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Approximation of Random Fields in High Dimension

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Outline

- 1 Main objects and setup
 - Tensor product-type random fields
 - Average case information complexity
- 2 Exact asymptotic expression for the information complexity
 - Auxiliary probabilistic construction
 - Main result
- 3 Proof
- 4 Some references

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Definition of tensor product-type random fields

- ▶ for any $d = 1, 2, \dots$: separable zero-mean random function $\{X^{(d)}(t)\}$, $t \in [0, 1]^d$ with the covariance function

$$\mathcal{K}^{(d)}(s, t) = \prod_{\ell=1}^d \mathcal{K}_{\ell}(s_{\ell}, t_{\ell})$$

for all $s_{\ell}, t_{\ell} \in [0, 1]$, $s = (s_1, \dots, s_d)$, $t = (t_1, \dots, t_d)$.

- ▶ family of tensor product-type random fields

$$\mathbb{X} = \left\{ X^{(d)}(t), t \in [0, 1]^d \right\}, \quad d = 1, 2, \dots$$

$d \rightarrow \infty$.

Some examples of tensor product-type random fields

- ▶ Wiener-Chentsov random field:

$$\mathcal{K}_{W^{(d)}}(s, t) = \prod_{\ell=1}^d \min\{s_{\ell}, t_{\ell}\}.$$

- ▶ d -variate completely tugged Brownian sheet (Hoefding, Blum, Kiefer and Rosenblatt process):

$$\mathcal{K}_{B^{(d)}}(s, t) = \prod_{l=1}^d (\min\{s_{\ell}, t_{\ell}\} - s_{\ell}t_{\ell}).$$

$$s_{\ell}, t_{\ell} \in [0, 1], s = (s_1, \dots, s_d), t = (t_1, \dots, t_d).$$

Karhunen-Loève expansion

$$\Lambda := \sum_{i=1}^{\infty} \lambda_i^2 < \infty.$$

λ_i^2 and $\varphi_i(\cdot)$: $\lambda_i^2 \varphi_i(t) = \int_0^1 \mathcal{K}(s, t) \varphi_i(s) ds$, $t \in [0, 1]$.

The covariance operator of $X^{(d)}$ has the **eigenvalues**:

$$\lambda_{\mathbf{k}}^2 := \lambda_{k_1}^2 \lambda_{k_2}^2 \dots \lambda_{k_d}^2, \quad \mathbf{k} = (k_1, k_2, \dots, k_d) \in \mathbb{N}^d.$$

Karhunen-Loève expansion: with $\varphi_{\mathbf{k}}(\mathbf{t}) = \varphi_{k_1}(t_1) \dots \varphi_{k_d}(t_d)$

$$X^{(d)}(\mathbf{t}) = \sum_{\mathbf{k} \in \mathbb{N}^d} \xi_{\mathbf{k}} \lambda_{\mathbf{k}} \varphi_{\mathbf{k}}(\mathbf{t}), \quad \mathbf{t} = (t_1, \dots, t_d) \in [0, 1]^d, \quad (1.1)$$

where $\{\xi_{\mathbf{k}}\}$ is an array of non-correlated $\mathcal{N}(0, 1)$ r.v.'s.

Average case information complexity

Let $X := X^{(d)}$ and X_n be the partial sum of the Karhunen-Loève expansion (1.1) corresponding to n maximal eigenvalues.

The average case information complexity:

$$n(\varepsilon, d) := \min\{n : \mathbb{E}\|X - X_n\|^2 \leq \varepsilon^2 \Lambda^d\},$$

where

$$\Lambda^d = \left(\sum_{i=1}^{\infty} \lambda_i^2 \right)^d = \sum_{\mathbf{k} \in \mathbb{N}^d} \lambda_{\mathbf{k}}^2 = \mathbb{E}\|X\|^2.$$

Aim: the asymptotic behavior of $n(\varepsilon, d)$, as $d \rightarrow \infty$.

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Auxiliary sequence of i.i.d. r.v.'s

i.i.d. $\{U_\ell\}$: $\mathbb{P}(U_\ell = -\log \lambda_i) = \lambda_i^2/\Lambda$, $i = 1, 2, \dots$.

(3dM): $\sum_{i=1}^{\infty} |\log \lambda_i|^3 \lambda_i^2 < \infty$, then $\mathbb{E}|U_\ell|^3 < \infty$.

$$M := \mathbb{E}U_\ell = -\sum_{i=1}^{\infty} \log \lambda_i \frac{\lambda_i^2}{\Lambda},$$

$$\sigma^2 := \text{Var } U_\ell = \sum_{i=1}^{\infty} |\log \lambda_i|^2 \frac{\lambda_i^2}{\Lambda} - M^2.$$

$$\alpha^3 := \mathbb{E}(U_\ell - M)^3 = -\sum_{i=1}^{\infty} (\log \lambda_i)^3 \frac{\lambda_i^2}{\Lambda} - 3M\sigma^2 - M^3.$$

Explosion coefficient

$$\mathcal{E} := \Lambda e^{2M}.$$

Lemma

$$\mathcal{E} > 1.$$

$$\mathcal{E} = 1 \iff \sigma = 0.$$

Proof: by concavity of the logarithmic function.

Information complexity

Recall: $\mathcal{E} = \Lambda e^{2M} > 1$.

Theorem

Let $\{\lambda_i\}_{i \geq 1}$ satisfy (3dM). Then for every $\varepsilon \in (0, 1)$ it holds

$$n(\varepsilon, d) = K \phi(q_\varepsilon/\sigma) \mathcal{E}^d e^{2q_\varepsilon \sqrt{d}} d^{-1/2} (1 + o(1)), \quad d \rightarrow \infty,$$

where $\phi(\cdot)$ is the standard normal density, the constant K is known and depends on whether U_ℓ has a lattice distribution or not, and the quantile $q = q_\varepsilon$ is chosen from the equation

$$1 - \Phi(q/\sigma) = \varepsilon^2.$$

Curse of dimensionality: $\log n(\varepsilon, d) = d \log \mathcal{E} (1 + o(1)), \quad d \rightarrow \infty$.

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First step: sum of units

$\zeta := \zeta(\varepsilon, d)$: the maximal positive number s.t.:

$$\sum_{\mathbf{k} \in \mathbb{N}^d: \lambda_{\mathbf{k}} < \zeta} \lambda_{\mathbf{k}}^2 = \mathbb{E} \|X - X_n\|^2 \leq \varepsilon^2 \Lambda^d.$$

$$A = A(\varepsilon, d) := \left\{ \mathbf{k} \in \mathbb{N}^d : \lambda_{\mathbf{k}} \geq \zeta \right\} = \left\{ \mathbf{k} \in \mathbb{N}^d : \prod_{\ell=1}^d \lambda_{k_\ell} \geq \zeta \right\}.$$

First step: sum of units. (Cont.)

Remember: $\lambda_i^2 = \Lambda \mathbb{P}(U_\ell = -\log \lambda_i)$, $i = 1, 2, \dots$

$\lambda_{\mathbf{k}} > 0$ for any $\mathbf{k} \in A$:

$$\begin{aligned} n(\varepsilon, d) &= \text{card}(A) = \sum_{\mathbf{k} \in A} \frac{\lambda_{\mathbf{k}}^2}{\lambda_{\mathbf{k}}} \\ &= \sum_{\mathbf{k} \in \mathbb{N}^d: \sum U_\ell \leq -\log \zeta} \exp \left\{ -2 \sum_{\ell=1}^d \log \lambda_{k_\ell} \right\} \Lambda^d \prod_{\ell=1}^d \mathbb{P}(U_\ell = -\log \lambda_{k_\ell}) \\ &= \Lambda^d \mathbb{E} \exp \left\{ 2 \sum_{\ell=1}^d U_\ell \right\} \mathbb{I}_{\{\sum_{\ell=1}^d U_\ell \leq -\log \zeta\}}. \end{aligned}$$

Second step: quantile convergence

For any $d \in \mathbb{N}$, $z \in \mathbb{R}^1$ and fixed $\varepsilon \in (0, 1)$:

$$\begin{aligned} \sum_{\mathbf{k} \in \mathbb{N}^d: \lambda_{\mathbf{k}} < z} \lambda_{\mathbf{k}}^2 &= \Lambda^d \mathbb{P} \left(\sum_{l=1}^d U_l > -\log z \right) \\ &= \Lambda^d \mathbb{P} \left(Z_d > -\frac{\log z + dM}{\sigma\sqrt{d}} \right) = \Lambda^d \mathbb{P} (Z_d > \theta_z) \leq \varepsilon^2 \Lambda^d, \end{aligned}$$

where $\theta_z = -\frac{\log z + dM}{\sigma\sqrt{d}}$, and $Z_d := \frac{\sum_{\ell=1}^d U_{\ell} - dM}{\sigma\sqrt{d}}$.

Then $\theta = \theta(\varepsilon, d)$ is the $(1 - \varepsilon^2)$ -quantile of the d.f. of Z_d .
By the CLT: $\theta(\varepsilon, d) \rightarrow q_{\varepsilon}/\sigma$, $d \rightarrow \infty$, where $q = q_{\varepsilon}$ is the quantile of the normal d.f.

Thus, for the information complexity we obtained:

$$\begin{aligned}n(\varepsilon, d) &= \mathcal{E}^d \mathbb{E} \exp\{2\sigma\sqrt{d}Z_d\} \mathbb{I}_{\{Z_d \leq \theta\}} \\ &= \mathcal{E}^d \exp\{2\sigma\sqrt{d}\theta\} \int_{-\infty}^{\theta} \exp\{2\sigma\sqrt{d}(z - \theta)\} dF_d(z),\end{aligned}$$

where $F_d(z) = \mathbb{P}(Z_d < z)$, $\mathcal{E} := \Lambda e^{2M}$,

$$Z_d := \frac{\sum_{\ell=1}^d U_{\ell} - dM}{\sigma\sqrt{d}},$$

$$\theta = \theta(\varepsilon, d) := -\frac{\log \zeta + dM}{\sigma\sqrt{d}}.$$

Third step: application of the Edgeworth-type expansions

- ▶ The rest of the proof:
 - integrate by parts
 - apply the Cramér-Esseen Theorems:

if the distribution of U_ℓ is not lattice, it holds (uniformly in z):

$$F_d(z) = \Phi(z) - \frac{e^{-z^2/2}}{\sqrt{2\pi}} \frac{\alpha^3(z^2 - 1)}{6\sigma^3\sqrt{d}} + o\left(\frac{1}{\sqrt{d}}\right)$$

and with the additional periodic term otherwise:




$$\frac{e^{-z^2/2} h S\left(\frac{(z\sigma\sqrt{d} - da)/h}{\sigma\sqrt{2\pi d}}\right)}{\sigma\sqrt{2\pi d}},$$

where $S(x) := [x] - x + \frac{1}{2}$, a is a shift and h is the maximal span of the distribution of $U_\ell - M$.





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