

Higher Order Concentration of Measure

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joint work with:

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Concentration of measure and its applications

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Topics

Higher Order Concentration Bounds in High Dimensions:

- Euclidean Spaces: Logarithmic Sobolev and Poincaré inequalities
- Spheres
- Functions of independent random variables
- Functions of weakly dependent random variables

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Higher Order Concentration of Measure

Recall: Logarithmic Sobolev inequality

μ sat. LSI(σ^2) if $\forall f$ bounded, locally Lipschitz

$$\text{Ent}_\mu(f^2) := \int f^2 \log\left(\frac{f^2}{\int f^2 d\mu}\right) d\mu \leq 2\sigma^2 \int |\nabla f|^2 d\mu$$

$$\Rightarrow \mu(|f - \int f d\mu| \geq t) \leq 2e^{-c\sigma^2 t^2 / \|\nabla f\|_\infty^2}$$

Example: $\mu := \otimes_{i=1}^n \mu_i$, μ_i unif. distr. on $[-1, 1]$ μ sat. LSI($\Theta(1)$)

$$f(x) := \sum_{i < j} x_i x_j \rightsquigarrow \sup |\nabla f(x)| = \Theta(n^{3/2})$$

$$\Rightarrow \mu(|f|/n^{3/2} \geq t) \leq 2e^{-ct^2} \text{ wrong order!}$$

Note that $|f''(x)|_{\text{HS}} \leq n$ \rightsquigarrow Ideas:

- Use higher order derivatives
- study fluctuations of $f - \mathbb{E}f - f_1 - \dots - f_{d-1}$ w.r.t. suitable decomp.

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Higher Order Derivatives in Concentration of Measure

Previous work:

- First Order Concentration: Sudakov, Milman, Schechtman, Talagrand, Ledoux
- Higher Order Derivatives: Adamczak–Wolff (2015)
- Second Order Concentration on the Sphere: Bobkov–Chistyakov–Götze (2017)
- Higher Order Poincaré and Superconcentration: Chatterjee (2007)
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Higher Order Concentration on Euclidean Spaces

Notation: $G \subset \mathbb{R}^n$ open set, $d \in \mathbb{N}$, $f \in \mathcal{C}^d(G)$.

hypermatrix of d -fold partial derivatives $f^{(d)}$

$$f_{i_1 \dots i_d}^{(d)}(x) = \partial_{i_1 \dots i_d} f(x).$$

Use operator and Hilbert–Schmidt type norms

$$|f^{(d)}|_{\text{Op}} := \sup \left\{ f^{(d)}(x)[v_1, \dots, v_d] : |v_1| = \dots = |v_d| = 1 \right\},$$

$$|f^{(d)}|_{\text{HS}} := \left(\sum_{i_1, \dots, i_d} (\partial_{i_1 \dots i_d} f(x))^2 \right)^{1/2}.$$

For a prob. measure μ on G ,

$$\|f^{(d)}\|_{\text{Op}, p} \equiv \| |f^{(d)}|_{\text{Op}} \|_p \equiv \left(\int_G |f^{(d)}|_{\text{Op}}^p d\mu \right)^{1/p},$$

any $p \in (0, \infty]$. Similarly: $\|f^{(d)}\|_{\text{HS}, p}$.

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Theorem (Bobkov-Götze-S (2017))

Let μ with $LSI(\sigma^2)$, $d \in \mathbb{N}$. $f: G \rightarrow \mathbb{R}$, $f \in \mathcal{C}^d(G)$, $\int f d\mu = 0$. Assume

$$\begin{aligned}\|f^{(k)}\|_{Op,2} &\leq \min(1, \sigma^{d-k}) \quad \forall k = 1, \dots, d-1, \\ \|f^{(d)}\|_{Op,\infty} &\leq 1.\end{aligned}\tag{0.1}$$

Then, for some $c > 0$

$$\int_G \exp\left(\frac{c}{\sigma^2} |f|^{2/d}\right) d\mu \leq 2.$$

By Chebychev's inequality: $\mu(|f| \geq t) \leq 2e^{-c\sigma^2 t^{2/d}}$

may somewhat sharpen this bound by analyzing the proof

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Euclidean Spaces: LSI

Simplify conditions: Assume

$$\int_G \partial_{i_1 \dots i_k} f \, d\mu = 0 \quad \forall k = 1, \dots, d-1, \quad \forall 1 \leq i_1, \dots, i_k \leq n \quad (0.2)$$

\rightsquigarrow (0.1) can be replaced by $\|f^{(d)}\|_{HS,2} \leq 1$.

Idea: remove “lower order terms” by suitable “projections”, e.g. ($d = 2$)

$$\tilde{f}(x) := f(x) - \left(\mu(f) + \sum_{i=1}^n \mu(\partial_i f)(x_i - \mu(x_i)) \right).$$

Gaussian case: (0.2) \Leftrightarrow orthogonality to all polynomials of degree $\leq d-1$

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Step 1: For any $g: G \rightarrow \mathbb{R}$ locally Lipschitz s. th. $\|g\|_p < \infty$

$$\|g\|_p^2 - \|g\|_2^2 \leq \sigma^2(p-2)\|\nabla g\|_p^2, \quad p \geq 2.$$

Proof: (Aida-Stroock (1994)) Assume g bounded. Then,

$$\frac{d}{dp}\|g\|_p^2 = \frac{2}{p^2}\|g\|_p^{2-p} \text{Ent}_\mu(|g|^p), \quad p > 2.$$

Set $u = |g|^{p/2}$ with $|\nabla u|^2 \leq \frac{p^2}{4}|g|^{p-2}|\nabla g|^2$. Hence, by Hölder's inequality,

$$\int_G |\nabla u|^2 d\mu \leq \frac{p^2}{4}\|g\|_p^{p-2}\|\nabla g\|_p^2.$$

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$$\text{Ent}_\mu(|g|^p) = \text{Ent}_\mu(u^2) \leq 2\sigma^2 \int_G |\nabla u|^2 d\mu \leq \frac{p^2\sigma^2}{2}\|g\|_p^{p-2}\|\nabla g\|_p^2.$$

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Sketch of Proof II

Step 2: For $f: G \rightarrow \mathbb{R}$ C^d -function

$$|\nabla|f^{(k-1)}(x)|_{\text{Op}}| \leq |f^{(k)}(x)|_{\text{Op}}, \quad \text{all } x \in G.$$

Follows by Taylor expansion.

Step 3: Apply Step 1 to $g := |f^{(k-1)}|_{\text{Op}}$ and combine with Step 2:

$$\|f^{(k-1)}\|_{\text{Op},p}^2 - \|f^{(d-1)}\|_{\text{Op},2}^2 \leq \sigma^2(p-2) \|\nabla|f^{(d-1)}|_{\text{Op}}\|_p^2 \leq \sigma^2(p-2) \|f^{(d)}\|_{\text{Op},p}^2.$$

By iteration and the Poincaré inequality,

$$\begin{aligned} \|f\|_p^2 &\leq \|f\|_2^2 + \sum_{k=1}^{d-1} (\sigma^2(p-2))^k \|f^{(k)}\|_{\text{Op},2}^2 + (\sigma^2(p-2))^d \|f^{(d)}\|_{\text{Op},p}^2 \\ &\leq \sum_{k=1}^{d-1} (\sigma^2 p)^k \|f^{(k)}\|_{\text{Op},2}^2 + (\sigma^2 p)^d \|f^{(d)}\|_{\text{Op},p}^2. \end{aligned}$$

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Step 3: Apply Step 1 to $g := |f^{(k-1)}|_{\text{Op}}$ and combine with Step 2:

$$\|f^{(k-1)}\|_{\text{Op},p}^2 - \|f^{(d-1)}\|_{\text{Op},2}^2 \leq \sigma^2(p-2) \|\nabla|f^{(d-1)}|_{\text{Op}}\|_p^2 \leq \sigma^2(p-2) \|f^{(d)}\|_{\text{Op},p}^2.$$

By iteration and the Poincaré inequality,

$$\begin{aligned} \|f\|_p^2 &\leq \|f\|_2^2 + \sum_{k=1}^{d-1} (\sigma^2(p-2))^k \|f^{(k)}\|_{\text{Op},2}^2 + (\sigma^2(p-2))^d \|f^{(d)}\|_{\text{Op},p}^2 \\ &\leq \sum_{k=1}^{d-1} (\sigma^2 p)^k \|f^{(k)}\|_{\text{Op},2}^2 + (\sigma^2 p)^d \|f^{(d)}\|_{\text{Op},p}^2. \end{aligned}$$

Sketch of Proof II

Step 2: For $f: G \rightarrow \mathbb{R}$ C^d -function

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Use $\|f^{(k)}\|_{\text{Op},2}^2 \leq \min(1, \sigma^{2(d-k)})$ for all $k = 1, \dots, d-1$ and $\|f^{(d)}\|_{\text{Op},\infty} \leq 1$ to get

$$\|f\|_p^2 \leq \sigma^{2d} \sum_{k=1}^d p^k \leq \frac{1}{1-p^{-1}} (\sigma^2 p)^d \leq 2(\sigma^2 p)^d,$$

i.e. $\|f\|_p \leq (2\sigma^2 p)^{d/2}$ for all $p \geq 2$. If $p < 2$, by Hölder's inequality

$$\|f\|_p \leq \|f\|_2 \leq (4\sigma^2)^{d/2}.$$

Consider $p = 2k/d$, $k = 1, 2, \dots$, and take $2/d$ -th root:

$$\||f|^{2/d}\|_k \leq \gamma k, \quad \gamma = 4\sigma^2.$$

It follows

$$\int \exp\left(\frac{c}{\sigma^2}|f|^{2/d}\right) d\mu \leq 2,$$

$c > 0$ constant (possible choice: $c = 1/(8e)$).

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Applications: Random Matrices

- Symmetric random matrices $\Xi = (\xi_{jk}/\sqrt{N})_{1 \leq j, k \leq N}$, where $\xi_{jk}, j \leq k$, independent with $\text{LSI}(\sigma^2)$
 $\rightsquigarrow \mu = \text{distr. of eigenvalues } \lambda = (\lambda_1 \leq \dots \leq \lambda_N) \text{ sat. LSI}(2\sigma^2/N)$
- β -ensembles: density on $\{\lambda \in \mathbb{R}^N : \lambda_1 < \dots < \lambda_N\}$

$$\mu(d\lambda) := \frac{1}{Z_N} e^{-\beta N \mathcal{H}(\lambda)} d\lambda, \quad \mathcal{H}(\lambda) := \frac{1}{2} \sum_{k=1}^N V(\lambda_k) - \frac{1}{N} \sum_{1 \leq k < l \leq N} \log |\lambda_l - \lambda_k|.$$

Here, $V: \mathbb{R} \rightarrow \mathbb{R}$ is a smooth strictly convex potential. Then

$$\mathcal{H}''(\lambda) \geq c \text{Id}$$

uniformly in λ and μ sat. $\text{LSI}(1/(\beta c N))$ (Bakry–Emery).

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$f: \mathbb{R} \rightarrow \mathbb{R}$, C^1 -smooth. Self-normalizing sums

$$S_N := \sum_{j=1}^N (f(\lambda_j) - \mathbb{E}f(\lambda_j)) \Rightarrow \mathcal{N}(0, \sigma_f^2)$$

“linear eigenvalue statistics” (cf. Johansson (1998), Pastur et al. (1996), Guionnet–Zeitouni (2000))

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Idea: Study quadratic eigenvalue statistics, i. e. $T_N := \sum_{j \neq k} g(\lambda_j, \lambda_k)$ for $g: \mathbb{R}^2 \rightarrow \mathbb{R}$ smooth. “Recenter” according to (0.2) \rightsquigarrow

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Higher Order Concentration on Spheres

$S^{n-1} = \{x \in \mathbb{R}^n : |x| = 1\}, \quad n \geq 2, \quad \sigma_{n-1}$ uniform measure

- σ_{n-1} satisfies LSI with constant $1/(n-1)$
- Any $\mathcal{C}^d(S^{n-1})$ -smooth fct. can be extended to \mathcal{C}^d -smooth fct. on $\mathbb{R}^n \setminus \{0\}$

Theorem (Bobkov-Götze-S (2017))

f \mathcal{C}^d -smooth in open nbhd. of S^{n-1} , $\int_{S^{n-1}} f d\sigma_{n-1} = 0$. If

$$\|f^{(k)}\|_{\text{op},2} \leq n^{-(d-k)/2} \quad \forall k = 1, \dots, d-1,$$

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$$Q_{d,a}(\theta) := \sum_{i=1}^n a_i \theta_i^d, \quad \theta \in S^{n-1},$$

$d \geq 3$, $a \in \mathbb{R}^n$ s. th. $n^{-1} \sum_{i=1}^n a_i^2 = 1$.

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In part., $Q_{d,a} - \bar{Q}_{d,a} = O_{\sigma_{n-1}}(n^{-(d-1)/2})$.

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$$h_n(\theta) := \mathbb{E} H\left(\sum_{i=1}^n \theta_i X_i\right),$$

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Euclidean Spaces: Poincaré inequality

Recall: Poincaré inequality

μ sat. PI(σ^2) if $\forall f$ bounded, locally Lipschitz

$$\text{Var}_\mu(f) := \int f^2 d\mu - \left(\int f d\mu \right)^2 \leq \sigma^2 \int |\nabla f|^2 d\mu$$

$$\text{LSI}(\sigma^2) \Rightarrow \text{PI}(\sigma^2), \quad \text{PI}(\sigma^2) \Rightarrow \mu(|f - \int f d\mu| \geq t) \leq 2e^{-c\sigma t/\|\nabla f\|_\infty}$$

Theorem (Götze-S (2018))

Let μ with PI(σ^2), $d \in \mathbb{N}$. $f: G \rightarrow \mathbb{R}$, $f \in \mathcal{C}^d(G)$, $\int f d\mu = 0$. Assume

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μ sat. PI(σ^2) if $\forall f$ bounded, locally Lipschitz

$$\text{Var}_\mu(f) := \int f^2 d\mu - \left(\int f d\mu \right)^2 \leq \sigma^2 \int |\nabla f|^2 d\mu$$

$$\text{LSI}(\sigma^2) \Rightarrow \text{PI}(\sigma^2), \quad \text{PI}(\sigma^2) \Rightarrow \mu(|f - \int f d\mu| \geq t) \leq 2e^{-c\sigma t/\|\nabla f\|_\infty}$$

Theorem (Götze-S (2018))

Let μ with PI(σ^2), $d \in \mathbb{N}$. $f: G \rightarrow \mathbb{R}$, $f \in \mathcal{C}^d(G)$, $\int f d\mu = 0$. Assume

$$\|f^{(k)}\|_{\text{Op},2} \leq \min(1, \sigma^{d-k}) \quad \forall k = 1, \dots, d-1,$$

$$\|f^{(d)}\|_{\text{Op},\infty} \leq 1.$$

Then, for some $c > 0$

$$\int_G \exp\left(\frac{c}{\sigma} |f|^{1/d}\right) d\mu \leq 2.$$

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Sketch of Proof

Lemma: For any $g: G \rightarrow \mathbb{R}$ locally Lipschitz s.th. $\|g\|_p < \infty$

$$\|g\|_p - \|g\|_2 \leq \frac{\sigma p}{\sqrt{2}} \|\nabla g\|_p, \quad p \geq 2.$$

Proof: Assume g is \mathcal{C}^1 -smooth. Apply Poincaré inequality for $\mu \otimes \mu$ (tensorization!) to fct.

$$u(x, y) := |g(x) - g(y)|^{p/2} \text{sign}(g(x) - g(y))$$

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Lemma: For any $g: G \rightarrow \mathbb{R}$ locally Lipschitz s.th. $\|g\|_p < \infty$

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Functions of independent random variables

Use difference operators instead of derivatives!

$$\mathfrak{h}_i f(X) = \frac{1}{2} \|f(X) - T_i f(X)\|_{i,\infty}, \quad \mathfrak{h} f = (\mathfrak{h}_1 f, \dots, \mathfrak{h}_n f),$$

$T_i f(X) = f(X_1, \dots, X_{i-1}, \bar{X}_i, X_{i+1}, \dots, X_n)$, $\bar{X}_1, \dots, \bar{X}_n$ indep. copies

$$\mathfrak{h}_{i_1 \dots i_d} f(X) = \frac{1}{2^d} \left\| \prod_{s=1}^d (Id - T_{i_s}) f(X) \right\|_{i_1, \dots, i_d, \infty},$$

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$|\mathfrak{h}^{(d)} f|_{HS}$, $\|\mathfrak{h}^{(d)} f\|_{HS,p}$ as above

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Functions of independent random variables

Theorem (Bobkov-Götze-S 2017)

X_1, \dots, X_n indep. r.v., $f = f(X) \in L^\infty(\mathbb{P})$, $\mathbb{E}f = 0$. Assume

$$\|\mathfrak{h}^{(k)}f\|_{\text{HS},2} \leq 1 \quad (k = 1, \dots, d-1) \quad (0.3)$$

$$\|\mathfrak{h}^{(d)}f\|_{\text{HS},\infty} \leq 1 \quad (0.4)$$

Then exists constant $c > 0$

$$\mathbb{E} \exp \left(c |f|^{2/d} \right) \leq 2.$$

Remarks:

- If f is multilinear polynomial $f(X) = \sum_{|I|=d} \alpha_I X_I + \dots$ with X_1, \dots, X_n standardized bounded, then (0.4) \Rightarrow (0.3).
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Sketch of Proof

Step 1: Introduce difference operators

$$\mathfrak{h}_i^\pm f(X) = \frac{1}{2} \|(f(X) - T_i f(X))_\pm\|_{i,\infty}$$

Boucheron / Bousquet / Lugosi / Massart (2005): For any real $p \geq 2$,

$$\|(f - \mathbb{E}f)_+\|_p \leq \sqrt{8\kappa p} \|\mathfrak{h}^+ f\|_p, \quad \|(f - \mathbb{E}f)_-\|_p \leq \sqrt{8\kappa p} \|\mathfrak{h}^- f\|_p$$

with $\kappa < 1.271$ absolute constant.

Consequence: By Hölder's inequality and triangular ineq.,

$$\|f\|_p \leq \|f\|_2 + \sqrt{32\kappa p} \|\mathfrak{h} f\|_p, \quad p \geq 2.$$

Step 2: For any $d \geq 2$,

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Then proceed in a similar way as in Euclidean case



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U-Statistics

Application: *U*-statistics. Combine elements of the proof of Theorem [B-G-S (2017)] for independent Rademacher variables and results on randomized *U*-statistics by de la Peña / Giné:

Corollary

X_1, \dots, X_n i.i.d. r.v., values in (S, \mathcal{S}) , $d \in \mathbb{N}$, $h: S^d \rightarrow \mathbb{R}$ bounded by $M > 0$.

Assume h is completely degenerate, i.e. $\mathbb{E}_i h(X_1, \dots, X_d) = 0$, $i = 1, \dots, d$.

Let

$$f(X_1, \dots, X_n) := \frac{(n-d)!}{n!} \sum_{h \neq \dots \neq l_d} h(X_{i_1}, \dots, X_{i_d}).$$

Then $\exists c = c(d, M) > 0$ constant s.th.

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Additive Functionals of Partial Sums

Application: Additive Functionals of Partial Sums (e.g. random walks), i.e.

$$S_f(X) := \sum_{i=1}^n f\left(\sum_{j=1}^i X_j\right).$$

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X_1, \dots, X_n indep. r.v., $f: \mathbb{R} \rightarrow \mathbb{R}$ bounded measurable. Then, for any $t \geq 0$,

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Proof: calculating out difference operators & counting arguments.

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Functions of weakly dependent random variables

Dependent r.v.: define L^2 - and L^∞ -difference operators \mathfrak{d} , \mathfrak{h} using disintegration theorem for product spaces

Consider $X = (X_1, \dots, X_n)$ dependent r.v., $X \sim \mu$

Theorem (Götze-S-Sinulis (2018))

Let μ with $LSI_0(\sigma^2)$, $d \in \mathbb{N}$, $f \in L^\infty(\mu)$. Assume

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Ising model

Example: Ising model q^n p.m. on $\{\pm 1\}^n$ def. by normalizing

$\pi(\sigma) = \exp(\frac{1}{2} \sum_{i,j} J_{ij} \sigma_i \sigma_j)$, where $J = (J_{ij})$ s.th. $J_{ii} = 0$ and

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To establish LSI: approximate tensorization result (Marton)

$$\text{Ent}_{q^n}(f) \leq \frac{2C}{\beta} \sum_{i=1}^n \int \text{Ent}_{q_i(\cdot | \bar{y}_i)}(f(\bar{y}_i, \cdot)) d\bar{q}_i(\bar{y}_i)$$

β minimal cond. probab., $C = C(A)$ if A coupling matrix

$$A_{ik} := \sup_{\substack{x,z \in \{\pm 1\}^n \\ x=z \text{ off } k}} d_{TV}(q_i(\cdot | \bar{x}_i), q_i(\cdot | \bar{z}_i))$$

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Ising model

Have $\|A\|_{\text{Op}} \leq \|J\|_{1 \rightarrow 1} \leq 1 - \alpha$ and $q_i(\cdot | \bar{\sigma}_i) \in (c_\alpha, 1 - C_\alpha)$
 \rightsquigarrow LSI with const. dep. on α only.

Application: $f = \sum_{|\mathbf{l}|=d} a_{\mathbf{l}} \sigma_{\mathbf{l}}$ d -homog. polynomial, $|a_{\mathbf{l}}| \leq 1$

$$\rightsquigarrow q^n(|f - \mathbb{E}f| \geq t) \leq 2 \exp\left(-\frac{t^{2/d}}{cn}\right)$$

improves upon older results by removing logarithmic dependencies

Similar results in presence of external field, i. e. normalize

$$\pi(\sigma) = \exp\left(\frac{1}{2} \sum_{i,j} J_{ij} \sigma_i \sigma_j + \sum_{i=1}^n h_i \sigma_i\right)$$

with h s. th. $\|h\|_\infty \leq \tilde{\alpha}$

Ising model

Have $\|A\|_{\text{Op}} \leq \|J\|_{1 \rightarrow 1} \leq 1 - \alpha$ and $q_i(\cdot | \bar{\sigma}_i) \in (c_\alpha, 1 - C_\alpha)$
 \rightsquigarrow LSI with const. dep. on α only.

Application: $f = \sum_{|I|=d} a_I \sigma_I$ d -homog. polynomial, $|a_I| \leq 1$

$$\rightsquigarrow q^n(|f - \mathbb{E}f| \geq t) \leq 2 \exp\left(-\frac{t^{2/d}}{cn}\right)$$

improves upon older results by removing logarithmic dependencies

Similar results in presence of external field, i. e. normalize

$$\pi(\sigma) = \exp\left(\frac{1}{2} \sum_{i,j} J_{ij} \sigma_i \sigma_j + \sum_{i=1}^n h_i \sigma_i\right)$$

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Thank you!