

# On Infinite-Dimensional Integration in Weighted Hilbert Spaces

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# Introduction

Finite-dimensional integration:

- $f(x_1, \dots, x_s) = \sum_{u \subseteq \{1, \dots, s\}} f_u(\mathbf{x}_u), \mathbf{x}_u = (x_j)_{j \in u}$

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$$\|f\| = \sum_{u \subseteq \{1, \dots, s\}} \gamma_u^{-1/2} \|f_u\|$$

- $\rightarrow$  *reproducing kernel Hilbert spaces*

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Infinite-dimensional integration:

- $f(\mathbf{x}) = \sum_u f_u(\mathbf{x}_u)$ , finite subsets  $u \subseteq \mathbb{N}$

- How to define integration for infinitely many variables?

- 1 The Setting
- 2 The Function Spaces
- 3 Finite-dimensional Integration
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## Basic assumptions, domain

(A1)  $k \neq 0$  reproducing kernel on  $D \times D$  where  $D \neq \emptyset$

(A2)  $H(k) \cap H(1) = \{0\}$

(A3)  $(\gamma_u)_u$  family of non-negative weights such that

$$\sum_u \gamma_u m^{|u|} < \infty, \quad m := \inf_{x \in D} k(x, x)$$

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The domain for functions of infinitely-many variables:

$$\mathfrak{X} = \left\{ \mathbf{x} \in D^{\mathbb{N}} : \sum_u \gamma_u \prod_{j \in u} k(x_j, x_j) < \infty \right\} \quad (\text{A3}) \Leftrightarrow \mathfrak{X} \neq \emptyset$$

# The kernel

Notation:  $1 : s = \{1, \dots, s\}$

Define reproducing kernels on  $\mathfrak{X} \times \mathfrak{X}$  by

$$k_u(\mathbf{x}, \mathbf{y}) := \prod_{j \in u} k(x_j, y_j)$$

$$K_{1:s}(\mathbf{x}, \mathbf{y}) := \sum_{u \subseteq 1:s} \gamma_u k_u(\mathbf{x}, \mathbf{y})$$

$$K(\mathbf{x}, \mathbf{y}) := \sum_u \gamma_u k_u(\mathbf{x}, \mathbf{y}) = \lim_{s \rightarrow \infty} K_{1:s}(\mathbf{x}, \mathbf{y})$$

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# The function spaces

- **Proposition** *Wasilkowski, Woźniakowski (2004)*  
Every  $f \in H(K_{1:s})$  has a unique decomposition

$$f = \sum_{u \subseteq 1:s} f_u, \quad f_u \in H(\gamma_u k_u).$$

Moreover,

$$\|f\|_{K_{1:s}} = \sum_{u \subseteq 1:s} \gamma_u^{-1/2} \|f_u\|_{k_u}.$$

- proof based on inductive argument

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- proof based on inductive argument
- **Proposition** *Gnewuch, Mayer, Ritter (2012)* Given (A1) - (A3):

$$H(K) = \bigoplus_u H(\gamma_u k_u).$$

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## Further basic assumptions

In addition to (A1) - (A3):

(A4)  $\rho$  a probability measure on  $D$

(A5)  $k$  measurable and  $\rho(\{x \in D : k(x, x) = 0\}) = 0$

(A6)  $H(k) \subseteq L_1(\rho)$

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(A6)  $H(k) \subseteq L_1(\rho)$

$H(k) \subseteq L_1(\rho)$  holds true if e.g.  $k$  square-root integrable along the diagonal

$$\int_D \sqrt{k(x, x)} \rho(dx) < \infty.$$

- $\left\{ \begin{array}{l} H(k) \subseteq L_1(\rho), \\ \text{closed graph theorem} \end{array} \right\} \implies J : H(k) \rightarrow L_1(\rho) \text{ bounded}$

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- representer  $h \in H(k)$ :

$$\int_D g \, d\rho = \langle g, h \rangle_k$$

$$\|h\|_k^2 = \int_D \int_D k(x, y) \rho(dx) \rho(dy)$$



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- if  $k \geq 0$  then  $\|J\| = \|h\|_k$ .  
(see e.g. *Hinrichs (2010)*, *Gnewuch, Mayer, Ritter (2012)*)

## Integration on $H(k_u)$ ?

- For finite  $u \subseteq \mathbb{N}$  consider

$$h_u(\mathbf{x}) = \prod_{j \in u} h(x_j), \quad \mathbf{x} \in D^u.$$

- $h_u \in H(k_u)$ ,  $\|h_u\|_{k_u} = \|h\|_k^{|u|}$ .

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- Straight-forward calculation: for  $f \in \text{span} \{k_u(\cdot, \mathbf{x}) : \mathbf{x} \in D^u\}$

$$\int_{D^u} f \, d\rho^u = \langle f, h_u \rangle_{k_u}$$

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- What about the closure? I.e. is

$$H(k_u) \subseteq L_1(\rho^u)?$$

# Continuous embedding $H(k_u) \subseteq L_1(\rho)$

Proposition (*Gnewuch, Mayer, Ritter (2012)*)

Suppose  $u \neq \emptyset$ . Then  $H(k_u) \subseteq L_1(\rho^u)$  and

$$\|h\|_k^{|u|} \leq \|J_u\| \leq \left(\sqrt{\pi/2} \|J\|\right)^{|u|}.$$

Furthermore,

$$\int_{D^u} f \, d\rho^u = \langle f, h_u \rangle_{k_u}, \quad f \in H(k_u).$$

- Proof: Little Grothendieck Theorem, change of measure argument; cf. *Hinrichs (2010)*.

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- $k \geq 0 \implies \|J_u\| = \|J\|^{|u|} = \|h\|_k^{|u|}$ .

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# Infinite-dimensional integration

- Put  $\mu = \rho^{\mathbb{N}}$ .
- Perspective 1: Integration with respect to  $\mu$ ,

$$I_1(f) = \int_{\mathfrak{X}} f(\mathbf{x}) \mu(d\mathbf{x})$$

if  $\mu(\mathfrak{X}) = 1$ .

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- Perspective 2: Integral as limit of representers. If

$$h^* = \sum_u \gamma_u h_u$$

exists in  $H(K)$ , define

$$I_2(f) = \langle f, h^* \rangle_K.$$

See *Kuo, Sloan, Wasilkowski, Woźniakowski (2010)*,  
*Plaskota, Wasilkowski (2011)*.

# Integration with respect to the product measure

- For  $f = \sum_u f_u \in H(K)$  let

$$f^{(s)} : D^{\mathbb{N}} \rightarrow \mathbb{R}, \quad f^{(s)}(\mathbf{y}) = \sum_{u \subseteq 1:s} f_u(y_u).$$

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- **Definition:** We say

$$H(K) \hookrightarrow L_1(\mu)$$

if  $H(K) \subseteq L_1(\mu)$  and

$$T : H(K) \rightarrow L_1(\mu), \quad Tf = \lim_{s \rightarrow \infty} f^{(s)}$$

defines a bounded linear mapping.

# Integration with respect to the product measure

## Lemma

If  $\mu(\mathfrak{X}) = 1$  and  $H(K) \subseteq L_1(\mu)$  then  $H(K) \hookrightarrow L_1(\mu)$  and

$$\int_{D^{\mathbb{N}}} Tf \, d\mu = \int_{\mathfrak{X}} f \, d\mu|_{\mathfrak{X}}.$$

- $\mu(\mathfrak{X}) = 1$  and  $H(K) \subseteq L_1(\mu)$  are satisfied if e.g.

$$\sum_u \gamma_u \left( \int_D k(x, x) \, d\rho(x) \right)^{|u|} < \infty.$$

### Proposition (Gnewuch, Mayer, Ritter (2012))

Let  $f = \sum_u f_u \in H(K)$ ,  $f_u \in H(\gamma_u k_u)$ .

- If  $\sum_u \gamma_u \|h\|_k^{2|u|} < \infty$  then  $h^* = \sum_u \gamma_u h_u \in H(K)$  and

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- If  $H(K) \hookrightarrow L_1(\mu)$  then

$$\langle f, h^* \rangle = \int_{D^{\mathbb{N}}} Tf(\mathbf{x}) \mu(d\mathbf{x}).$$

## Example 1: Non-negative Kernel

- $\rho$  the uniform distribution on  $D = [0, 1]$
- $k(x, y) = \min(x, y)$
- Fix  $\alpha > 0$ ;  $\gamma_{1:s} = \alpha^s$ ;  $\gamma_u = 0$  otherwise.

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$$\mu(\mathfrak{X}) = 1 \Leftrightarrow \alpha < \exp(1).$$

- For  $\alpha \in [\exp(1), 3)$ ,  $H(K) \hookrightarrow L_1(\mu)$  holds true, but  $\mu(\mathfrak{X}) = 0$ .

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$$\mathfrak{X} = D^{\mathbb{N}} \Leftrightarrow \alpha < 1.$$

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- $H(K) \hookrightarrow L_1(\mu) \not\Rightarrow \sum_u \gamma_u \left( \int_D k(x, x) d\rho(x) \right)^{|u|} < \infty$



## Example 2: ANOVA-type kernel

- $\rho$  uniform distribution on  $D = [-1, 1]$
- $k(x, y) = x \cdot y$
- $\gamma_u$  as before

$$\sum_u \gamma_u \|h\|_k^{2|u|} < \infty \Leftrightarrow \alpha > 0 \quad (\text{trivial}),$$

$$\alpha < 4 \Leftrightarrow \sum_u \gamma_u \|J_u\|^2 < \infty \Rightarrow H(K) \hookrightarrow L_1(\mu) \Rightarrow \alpha \leq 4,$$

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For  $\alpha \in (4, \exp(2))$

- $\langle f, h^* \rangle_K$  well-defined and  $\mu(\mathfrak{X}) = 1$ .
- $H(K) \hookrightarrow L_1(\mu)$  not satisfied.

# Summary, open questions

- Sometimes the number

$$I_2(f) = \langle f, h^* \rangle = \lim_{s \rightarrow \infty} \sum_{u \subseteq \mathbb{1}:s} \int_{D^u} f_u \, d\rho^u$$

may exist, but can not be interpreted as

$$I_1(f) = \int_{D^{\mathbb{N}}} Tf \, d\mu.$$

- $I_1(f)$  can make sense non-trivially even if  $\mu(\mathfrak{X}) = 0$ .

Open questions:

- How to characterize  $H(k) \subseteq L_1(\rho)$ ?
- A necessary condition for  $H(K) \hookrightarrow L_1(\mu)$ ?