

Concentration of measure for U-statistics with applications

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The setup

- (Σ, \mathcal{F}) – a Polish space with the Borel σ -field
- X_1, X_2, \dots, X_n - i.i.d. Σ -valued random variables
- $h: \Sigma^d \rightarrow \mathbb{R}$ – a Borel measurable function
- $I_n^d = \{\mathbf{i} = (i_1, \dots, i_d) : i_j \in \mathbb{N}, 1 \leq i_j \leq n, i_j \neq i_k \text{ if } j \neq k\}$
- a U-statistic:

$$Z = \sum_{\mathbf{i} \in I_n^d} h(X_{i_1}, \dots, X_{i_d})$$

Additional assumptions

We assume

- **Symetry** – h invariant under permutation of coordinates
- **Complete degeneracy**

$$\mathbb{E}h(X_1, X_2, \dots, X_d) = 0.$$

A natural assumption in view of the Hoeffding decomposition.

Aim

Our aim is to find good (exponential) estimates on

$$\mathbb{P}(|Z| \geq t)$$

How to do it?

- Estimate moments $\|Z\|_p$
- Use Chebyshev's Inequality and optimize in p

Previously known results

Theorem (Bernstein's inequality, $d = 1$)

$$Z = \sum_{i=1}^n h(X_i)$$

$$\mathbb{P}(|Z| \geq t) \leq K \exp\left(-\frac{1}{K} \min\left[\frac{t^2}{n\mathbb{E}h^2}, \frac{t}{\|h\|_\infty}\right]\right)$$

Previously known results

Theorem (Giné, Latała, Zinn (Houdré, Reynaud-Bouret with a different proof), $d = 2$)

$$Z = \sum_{i \neq j} h(X_i, X_j)$$

$$\mathbb{P}(|Z| \geq t) \leq$$

$$K \exp \left(- \frac{1}{K} \min \left[\frac{t^2}{n^2 \mathbb{E} h^2}, \frac{t}{n \|h\|_{L^2 \rightarrow L^2}}, \frac{t^{2/3}}{(n \|\mathbb{E}_{X_1} h^2\|_{\infty})^{1/3}}, \frac{t^{1/2}}{\|h\|_{\infty}^{1/2}} \right] \right)$$

Previously known results

Actually Giné, Latała, Zinn prove

Theorem

$$\begin{aligned} \mathbb{E}|Z|^p &\leq K^p [p^{p/2}(n^2 \mathbb{E}h^2)^{p/2} + p^p(n\|h\|_{L^2 \rightarrow L^2})^p \\ &\quad + p^{3p/2} \mathbb{E}_X \max_{i \leq n} (n \mathbb{E}_Y h^2(X_i, Y))^p \\ &\quad + p^{2p} \mathbb{E} \max_{i, j \leq n} |h(X_i, Y_j)|^p], \end{aligned}$$

where (Y_i) - independent copy of (X_i) .

Notation and definitions

- I - a finite, nonempty set,
- \mathcal{P}_I - set of partitions of I into disjoint, nonempty sets
- $\mathcal{J} = \{J_1, \dots, J_k\} \in \mathcal{P}_I$
- For $I = \emptyset$, $\mathcal{P}_I = \{\emptyset\}$
- $\deg \mathcal{J} = \# \mathcal{J}$.

Notation and definitions

With each partition we associate a norm $\|h\|_{\mathcal{J}}$.
 It is best to define it by examples

$$\|h(X_1, X_2, X_3)\|_{\{1,2,3\}} = \sup\{\mathbb{E}h(X_1, X_2, X_3)f(X_1, X_2, X_3) : \mathbb{E}f(X_1, X_2, X_3)^2 \leq 1\}$$

$$\|h(X_1, X_2, X_3)\|_{\{1,2\}\{3\}} = \sup\{\mathbb{E}h(X_1, X_2, X_3)f(X_1, X_2)g(X_3) : \mathbb{E}f(X_1, X_2)^2, \mathbb{E}g(X_3)^2 \leq 1\}$$

$$\|h(X_1, X_2, X_3)\|_{\{1\}\{2\}\{3\}} = \sup\{\mathbb{E}h(X_1, X_2, X_3)f(X_1)g(X_2)k(X_3) : \mathbb{E}f(X_1)^2, \mathbb{E}g(X_2)^2, \mathbb{E}k(X_3)^2 \leq 1\}.$$

Notation and definitions

We can see that

- $\|h\|_{\{1,2,3\}} = \|h\|_2$
- $\|h\|_{\{1\},\{2\},\{3\}}$ is the norm of h viewed as a 3-linear functional on $L^2(X_1) \times L^2(X_2) \times L^2(X_3)$.
- $\|h\|_{\{1,2\},\{3\}}$ is the norm of h as a 2-linear functional on $L^2(X_1, X_2) \times L^2(X_3)$.

Notation and definitions

Similarly we define e.g.

$$\|h(X_1, X_2, X_3)\|_{\{1\}\{2\}} = \sup\{\mathbb{E}_{X_1, X_2} h(X_1, X_2, X_3) f(X_1) g(X_2) : \mathbb{E}f^2, \mathbb{E}g^2 \leq 1\}$$

$$\|h(X_1, X_2, X_3)\|_{\{3\}} = \sup\{\mathbb{E}_{X_3} h(X_1, X_2, X_3) f(X_3) : \mathbb{E}f^2 \leq 1\},$$

but now they are **random variables**.

Finally (to simplify the statements of theorems) we define

$$\|h(X_1, X_2, X_3)\|_{\emptyset} = |h(X_1, X_2, X_3)|.$$

Moments estimates

$$Z = \sum_{i \neq j \neq k} h(X_i, X_j, X_k).$$

Theorem

For all $p \geq 2$ we have

$$\begin{aligned} & \mathbb{E}|Z|^p \\ & \leq K^p \sum_{I \subseteq \{1,2,3\}} \sum_{\mathcal{J} \in \mathcal{P}_I} n^{\#I p/2} p^{p(\deg(\mathcal{J})/2 + \#\mathcal{I}^c)} \mathbb{E}_{I^c} \max_{i \in \mathcal{I}^c} \|h(X_i, Y_j, Z_k)\|_{\mathcal{J}}^p, \end{aligned}$$

where $(Y_j), (Z_k)$ – independent copies of (X_i) , $i = (i, j, k)$.

A closer look at the right-hand side

- $I = \{1, 2, 3\}$
 - $p^{p/2} n^{3p/2} \|h\|_{\{1,2,3\}}^p \sim p^{p/2} (\mathbb{E}|Z|^2)^{p/2}$
 - $p^p n^{3p/2} \|h\|_{\{1,2\},\{3\}}^p$,
 - $p^{3p/2} n^{3p/2} \|h\|_{\{1\}\{2\}\{3\}}^p$,
- $I = \{1, 2\}$
 - $p^{3p/2} n^p \mathbb{E}_Y \max_{k \leq n} \|h(X_1, X_2, Y_k)\|_{\{1,2\}}^p =$
 $p^{3p/2} n^p \mathbb{E}_Y \max_{k \leq n} (\mathbb{E}_{X_1, X_2} h(X_1, X_2, Y_k)^2)^{p/2}$
 - $p^{2p} n^p \mathbb{E}_Y \max_{k \leq n} \|h(X_1, X_2, Y_k)\|_{\{1\}\{2\}}^p$

A closer look at the right-hand side

- $I = \{1\}$
 - $p^{5p/2} n^{p/2} \mathbb{E}_{Y,Z} \max_{j,k \leq n} \|h(X_1, Y_j, Z_k)\|_{\{1\}}^p =$
 $p^{5p/2} n^{p/2} \mathbb{E}_{Y,Z} \max_{j,k \leq n} (\mathbb{E}_{X_1} h^2(X_1, Y_j, Z_k)^2)^{p/2}$
- $I = \emptyset,$
 - $p^{3p} \mathbb{E} \max_{i,j,k \leq n} |h(X_i, Y_j, Z_k)|^p.$

Tail estimates

Theorem

$$\mathbb{P} \left(\left| \sum_i h_i \right| \geq t \right) \leq K \exp \left[-\frac{1}{K} \min_{I \subseteq I_d, \mathcal{J} \in \mathcal{P}_I} \left(\frac{t}{n^{\#I/2} \| \| h \|_{\mathcal{J}} \|_{\infty}} \right)^{2/(\deg(\mathcal{J}) + 2\#I^c)} \right].$$

Gaussian chaoses

Let (a_{ijk}) be a 3-indexed symmetric matrix with zeros on the diagonal, and g_1, g_2, \dots – independent $\mathcal{N}(0, 1)$ Gaussian variables. Consider

$$Z = \sum_{ijk} a_{ijk} g_i g_j g_k.$$

Define for $\mathcal{J} = \{J_1, \dots, J_m\} \in \mathcal{P}_{\{1,2,3\}}$

$$\|(a_{ijk})\|_{\mathcal{J}} = \sup \left\{ \sum_{ijk} a_{ijk} \prod_{l=1}^m x_{i_{J_l}}^{(l)} : \sum_{i_{J_l}} (x_{i_{J_l}}^{(l)})^2 \leq 1 \right\}$$

e.g.

$$\|(a_{ijk})\|_{\{1,2\}\{3\}} = \sup \left\{ \sum_{ijk} a_{ijk} x_{ij} y_k : \sum x_{ij}^2 \leq 1, \sum y_k^2 \leq 1 \right\}$$

Gaussian chaoses

Theorem (Latała)

For $p \geq 2$

$$\|Z\|_p \sim \sum_{\mathcal{J} \in \mathcal{P}_{\{1,2,3\}}} p^{\deg \mathcal{J}/2} \|(\mathbf{a}_{ijk})\|_{\mathcal{J}}.$$

In consequence for $t \geq 0$

$$\mathbb{P}(|Z| \geq t) \leq K \exp \left[-\frac{1}{K} \min_{\mathcal{J} \in \mathcal{P}_{\{1,2,3\}}} \left(\frac{t}{\|(\mathbf{a}_{ijk})\|_{\mathcal{J}}} \right)^{2/\deg \mathcal{J}} \right].$$

Tools

- Decoupling Inequalities (de la Peña, Montgomery-Smith)
- Talagrand's Inequality for suprema of empirical processes (in the moments version, Giné, Latała, Zinn & Boucheron, Bousquet, Lugosi, Massart)
- Estimates between weak and strong variance for empirical processes
- Estimates on Gaussian averages in operator spaces (Latała)

Crucial Lemma

Lemma

Let Z_k be independent random variables and $A_k(Z_k) = (a_{ijk}(Z_k))_{ij}$ - independent centered random matrices. Then for $p \geq 2$

$$\begin{aligned} \mathbb{E} \left\| \sum_k A_k(Z_k) \right\| &\leq K \left[\frac{1}{\sqrt{p}} \|(a_{ijk}(Z_k))\|_{\{1,2,3\}} \right. \\ &\quad + \|(a_{ijk}(Z_k))\|_{\{1,3\}\{2\}} + \|(a_{ijk}(Z_k))\|_{\{1\}\{2,3\}} \\ &\quad + \sqrt{p} \|(a_{ijk}(Z_k))\|_{\{1\}\{2\}\{3\}} \\ &\quad \left. + p \sqrt{\mathbb{E} \max_k \|(A_k(Z_k))\|^2} \right] \end{aligned}$$

Crucial Lemma

$$\|(\mathbf{a}_{ijk}(Z_k))\|_{\{1,2,3\}} = \sqrt{\mathbb{E} \sum_{ijk} \mathbf{a}_{ijk}(Z_k)^2}$$

$$\|(\mathbf{a}_{ijk}(Z_k))\|_{\{1,3\},\{2\}} = \sqrt{\sup_{\|x\|_2 \leq 1} \sum_{i,k} \mathbb{E} \left(\sum_j \mathbf{a}_{ijk}(Z_k) x_j \right)^2}$$

$$\|(\mathbf{a}_{ijk}(Z_k))\|_{\{1\}\{2\}\{3\}} = \sqrt{\sup_{\|x\|_2, \|y\|_2 \leq 1} \sum_k \mathbb{E} \left(\sum_{i,j} \mathbf{a}_{ijk}(Z_k) x_i y_j \right)^2.}$$

- Tail estimates for multiple stochastic integrals with respect to processes with independent increments and uniformly bounded jumps (in the spirit of inequalities by Houdré, Reynaud-Bouret for $d = 2$)
- Law of the iterated logarithm for kernel density estimators (Giné, Mason), $d = 2$.
- Law of the iterated logarithm for canonical U-statistics

Definition

For $u \geq 0$ let us define

$$\|h(X_1, X_2, X_3)\|_{\{1,2,3\},u} = \sup\{\mathbb{E}h(X_1, X_2, X_3)f(X_1, X_2, X_3) : \|f\|_2 \leq 1, \|f\|_\infty \leq u\}$$

$$\|h(X_1, X_2, X_3)\|_{\{1,2\}\{3\},u} = \sup\{\mathbb{E}h(X_1, X_2, X_3)f(X_1, X_2)g(X_3) : \|f\|_2, \|g\|_2 \leq 1, \|f\|_\infty, \|g\|_\infty \leq u\}$$

Theorem (Latała, R.A. (Giné, Kwapień, Latała, Zinn for $d = 2$))

The $h: \Sigma^d \rightarrow \mathbb{R}$ be arbitrary kernel. Then the LIL

$$\limsup_{n \rightarrow \infty} \frac{1}{n^{d/2} \log \log^{d/2} n} \left| \sum_{i \in I_n^d} h(X_{i_1}, \dots, X_{i_d}) \right| < \infty$$

holds if and only if h is completely degenerated and for all $\mathcal{J} \in \mathcal{P}_{\{1, \dots, d\}}$ we have

$$\limsup_{u \rightarrow \infty} \frac{\|h\|_{\mathcal{J}, u}}{\log \log^{(d - \deg \mathcal{J})/2} u} < \infty$$

A few questions

- Prove estimates for suprema of U-statistics (U-processes) at least over VC classes of kernels or for U-statistics in Banach spaces of type 2 (known for Hilbert spaces).
- Identify the limit in the LIL for $d \geq 2$
- Prove tail estimates for chaoses generated by other variables (e.g. stable \rightarrow consequences in stochastic processes, Bernoulli \rightarrow consequences in random graphs theory)