat{\mathfrak{s}}_\cdot = \mathfrak{s}_\cdot + k^{-1/2}\dot{W}_\cdot \sim N_{\mathfrak{s}_\cdot}^k$ and $\widehat{g}_\cdot = g_\cdot + n^{-1/2}\dot{B}_\cdot \sim N_{\theta_{\mathfrak{s}_\cdot}}^n$ is a noisy version of $\mathfrak{s}_\cdot \in \mathcal{S} \subseteq \mathbb{R}_{\neq 0}^{\mathbb{N}} \cap \ell_\infty$ and $g_\cdot = \mathfrak{s}_\cdot \theta \in \text{dom}(M_{\mathfrak{s}_\cdot}) \subseteq \ell_2$ with $\theta \in \Theta \subseteq \ell_2$, respectively, where $\dot{B}_\cdot \sim N_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{W}_\cdot \sim N_{(0,1)}^{\otimes \mathbb{N}}$ are *independent*. Consequently, $(\widehat{g}_\cdot, \widehat{\mathfrak{s}}_\cdot)$ admits a joint $N_{\theta_{\mathfrak{s}_\cdot}}^{n \otimes k} = N_{\theta_{\mathfrak{s}_\cdot}}^n \otimes N_{\mathfrak{s}_\cdot}^k$ distribution belonging to the family $N_{\theta_{\mathfrak{s}_\cdot}}^{n \otimes k} := (N_{\theta_{\mathfrak{s}_\cdot}}^n \otimes N_{\mathfrak{s}_\cdot}^k)_{\theta \in \Theta, \mathfrak{s}_\cdot \in \mathcal{S}}$. Summarising the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, N_{\theta_{\mathfrak{s}_\cdot}}^{n \otimes k})$ where $\Theta \subseteq \ell_2$ and $\mathcal{S} \subseteq \mathbb{R}_{\neq 0}^{\mathbb{N}} \cap \ell_\infty$. Under Assumption §0702.32 (which is implied by Assumption §1000.14) in **Corollary §0702.37** an upper bound for the maximal global \mathfrak{v} -risk of a tOPE is shown. More precisely, the performance of the tOPE $\widehat{\theta}_\cdot^m = \widehat{\mathfrak{s}}_\cdot^{(k) \dagger} \widehat{g}_\cdot^m \in \ell_2(\mathfrak{v}^2)$ with dimension $m \in \mathbb{N}$ is measured by its maximal global \mathfrak{v} -risk over the ellipsoid $\Theta = \ell_2^{\mathfrak{a}, \mathfrak{r}}$ with $\mathfrak{r} \in \mathbb{R}_{>0}$ and the link condition $\mathbb{M}_{\mathfrak{t}, \mathfrak{d}}$ with $\mathfrak{d} \in \mathbb{R}_{\geq 1}$, that is

$$\mathcal{R}_{n,k}^{\mathfrak{v}}[\widehat{\theta}_\cdot^m | \ell_2^{\mathfrak{a}, \mathfrak{r}}, \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}] := \sup \{ N_{\theta_{\mathfrak{s}_\cdot}}^{n \otimes k} (\|\widehat{\theta}_\cdot^{m^*} - \theta\|_{\mathfrak{v}}^2) : \theta \in \ell_2^{\mathfrak{a}, \mathfrak{r}}, M_{\mathfrak{s}_\cdot} \in \mathbb{M}_{\mathfrak{t}, \mathfrak{d}} \} \quad \forall n, k, m \in \mathbb{N}.$$

The tOPE $\widehat{\theta}_\cdot^{m^*} = \widehat{\mathfrak{s}}_\cdot^{(k) \dagger} \widehat{g}_\cdot^{m^*} \in \ell_2(\mathfrak{v}^2)$ with optimally chosen dimension $m_n^* = m_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ as in (11.22) fulfills $\mathcal{R}_{n,k}^{\mathfrak{v}}[\widehat{\theta}_\cdot^{m^*} | \ell_2^{\mathfrak{a}, \mathfrak{r}}, \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}] \leq R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) \vee \|(\mathfrak{a}\mathfrak{v})_\cdot^2 (1 \vee k\mathfrak{t}_\cdot^2)^{-1}\|_{\ell_\infty} \times (\mathfrak{r}^2 + 4\mathfrak{d}^2 + 12\mathfrak{r}^2\mathfrak{d}^2)$ for all $n, k \in \mathbb{N}$. \square

§1101.26 **Corollary** (GdiSM with noisy operator §1101.25 continued). *Consider $(\widehat{g}_\cdot = g_\cdot + n^{-1/2}\dot{B}_\cdot, \widehat{\mathfrak{s}}_\cdot = \mathfrak{s}_\cdot + k^{-1/2}\dot{W}_\cdot) \sim N_{\theta_{\mathfrak{s}_\cdot}}^{n \otimes k} = N_{\theta_{\mathfrak{s}_\cdot}}^n \otimes N_{\mathfrak{s}_\cdot}^k$ as in Model §1101.25, where $\dot{B}_\cdot \sim N_{(0,1)}^{\otimes \mathbb{N}}$ and $\dot{W}_\cdot \sim N_{(0,1)}^{\otimes \mathbb{N}}$ are independent. Let Assumption §1000.14 be satisfied. Then we have*

$$\inf_{\widehat{\theta}_\cdot} \mathcal{R}_{n,k}^{\mathfrak{v}}[\widehat{\theta}_\cdot | \ell_2^{\mathfrak{a}, \mathfrak{r}}, \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}] \geq \|(\mathfrak{a}\mathfrak{v})_\cdot^2 (1 \vee k\mathfrak{t}_\cdot^2)^{-1}\|_{\ell_\infty} \times \frac{\mathfrak{r}^2}{32\mathfrak{d}^2} (1 \wedge 8(1 - \mathfrak{d}^{-1})^2) \quad \forall n, k \in \mathbb{N} \quad (11.17)$$

where the infimum is taken over all possible estimators.

§1101.27 **Proof of Corollary §1101.26.** Given in the lecture. \square

§1102 Lower bound based on m hypothesis

§1102.01 **Notation.** For $m \in \mathbb{N}$ set $\mathcal{T}_m := \{-1, 1\}^m$ and for each $\tau := (\tau_j)_{j \in \llbracket m \rrbracket} \in \mathcal{T}_m$ and $j \in \llbracket m \rrbracket$ introduce $\tau^{(j)} \in \mathcal{T}_m$ given by $\tau_j^{(j)} := -\tau_j$ and $\tau_l^{(j)} := \tau_l$ for $l \in \llbracket m \rrbracket \setminus \{j\}$. \square

§1102.02 **Lemma (Assouad's cube technique).** *Given a noise level $n \in \mathbb{N}$ let $\mathbb{P}_\Theta^n := (\mathbb{P}_\theta^n)_{\theta \in \Theta}$ be a family of probability measures. Suppose there exist $m \in \mathbb{N}$ and distances $\mathfrak{d}_{\text{ist}}^{(j)}(\cdot, \cdot)$, $j \in \llbracket m \rrbracket$ such that $\mathfrak{d}_{\text{ist}}^2(\cdot, \cdot) \geq \sum_{j \in \llbracket m \rrbracket} |\mathfrak{d}_{\text{ist}}^{(j)}(\cdot, \cdot)|^2$. We measure the accuracy of an estimator θ by its maximal risk*

$$\mathcal{R}_n[\widetilde{\theta} | \Theta] := \sup \{ \mathbb{E}_\theta^n (\mathfrak{d}_{\text{ist}}^2(\widetilde{\theta}, \theta)) : \theta \in \Theta \}.$$

(i) *If there exists $\{\theta^\tau : \tau \in \mathcal{T}_m\} \subseteq \Theta$ such that for all $\tau \in \mathcal{T}_m$ and $j \in \llbracket m \rrbracket$ we have Assumption §1100.11 and $H(\mathbb{P}_{\theta^\tau}^n, \mathbb{P}_{\theta^{(j)}}^n) \leq 1$ then we obtain*

$$\inf_{\widetilde{\theta}} \mathcal{R}_n[\widetilde{\theta} | \Theta] \geq 2^{-m} \sum_{\tau \in \mathcal{T}_m} \frac{1}{16} \sum_{j \in \llbracket m \rrbracket} |\mathfrak{d}_{\text{ist}}^{(j)}(\theta^\tau, \theta^{\tau^{(j)}})|^2 \quad (11.18)$$

where the infimum is taken over all possible estimators.

(ii) Let $n \in \mathbb{N}_{\geq 2}$ and for each $\theta \in \Theta$ let $\mathbb{P}_\theta^n = \otimes_{j \in \llbracket n \rrbracket} \mathbb{P}_\theta$ be a product probability measure with identically \mathbb{P}_θ -distributed marginals. If there exists $\{\theta^\tau: \tau \in \mathcal{T}_m\} \subseteq \Theta$ $\tau \in \mathcal{T}_m$ and $j \in \llbracket m \rrbracket$ we have Assumption §1100.11 and the marginals satisfy $H(\mathbb{P}_\theta, \mathbb{P}_{\theta^{(j)}}) \leq 2n^{-1}$ then we have

$$\inf_{\tilde{\theta}} \mathcal{R}_n[\tilde{\theta} | \Theta] \geq 2^{-m} \sum_{\tau \in \mathcal{T}_m} \frac{1}{64} \sum_{j \in \llbracket m \rrbracket} |\mathfrak{d}_{\text{ist}}^{(j)}(\theta^\tau, \theta^{\tau^{(j)}})|^2 \quad (11.19)$$

where the infimum is taken over all possible estimators.

§1102.03 **Proof** of Lemma §1102.02. Given in the lecture. \square

§1102.04 **Remark** (Lower bound for a global \mathfrak{v} -risk). For $a, b \in \ell_2(\mathfrak{v}^2)$ consider $\mathfrak{d}_{\text{ist}}(a, b) = \|a - b\|_{\mathfrak{v}}$. Evidently, for each $j \in \mathbb{N}$ setting $\mathfrak{d}_{\text{ist}}^{(j)}(a, b) := |\mathfrak{v}_j(a_j - b_j)|$ we have

$$\mathfrak{d}_{\text{ist}}^2(a, b) = \|a - b\|_{\mathfrak{v}}^2 \geq \sum_{j \in \llbracket m \rrbracket} \mathfrak{v}_j^2 |a_j - b_j|^2 = \sum_{j \in \llbracket m \rrbracket} |\mathfrak{d}_{\text{ist}}^{(j)}(a, b)|^2 \quad \forall m \in \mathbb{N}.$$

Consequently, a lower bound for a global \mathfrak{v} -risk can be obtained elementarily from Lemma §1102.02, which in case (i) for any estimator $\tilde{\theta}$ states

$$\mathcal{R}_n^{\mathfrak{v}}[\tilde{\theta} | \Theta] := \sup \{ \mathbb{P}_\theta^n (\|\tilde{\theta} - \theta\|_{\mathfrak{v}}^2) : \theta \in \Theta \} \geq 2^{-m} \sum_{\tau \in \mathcal{T}_m} \frac{1}{16} \sum_{j \in \llbracket m \rrbracket} \mathfrak{v}_j^2 |\theta_j^\tau - \theta_j^{\tau^{(j)}}|^2$$

If we consider furthermore candidates $\{\theta^\tau := (\tau_j \theta_j^* \mathbf{1}_j^m)_{j \in \mathbb{N}} : \tau \in \mathcal{T}_m\} \subseteq \Theta \subseteq \ell_2(\mathfrak{v}^2)$ for some $\theta^* \in \ell_2(\mathfrak{v}^2)$, then it is easily seen that $\sum_{j \in \llbracket m \rrbracket} \mathfrak{v}_j^2 |\theta_j^\tau - \theta_j^{\tau^{(j)}}|^2 = 4 \sum_{j \in \llbracket m \rrbracket} \mathfrak{v}_j^2 |\theta_j^*|^2 = 4 \|\theta^* \mathbf{1}^m\|_{\mathfrak{v}}^2$ which in turn under the conditions of Lemma §1102.02 (i) implies

$$\inf_{\tilde{\theta}} \mathcal{R}_n^{\mathfrak{v}}[\tilde{\theta} | \Theta] \geq 2^{-m} \sum_{\tau \in \mathcal{T}_m} \frac{1}{4} \|\theta^* \mathbf{1}^m\|_{\mathfrak{v}}^2 = \frac{1}{4} \|\theta^* \mathbf{1}^m\|_{\mathfrak{v}}^2. \quad (11.20)$$

Similarly, under the conditions of Lemma §1102.02 (ii) we get

$$\inf_{\tilde{\theta}} \mathcal{R}_n^{\mathfrak{v}}[\tilde{\theta} | \Theta] \geq 2^{-m} \sum_{\tau \in \mathcal{T}_m} \frac{1}{16} \|\theta^* \mathbf{1}^m\|_{\mathfrak{v}}^2 = \frac{1}{16} \|\theta^* \mathbf{1}^m\|_{\mathfrak{v}}^2. \quad (11.21)$$

Often a minimax-optimal lower bound can be found by choosing the parameter m and constructing a candidate $\theta^* = \mathbf{U}\theta^*$ that have the largest possible $\|\theta^* \mathbf{1}^m\|_{\mathfrak{v}}^2$ -value although that the associated \mathbb{P}_θ^n , $\tau \in \mathcal{T}_m$ are still statistically indistinguishable in the sense that $H(\mathbb{P}_\theta^n, \mathbb{P}_{\theta^{(j)}}^n) \leq 1$ or $H(\mathbb{P}_\theta, \mathbb{P}_{\theta^{(j)}}) \leq 2n^{-1}$ for all $j \in \llbracket m \rrbracket$ and $\tau \in \mathcal{T}_m$. \square

§1102.05 **Reminder** (Maximal global \mathfrak{v} -risk in diSM (§1000.08 continued)). In Subsection §070101 we consider an orthogonal projection estimator (OPE) in a Diagonal inverse sequence model (diSM) as in Model §01104.08 (summarised in Model §1000.08). Here the observable noisy version \hat{g} satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, \mathbb{P}_{\Theta \times \{\mathfrak{s}\} \times \Sigma}^n := (\mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n)_{\theta \in \Theta, \mathfrak{s} \in \Sigma})$ where $\mathfrak{s} \in \ell_\infty$ is known, $\Theta \subseteq \ell_2$ and $\Sigma \subseteq \mathbb{R}_{>0}^{\mathbb{N}} \cap \ell_\infty$. Under Assumption §070101.30 (which is implied by Assumption §1000.14) in Corollary §070101.40 an upper bound for the maximal global \mathfrak{v} -risk of an OPE is shown. More precisely, assuming a multiplication operator $M_{\mathfrak{s}} \in \mathbb{L}(\mathbb{J})$ (compare Notation §01104.01), which fulfills a link condition $M_{\mathfrak{s}} \in \mathbb{M}_{\mathfrak{t}, \mathfrak{d}}$ for $\mathfrak{d} \in \mathbb{R}_{\geq 1}$ (see Assumption §0403.04), the performance of the OPE $\hat{\theta}^m = \mathfrak{s}^\dagger \hat{g} \mathbf{1}^m \in \text{dom}(\phi_{\mathbb{N}})$ with dimension $m \in \mathbb{N}$ is measured by its maximal global \mathfrak{v} -risk over the ellipsoid $\Theta = \ell_2^{\mathfrak{a}, \mathfrak{r}}$ with $\mathfrak{r} \in \mathbb{R}_{>0}$, that is

$$\mathcal{R}_n^{\mathfrak{v}}[\hat{\theta}^m | \ell_2^{\mathfrak{a}, \mathfrak{r}}, \{M_{\mathfrak{s}}\}, \{\sigma\}] := \sup \{ \mathbb{P}_{\theta|\mathfrak{s}|\sigma}^n (\|\hat{\theta}^m - \theta\|_{\mathfrak{v}}^2) : \theta \in \ell_2^{\mathfrak{a}, \mathfrak{r}} \} \quad \forall n, m \in \mathbb{N}.$$

For $n, m \in \mathbb{N}$ setting (as in (07.07))

$$\begin{aligned} \mathcal{R}_n^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) &:= (\mathbf{av})_{(m)}^2 \vee n^{-1} \|\mathbf{t}^{-1} \mathbf{1}^m\|_{\mathbf{v}}^2, \quad m_n^* := \arg \min \{ \mathcal{R}_n^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) : m \in \mathbb{N} \} \\ \text{and } \mathcal{R}_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) &:= \mathcal{R}_n^{m_n^*}(\mathbf{a}, \mathbf{t}, \mathbf{v}) = \min \{ \mathcal{R}_n^m(\mathbf{a}, \mathbf{t}, \mathbf{v}) : m \in \mathbb{N} \} \end{aligned} \quad (11.22)$$

the OPE $\widehat{\theta}_n^{m_n^*} = \mathfrak{s}^\dagger \widehat{g}_n \mathbf{1}^{m_n^*} \in \ell_2(\mathbf{v}^2)$ with optimally chosen dimension $m_n^* = m_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ as in (11.22) fulfills

$$\mathcal{R}_n^{\mathbf{v}}[\widehat{\theta}_n^{m_n^*} | \ell_2^{\mathbf{a}, \mathbf{r}}, \{M_s\}, \{\sigma_s\}] \leq (\mathbf{v}_\sigma \mathbf{d}^2 + \mathbf{r}^2) \mathcal{R}_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \quad \forall n \in \mathbb{N} \quad (11.23)$$

with $\|\sigma_s^2\|_{\ell_\infty} =: \mathbf{v}_\sigma \in \mathbb{R}_{>0}$. □

§1102.06 **Lemma.** Under Assumption §10100.14 for $m_n^* := m_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ and $\mathcal{R}_n^* := \mathcal{R}_n^{m_n^*}(\mathbf{a}, \mathbf{t}, \mathbf{v})$ as in (11.22) distinguish case i) : $\mathcal{R}_n^* = n^{-1} \|\mathbf{t}^{-1} \mathbf{1}^{m_n^*}\|_{\mathbf{v}}^2 > (\mathbf{av})_{(m_n^*)}^2$ and case ii) : $\mathcal{R}_n^* = (\mathbf{av})_{(m_n^*)}^2 \geq n^{-1} \|\mathbf{t}^{-1} \mathbf{1}^{m_n^*}\|_{\mathbf{v}}^2$. Then for all $n \in \mathbb{N}_{>(\mathbf{v}/\mathbf{t})_1^2(\mathbf{av})_1^2}$, i.e. $(\mathbf{av})_{(1)}^2 > n^{-1}(\mathbf{v}/\mathbf{t})_1^2$, in case i) we have $(\mathbf{av})_{(m_n^*-1)}^2 > n^{-1} \|\mathbf{t}^{-1} \mathbf{1}^{m_n^*}\|_{\mathbf{v}}^2$ and in case ii) setting

$$m_n^\diamond := \min \{ m \in \mathbb{N}_{>m_n^*} : n^{-1} \|\mathbf{t}^{-1} \mathbf{1}^m\|_{\mathbf{v}}^2 \geq (\mathbf{av})_{(m)}^2 \} \quad (11.24)$$

we obtain $(\mathbf{av})_{(m_n^*)}^2 = (\mathbf{av})_{(m_n^\diamond-1)}^2 \leq n^{-1} \|\mathbf{t}^{-1} \mathbf{1}^{m_n^\diamond}\|_{\mathbf{v}}^2$.

§1102.07 **Proof of Lemma §1102.06.** Given in the lecture. □

§1102.08 **Corollary** (diSM §1101.05 continued). For $\mathfrak{s} \in \ell_\infty$, $\theta_j \in \ell_2$, hence $g = \mathfrak{s}\theta \in \text{dom}(M_s) \subseteq \ell_2$, consider $\widehat{g} = g + n^{-1/2} \dot{\varepsilon} \sim P_{\theta|\mathfrak{s}|\sigma}^n$ as in Model §1101.05, where $\dot{\varepsilon}$ fulfills (SM:ub) and (SM:lb) with $C_\varepsilon, x_\varepsilon \in \mathbb{R}_{>0}$ $\|\sigma_s^{-2}\|_{\ell_\infty} =: \mathbf{v}_\sigma \in \mathbb{R}_{>0}$. For each $\theta_j^* \in \ell_2$ with $2n^{1/2} \mathbf{v}_\sigma^{1/2} \|\mathfrak{s}_j \theta_j^* \mathbf{1}_j^m\|_{\ell_\infty} \leq x_\varepsilon$ and for each $\tau \in \mathcal{T}_m = \{-1, 1\}^m$ as in Notation §1102.01 setting $\theta^\tau := (\tau_j \theta_j^* \mathbf{1}_j^m)_{j \in \mathbb{N}}$ the distribution $P_{\theta^\tau|\mathfrak{s}|\sigma}^n \in \mathcal{W}(\mathcal{B}^{\otimes \mathbb{N}})$ satisfies $H^2(P_{\theta^\tau|\mathfrak{s}|\sigma}^n, P_{\theta^{\tau^*}|\mathfrak{s}|\sigma}^n) \leq 4n \mathbf{v}_\sigma C_\varepsilon \|\mathfrak{s}_j \theta_j^* \mathbf{1}_j^m\|_{\ell_\infty}^2$ for all $j \in \llbracket m \rrbracket$.

§1102.09 **Proof of Corollary §1102.08.** Given in the lecture. □

§1102.10 **Proposition** (diSM §1101.05 continued). For $\mathfrak{s} \in \ell_\infty$, $\theta_j \in \ell_2$, hence $g = \mathfrak{s}\theta \in \text{dom}(M_s) \subseteq \ell_2$, consider $\widehat{g} = g + n^{-1/2} \dot{\varepsilon} \sim P_{\theta|\mathfrak{s}|\sigma}^n$ as in Model §1101.05, where $\dot{\varepsilon}$ fulfills (SM:ub) and (SM:lb) with $C_\varepsilon, x_\varepsilon \in \mathbb{R}_{>0}$ and $\|\sigma_s^{-2}\|_{\ell_\infty} =: \mathbf{v}_\sigma \in \mathbb{R}_{>0}$. Let Assumption §10100.14 and in addition (11.10) for $\mathbf{d} \in \mathbb{R}_{\geq 1}$ be satisfied. Then we have

$$\begin{aligned} \inf_{\widehat{g}} \mathcal{R}_n^{\mathbf{v}}[\widehat{g} | \ell_2^{\mathbf{a}, \mathbf{r}}, \{M_s\}, \{\sigma_s\}] &\geq \mathcal{R}_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v}) \\ &\times \frac{1}{16} (4C_{(\mathbf{av})} \mathbf{r}^2 \wedge \mathbf{v}_\sigma^{-1} \mathbf{d}^{-2} (C_\varepsilon^{-1} \wedge x_\varepsilon^2)) \quad \forall n \in \mathbb{N}_{>(\mathbf{v}/\mathbf{t})_1^2(\mathbf{av})_1^2} \end{aligned} \quad (11.25)$$

where the infimum is taken over all possible estimators.

§1102.11 **Proof of Proposition §1102.10.** Given in the lecture. □

§1102.12 **Comment.** By combining the lower bound in Proposition §1102.10 and the upper bound in Corollary §07101.40 for the maximal global \mathbf{v} -risk of an OPE in a diSM §1101.05 we have shown that $\mathcal{R}_n^*(\mathbf{a}, \mathbf{t}, \mathbf{v})$ is a minimax-rate and the OPE with optimally chosen dimension parameter is minimax-optimal (up to a constant). □

§1102.13 **GdiSM (§1101.11 continued).** Recall that the observations satisfy a statistical product experiment $(\mathbb{R}^{\mathbb{N}}, \mathcal{B}^{\otimes \mathbb{N}}, N_{\Theta \times \{\mathfrak{s}\}}^n)$ where $\Theta \subseteq \ell_2$. Under Assumption §07101.30 (which is implied by Assumption §10100.14) in Corollary §07101.38 an upper bound for the maximal global \mathbf{v} -risk of an OPE is shown. More precisely, assuming a multiplication operator $M_s \in \mathbb{L}(\mathbb{J})$ (compare Notation §01104.01), which fulfills a link condition $M_s \in \mathbb{M}_{\mathbf{t}, \mathbf{d}}$ for $\mathbf{d} \in \mathbb{R}_{\geq 1}$ (see Assumption §04103.04),

the performance of the OPE $\hat{\theta}^m = \mathfrak{s}^\dagger \hat{g} \mathbb{1}^m \in \ell_2(\mathfrak{v}^2)$ with dimension $m \in \mathbb{N}$ is measured by its maximal global \mathfrak{v} -risk over the ellipsoid $\Theta = \ell_2^{\mathfrak{a}, \mathfrak{r}}$ with $\mathfrak{r} \in \mathbb{R}_{>0}$, that is

$$\mathcal{R}_n^{\mathfrak{v}}[\hat{\theta}^m | \ell_2^{\mathfrak{a}, \mathfrak{r}}, \{M_s\}] := \sup \{N_{\theta|s}^n(\|\hat{\theta}^m - \theta\|_{\mathfrak{v}}^2) : \theta \in \ell_2^{\mathfrak{a}, \mathfrak{r}}\} \quad \forall n, m \in \mathbb{N}.$$

The OPE $\hat{\theta}^{m_n^*} = \mathfrak{s}^\dagger \hat{g} \mathbb{1}^{m_n^*} \in \ell_2(\mathfrak{v}^2)$ with optimally chosen dimension $m_n^* = m_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ as in (11.22) fulfills $\mathcal{R}_n^{\mathfrak{v}}[\hat{\theta}^{m_n^*} | \ell_2^{\mathfrak{a}, \mathfrak{r}}, \{M_s\}] \leq (d^2 + \mathfrak{r}^2) R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ for all $n \in \mathbb{N}$. \square

§1102.14 **Corollary** (GdiSM §1102.13 continued). *For $\mathfrak{s} \in \ell_\infty$, $\theta_s \in \ell_2$, hence $g_s = \mathfrak{s} \cdot \theta_s \in \text{dom}(M_s) \subseteq \ell_2$, consider $\hat{g} = g + n^{-1/2} \dot{B} \sim N_{\theta|s}^n$ as in Model §1101.11, where $\dot{B} \sim N_{(0,1)}^{\otimes \mathbb{N}}$. Let Assumption §1000.14 and in addition (??) be satisfied. Then we have*

$$\inf_{\hat{\theta}} \mathcal{R}_n^{\mathfrak{v}}[\hat{\theta}^m | \ell_2^{\mathfrak{a}, \mathfrak{r}}, \{M_s\}] \geq R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v}) \times \frac{1}{8} (2\mathfrak{r}^2 \wedge d^{-2}) \quad \forall n \in \mathbb{N}_{>(\mathfrak{v}/\mathfrak{t})^2(\mathfrak{a}\mathfrak{v})^{-2}} \quad (11.26)$$

where the infimum is taken over all possible estimators.

§1102.15 **Proof of Corollary** §1102.14. Given in the lecture. \square

§1102.16 **Comment**. By combining the lower bound in Corollary §1102.14 and the upper bound in Corollary §0701.38 for the maximal global \mathfrak{v} -risk of an OPE in a GdiSM §1102.13 we have shown that $R_n^*(\mathfrak{a}, \mathfrak{t}, \mathfrak{v})$ is a minimax-rate and the OPE with optimally chosen dimension parameter is minimax-optimal (up to a constant). \square

§1102.17 **Remark**. Let $\mathbb{P}_{\Theta \times \Xi}^{n \otimes k} = (\mathbb{P}_{\theta|\xi}^{n \otimes k})_{\theta \in \Theta, \xi \in \Xi}$ be a family of product measures $\mathbb{P}_{\theta|\xi}^{n \otimes k} = \mathbb{P}_{\theta|\xi}^n \otimes \mathbb{P}_\xi^k$ depending on a function of interest $\theta \in \Theta$, a nuisance parameter $\xi \in \Xi$ and noise levels $n, k \in \mathbb{N}$. Supposing there exist $m \in \mathbb{N}$ and distances $\mathfrak{d}_{\text{ist}}^{(j)}(\cdot, \cdot)$, $j \in \llbracket m \rrbracket$ such that $\mathfrak{d}_{\text{ist}}^2(\cdot, \cdot) \geq \sum_{j \in \llbracket m \rrbracket} |\mathfrak{d}_{\text{ist}}^{(j)}(\cdot, \cdot)|^2$ the Lemma §1102.02 allows us to bound from below the maximal risk for each nuisance parameter $\xi \in \Xi$ and noise level $n \in \mathbb{N}$. To be more precise, given noise levels $n, k \in \mathbb{N}$ for each $\tau \in \mathcal{T}_m$ consider $\theta^\tau \in \Theta$ with associated product probability measure $\mathbb{P}_{\theta^\tau|\xi}^{n \otimes k} = \mathbb{P}_{\theta^\tau|\xi}^n \otimes \mathbb{P}_\xi^k$, then for all $j \in \llbracket m \rrbracket$ we have $\rho(\mathbb{P}_{\theta^\tau|\xi}^{n \otimes k}, \mathbb{P}_{\theta^{\tau^{(j)}}|\xi}^{n \otimes k}) = \rho(\mathbb{P}_{\theta^\tau|\xi}^n \otimes \mathbb{P}_\xi^k, \mathbb{P}_{\theta^{\tau^{(j)}}|\xi}^n \otimes \mathbb{P}_\xi^k) = \rho(\mathbb{P}_{\theta^\tau|\xi}^n, \mathbb{P}_{\theta^{\tau^{(j)}}|\xi}^n)$ due to the independence. Consequently, if $H(\mathbb{P}_\tau^n, \mathbb{P}_{\tau^{(j)}}^n) \leq 1$ for all $\tau \in \mathcal{T}_m$ and $j \in \llbracket m \rrbracket$, then for any estimator $\tilde{\theta}$ we obtain

$$\mathcal{R}_{n,k}[\tilde{\theta} | \Theta, \{\xi\}] := \sup \{ \mathbb{E}_{\theta|\xi}^{n \otimes k}(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta \} \geq 2^{-m} \sum_{\tau \in \mathcal{T}_m} \frac{1}{16} \sum_{j \in \llbracket m \rrbracket} |\mathfrak{d}_{\text{ist}}^{(j)}(\theta^\tau, \theta^{\tau^{(j)}})|^2$$

due to Lemma §1102.02. It is worth noting that we obtain the same lower bound when disposing of the family $\mathbb{P}_{\Theta \times \{\xi\}}^n = (\mathbb{P}_{\theta|\xi}^n)_{\theta \in \Theta}$ only, in other words assuming the nuisance parameter $\xi \in \Xi$ is known in advance. \square

§1102.18 **Corollary (Lower bound based on m hypothesis)**. *Let $\mathbb{P}_{\Theta \times \Xi}^{n \otimes k} = (\mathbb{P}_{\theta|\xi}^{n \otimes k})_{\theta \in \Theta, \xi \in \Xi}$ be a family of product measures $\mathbb{P}_{\theta|\xi}^{n,k} = \mathbb{P}_{\theta|\xi}^n \otimes \mathbb{P}_\xi^k$ depending on a function of interest $\theta \in \Theta$, a nuisance parameter $\xi \in \Xi$ and noise levels $n, k \in \mathbb{N}$. Suppose there exist distances $\mathfrak{d}_{\text{ist}}^{(j)}(\cdot, \cdot)$, $j \in \llbracket m \rrbracket$ such that $\mathfrak{d}_{\text{ist}}^2(\cdot, \cdot) \geq \sum_{j \in \llbracket m \rrbracket} |\mathfrak{d}_{\text{ist}}^{(j)}(\cdot, \cdot)|^2$. If there exists $\{(\theta^\tau, \xi^\tau) : \tau \in \mathcal{T}_m\} \subseteq \Theta \times \Xi$ such that for all $\tau \in \mathcal{T}_m$ and $j \in \llbracket m \rrbracket$ Assumption §1100.11, (C1) $\mathbb{P}_{\theta^\tau|\xi^\tau}^n = \mathbb{P}_{\theta^{\tau^{(j)}}|\xi^{\tau^{(j)}}}^n$ and (C2) $H(\mathbb{P}_{\xi^\tau}^k, \mathbb{P}_{\xi^{\tau^{(j)}}}^k) \leq 1$ are fulfilled, then for any estimator $\tilde{\theta}$ we have*

$$\mathcal{R}_{n,k}[\tilde{\theta} | \Theta, \Xi] := \sup \{ \mathbb{E}_{\theta|\xi}^{n \otimes k}(\mathfrak{d}_{\text{ist}}^2(\tilde{\theta}, \theta)) : \theta \in \Theta, \xi \in \Xi \} \geq 2^{-m} \sum_{\tau \in \mathcal{T}_m} \frac{1}{16} \sum_{j \in \llbracket m \rrbracket} |\mathfrak{d}_{\text{ist}}^{(j)}(\theta^\tau, \theta^{\tau^{(j)}})|^2.$$

§1102.19 **Proof of Corollary** §1102.18. Given in the lecture. \square

§1102.20 **Remark.** The last assertion allows us often to derive a lower bound depending on the classes Θ and Ξ and the noise level k but not on the noise level n . Roughly speaking this means that we cover the influence of the estimation of the nuisance parameter. Typically we combine this lower bound with the lower bound obtained in **Lemma** §1102.02 where the nuisance parameter is assumed to be known in advance. \square

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Index

Density deconvolution

- additive on \mathbb{R} , 17
 - unknown error density, 35
- circular on $[0, 1)$, 15
 - unknown error density, 35
- multiplicative on $\mathbb{R}_{>0}$, 19
 - unknown error density, 36

Density estimation

- on $[0, 1)$, 8
- on \mathcal{D} , 6
- on \mathbb{R} , 10
- on $\mathbb{R}_{>0}$, 11

Empirical mean function, 4

Empirical mean model, 5, 23

- direct, 12
- inverse diagonal, 14, 63, 64, 66, 68, 72, 74
- noisy operator, 34, 76, 79, 81, 85, 87
- inverse non-diagonal, 20, 90, 92, 95, 97, 100
- noisy operator, 27, 103, 104, 107, 110, 114, 117

Functional linear regression, 28

- instrumental, 29
- second order stationarity, 36
 - instrumental, 37

Galerkin solution, 48, 51, 52, 54

- generalised, 52, 53

Inverse problem, 39

- ill-posed, 39

Noisy version

- image, 5
- operator, 23
 - non-diagonal, 31

Operator

- conditional expectation, 24
- convolution
 - additive on \mathbb{R} , 17

- circular on $[0, 1)$, 15

- multiplicative on $\mathbb{R}_{>0}$, 18

covariance, 3, 25

- second order stationarity, 31

cross-covariance, 25

- second order stationarity, 32

design, 26

Operator classes

- link condition, $\mathbb{M}_{t,d}$, 45

- link condition, $\mathbb{T}_{t,d}$, 48, 58

- generalised, $\mathbb{T}_{t,d,D}$, 52

- source condition, $\text{ran}((\mathbb{T}_{t,d}^* \mathbb{T}_{t,d})^{s/2})$, 49

- source condition, $\text{ran}((A^* A)^{s/2})$, 57

Regression

- instrumental, 28

- known design, 22

- uniform design, 7, 9

- inverse diagonal, 14

- inverse non-diagonal, 21

- unknown design, 30

Sequence model, 5

- Bivariate, 23

- Gaussian, 24

- direct, 12, 124

- direct Gaussian, 12, 123

- Gaussian, 6

- inverse diagonal, 14, 63, 66, 68, 71, 74,

- 124, 129, 130, 134, 135

- noisy operator, 34, 76, 79, 81, 84, 87,

- 125, 132

- inverse diagonal Gaussian, 14, 62, 66,

- 68, 71, 73, 130, 131, 136

- noisy operator, 34, 75, 76, 78, 81, 84,

- 87, 133

- inverse non-diagonal, 21, 89, 90, 92, 94,

- 97, 99

- noisy operator, 27, 103, 106, 110, 114,

- 117

- inverse non-diagonal Gaussian, 21, 89,

- 92, 94, 97, 99

noisy operator, 27, 102, 106, 109, 113,
117

Solution classes

abstract smoothness condition, $\mathbb{J}^{a,r}$, 43

Statistical problem

direct, 12

inverse diagonal, 13

noisy operator, 34

inverse non-diagonal, 20

noisy operator, 26, 100