

Approximation de complexité pour des champs additifs

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Introduction

- ▶ Calcul de complexité pour des problèmes multivariés
- ▶ Idée: Etudier la tractabilité

$$f(t), t \in T \subset \mathbb{R}^d$$
$$\|f - A(f)\|$$

A algorithm, $\|\cdot\|$ norme

- ▶ H. Wozniakowski

Cadre général

- ▶ Champ aléatoire

$$X(t) = \sum_{k=1}^{\infty} \xi_k \phi_k(t), \quad t \in T \subset \mathbb{R}^d,$$

ξ_k v.a. et $\phi_k : T \rightarrow \mathbb{R}$

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- ▶ Norme $\|\cdot\|$
- ▶ Approximation en moyenne

$$n^{avg}(\varepsilon) = \inf\{n; E\|X - X_n\|^2 \leq \varepsilon^2\}, \quad \varepsilon \rightarrow 0$$

Champs additifs

► Exemple

$$X(t) = \sum_{l=1}^d X_l(t_l), t \in \mathbb{R}^d$$

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- ▶ Notre cadre: champ à d paramètres, somme de champs dépendant chacun de b paramètres

► $d, b \in \mathbb{N}^*$

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$$X(t) = \sum_{A \in D_b} X_A(\Pi_A(t)),$$

X_A copies i.i.d. d'un champ b -paramétrique

Produit de tenseur

- ▶ $\{Y(u), u \in [0, 1]\}$ processus de dim 1, ordre 2 t.q.

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$$Y^{\otimes b}(t) = \sum_{k \in \mathbb{N}^b} \prod_{l=1}^b \lambda(k_l) \phi_{k_l}(t_l) \xi_k, \quad t \in T_b,$$

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d – Champ additif

$$\begin{aligned} X_{d,b}(t) &= \sum_{A \in D_b} \sum_{k \in \mathbb{N}^b} \left(\prod_{l=1}^b \lambda(k_l) \phi_{k_l}([\Pi_A(t)]_l) \right) \xi_k^A \\ &= \sum_{A \in D_b} \sum_{k \in \mathbb{N}^A} \left(\prod_{a \in A} \lambda(k_a) \prod_{a \in A} \phi_{k_a}(t_a) \right) \xi_k^A. \end{aligned}$$

\mathcal{K}_Y covariance de Y : $\text{Cov}(Y(u), Y(u')) = \mathcal{K}_Y(u, u')$

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► **Lemma**

$k, k' \in \mathbb{N}^d$ et $A, A' \subset D$, alors $\psi(t) = \prod_{a \in A} \phi_{k_a}(t_a)$ et $\psi'(t) = \prod_{a \in A'} \phi_{k'_a}(t_a)$ sont identiques ou orthogonales dans $L_2(T_d)$.

On peut alors réécrire



$$X_{d,b}(t) = \sum_{h=0}^b \sum_{\substack{C \subset D \\ |C|=h}} \left[\sum_{k \in (\mathbb{N}^*)^C} \prod_{a \in C} \phi_{k_a}(t_a) \prod_{a \in C} \lambda(k_a) \right] \eta^C,$$

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▶ $(\eta^C)_{C \in D}$ centrées, non corrélées

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$$\text{Var}(\eta^C) = C_{d-h}^{b-h} \lambda(0)^{2(b-h)}.$$

Approximation en moyenne

Approximation de $X_{d,b}$ par X_n (n v.p. maximales)

$$n^{avg}(\varepsilon, d, b) = \inf\{n; E\|X_{d,b} - X_n\|_{L_2(\mathcal{T}_d)}^2 \leq \varepsilon^2\}.$$

- ▶ $\Lambda := \sum_{i=0}^{\infty} \lambda(i)^2 < \infty,$
 $\lambda(i) \sim \mu i^{-r} (\log i)^q$
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- ▶ $\alpha = q/r$

Valeurs propres

- ▶ $h \in \{1, \dots, b\}$ et $k \in (\mathbb{N}^*)^h$

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- $(\bar{\lambda}_{n,h}^2, n \in \mathbb{N})$ réarrangement décroissant

$$\bar{\lambda}_{n,h}^2 \sim B_h^2 n^{-2r} (\log n)^{2r\beta}, \quad n \rightarrow \infty, \quad (1)$$

where

- $\alpha > -1$:
$$\begin{cases} B_h = \mu^h \left(\frac{\Gamma(\alpha+1)^h}{\Gamma(h(\alpha+1))} \right)^r \\ \beta = (h-1) + h\alpha \end{cases}$$
- $\alpha < -1$:
$$\begin{cases} B_h = \mu h^r \left[\sum_{i \geq 1} \lambda(i)^{1/r} \right]^{(h-1)r} \\ \beta = \alpha \end{cases}$$

(Csaki; Li; Karol, Nikitin & Nazarov; Lifshits & Tulyakova)

Proposition

- $\alpha > -1$

$$n^{\text{avg}}(\varepsilon, d, b) \sim [C_d^b]^{\frac{2r}{2r-1}} \left(\frac{B_b}{\sqrt{2}(r-1/2)^{r\beta+1/2}} \frac{|\log \varepsilon|^{r\beta}}{\varepsilon} \right)^{(r-1/2)^{-1}}, \varepsilon \rightarrow 0$$

- $\alpha < -1$

$$n^{\text{avg}}(\varepsilon, d, b) \sim \left(\frac{\sqrt{Q}}{\sqrt{2}(r-1/2)^{r\alpha+1/2}} \frac{|\log \varepsilon|^{r\alpha}}{\varepsilon} \right)^{(r-1/2)^{-1}}, \varepsilon \rightarrow 0$$

$$Q = \left(\sum_{h=1}^b C(h)^{\frac{1}{2r}} \right)^{2r} \quad \text{and} \quad C(h) = C_{d-h}^{b-h} [C_d^h]^{2r} \lambda(0)^{2(b-h)} B_h^2.$$

Dimension croissante

$d \rightarrow \infty$



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▶ b fixé et $d \rightarrow \infty$

▶ $b, d \rightarrow \infty$ $b/d \rightarrow l$

b fixé

Proposition

$$\tilde{n}^{avg}(\varepsilon, b, d) \sim \frac{d^b}{b!} \Lambda^{-b/(2r-1)} \left(\frac{B_b}{\sqrt{2}(r-1/2)^{r\beta+1/2}} \frac{|\log \varepsilon|^{r\beta}}{\varepsilon} \right)^{(r-1/2)^{-1}} .$$

$b \rightarrow \infty$ et $b/d \rightarrow l$

Proposition

$b/d \rightarrow l \in [0, 1]$

$$\lim_{d \rightarrow \infty} \frac{\log \tilde{n}^{avg}(\varepsilon, b, d)}{d} = \log V,$$

où

$$V = (1 - lp)^{-1+lp} l^{-lp} (1 - p)^{(1-p)l} A^l,$$

$$\tilde{\Lambda} = \sum_{i=1}^{\infty} \lambda(i)^2, \quad p = \frac{\tilde{\Lambda}}{\Lambda},$$

$$M = \sum_{i=0}^{\infty} (-\log \lambda(i)) \frac{\lambda(i)^2}{\tilde{\Lambda}}, \quad A = e^{2M} \Lambda.$$

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