Some results on average-case matrix discrepancy

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Motivations / applications

Combinatorics, computational geometry, experimental design, theory of approximation algorithms, ...

Matousek '09 ; Chen&al '14 ; ...

Discrepancy of a set system / a set of vectors

$$\operatorname{disc}(\mathcal{S}) = \min_{\varepsilon \in \{\pm 1\}^n} \max_{i \in [d]} |\langle \varepsilon, a_i \rangle| = \min_{\varepsilon \in \{\pm 1\}^n} \left\| \sum_{i=1}^n \varepsilon_i u_i \right\|_{\infty} \quad \begin{array}{l} a_i \coloneqq \mathbb{1}_{S_i} \in \{0, 1\}^n \\ (u_i)_j \coloneqq (a_j)_i \end{array}$$

Cf. e.g. talks of Dan Spielman

Set system
$$S \coloneqq \{S_1, \cdots, S_d\}$$

 $S_i \subseteq \{1, \cdots, n\}$ \longleftrightarrow Characteristic $\#i$

For a coloring $\chi \in \{-1,1\}^n$, we define its discrepancy

$$\operatorname{disc}(\chi) \coloneqq \max_{i \in [d]} \left| \sum_{j \in S_i} \chi(j) \right|$$



 $disc(\chi) = max(0, 0, 1, 2) = 2$

$$\operatorname{disc}(\mathcal{S}) = \min_{\varepsilon \in \{\pm 1\}^n} \max_{i \in [m]} |\langle \varepsilon, a_i \rangle| = \min_{\varepsilon \in \{\pm 1\}^n} \left\| \sum_{i=1}^n \varepsilon_i u_i \right\|_{\infty}$$

Theorem (Spencer '85)

$$||u_i||_{\infty} \le 1$$
 \square disc $(u_1, \cdots, u_n) \le 6\sqrt{n}$

- The scaling \sqrt{n} is **optimal** (up to constants)
- Bansal '10: polynomial-time algorithms





Random Spencer: the symmetric binary perceptron $||u_i||_{\infty} \leq 1 \implies \operatorname{disc}(u_1, \cdots, u_n) \leq C\sqrt{n}$

What about random vectors ?

Given
$$g_1, \dots, g_d \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(0, I_n)$$
, can we find $\varepsilon \in \{\pm 1\}^n$ such that $\max_{i \in [d]} |\langle g_i, \varepsilon \rangle| \le \kappa \sqrt{n}$

This is the Symmetric Binary Perceptron (SBP) Aubin, Perkins & Zdeborová '19

 $(u_i)_j \coloneqq (g_j)_i$



Matrix discrepancy



Bansal&al '23 Matrix Spencer holds if we further assume $rk(\mathbf{A}_i) \lesssim n/\log^3 n$

Average-case matrix discrepancy: "Random Matrix Spencer"



Also introduced in [Kunisky & Zhang '23]

 \square Matrix analog of the SBP ($\mathbf{W}_i = \text{Diag}(\mathbf{g}_i)$ recovers the Symmetric Binary Perceptron)

 $\Box \quad \text{Trivial bound: if } \kappa > 2, \ \mathbb{P}\left[\left\|\sum_{i=1}^{n} \mathbf{W}_{i}\right\|_{\mathrm{op}} \le \kappa \sqrt{n}\right] = \mathbb{P}_{\mathbf{W} \sim \mathrm{GOE}(d)}[\|\mathbf{W}\|_{\mathrm{op}} \le \kappa] = 1 - o(1)$



 This talk
 > Sharp satisfiability transitions ?

 Goals
 > Structure of solution space ?

 Polynomial-time solving algorithms ? [Kunisky & Zhang '23]

 Add structure to W_i ? To probe the Matrix Spencer conjecture ?

Results I: first moment asymptotics

 $n/d^2 \to \tau > 0$

Number of solutions / Partition function

$$Z_{\kappa} \coloneqq \# \left\{ \varepsilon \in \{\pm 1\}^n \text{ s.t. } \left\| \sum_{i=1}^n \varepsilon_i \mathbf{W}_i \right\|_{\text{op}} \le \kappa \sqrt{n} \right\}$$

Theorem:
$$\lim_{d \to \infty} \frac{1}{d^2} \log \mathbb{E} Z_{\kappa} = (\tau - \tau_1(\kappa)) \log 2$$
$$\tau_1(\kappa) \coloneqq \frac{1}{\log 2} \left[-\frac{\kappa^4}{128} + \frac{\kappa^2}{8} - \frac{1}{2} \log \frac{\kappa}{2} - \frac{3}{8} \right]$$



$$\sum_{\varepsilon \in \{\pm 1\}^n} \mathbb{P}\left[\left\| \sum_{i=1}^n \varepsilon_i \mathbf{W}_i \right\|_{\mathrm{op}} \le \kappa \sqrt{n} \right] = 2^n \mathbb{P}[\|\mathbf{W}\|_{\mathrm{op}} \le \kappa] \quad \longrightarrow \quad \text{Left } (\kappa < 2) \text{ large deviations of } \|\mathbf{W}\|_{\mathrm{op}}$$

▶ Intuition: The events $\|\mathbf{W}\|_{op} \leq \kappa$ are driven by large deviations of the whole spectral density.

 $\mathbb{P}[\mu_{\mathbf{W}} \simeq \mu] \simeq \exp\{-d^2 I(\mu)\}$ Ben Arous & Guionnet '97; Dean&Majumdar '06 '08;

Proof: technical adaptations of the proof of [BAG '97] see also [Anderson, Guionnet & Zeitouni '10]

$$\lim \frac{1}{d^2} \log \mathbb{P}[\|\mathbf{W}\|_{\mathrm{op}} \le \kappa] = -\inf_{\mu \in \mathcal{M}([-\kappa,\kappa])} I(\mu)$$

> Compute $\rho_{\kappa}(x)$ from **Tricomi's theorem**

Tricomi' 85; Dean&Majumdar '06 '08; Vivo&al '07, ...

Prove $\rho_{\kappa} = \operatorname*{arg\,min}_{\mu \in \mathcal{M}([-\kappa,\kappa])} I(\mu)$

 \geq

 $\mathbb{E}Z_{\kappa} =$

Classical tools of logarithmic potential theory Saff&Totik'13; Ben Arous & Guionnet '97

$$I(\mu) \coloneqq -\frac{1}{2} \int \mu(\mathrm{d}x)\mu(\mathrm{d}y) \log |x-y| + \frac{1}{4} \int \mu(\mathrm{d}x) \, x^2 - \frac{3}{8}$$

 $\mathbf{W} \sim \text{GOE}(d)$



Results II: Upper bounds via the second moment method



 $n/d^2 \to \tau > 0$

Second moment method: overview

$$\operatorname{disc}(\{\mathbf{W}_i\}) = \min_{\varepsilon \in \{\pm 1\}^n} \left\| \sum_{i=1}^n \varepsilon_i \mathbf{W}_i \right\|_{\operatorname{op}}$$

$$n/d^2 \to \tau > 0$$

Theorem
$$\tau > \tau_2(\kappa) \longrightarrow \mathbb{P}\left[Z_{\kappa} := \#\left\{\varepsilon \in \{\pm 1\}^n \text{ s.t. } \left\|\sum_{i=1}^n \varepsilon_i \mathbf{W}_i\right\|_{op} \le \kappa \sqrt{n}\right\} \ge 1\right] = 1 - o(1)$$

Explicit formula (non-optimal)
Proof
1 Sharp 1st moment $(1/d^2) \log \mathbb{E}Z_{\kappa} \to \cdots$
2 Second moment upper bound $\frac{\mathbb{E}[Z_{\kappa}^2]}{\mathbb{E}[Z_{\kappa}]^2} \le \left[1 - \frac{\tau_2(\kappa)}{\tau}\right]^{-1/2}$
3 Margin concentration $\operatorname{Var}[\operatorname{disc}(\mathbf{W}_1, \cdots, \mathbf{W}_n)] \lesssim \frac{1}{d} (\mathbb{E}[\operatorname{disc}(\mathbf{W}_1, \cdots, \mathbf{W}_n)])^2$
Results of [Altschuler'23], based on Talagrand's $L^1 - L^2$ inequality

 $\frac{\mathbb{E}[Z_{\kappa}^2]}{\mathbb{E}[Z_{\kappa}]^2} \lesssim \left[1 - \frac{\tau_2(\kappa)}{\tau}\right]^{-1/2}$

10

$$\frac{\mathbb{E}[Z_{\kappa}^{2}]}{\mathbb{E}[Z_{\kappa}]^{2}} = \frac{1}{2^{n}} \sum_{l=0}^{n} \binom{n}{l} \exp\{nG_{d}(q_{l})\}, \text{ where for } q \in [-1,1]: G_{d}(q) \coloneqq \frac{1}{n} \log \frac{\mathbb{P}\left[\|\mathbf{W}\|_{\mathrm{op}} \le \kappa \text{ and } \|q\mathbf{W} + \sqrt{1-q^{2}}\mathbf{Z}\|_{\mathrm{op}} \le \kappa\right]}{\mathbf{P}[\|\mathbf{W}\|_{\mathrm{op}} \le \kappa]^{2}}$$

$$q_{l} \coloneqq 2(l/n) - 1 \qquad \qquad \mathbf{W}, \mathbf{Z} \sim \langle \cdot \rangle_{q,\kappa}$$

Large deviations of the spectral norm of correlated GOE(d) matrices.

- \blacktriangleright Upper bound on $G_d(q)$
- Crude upper bound for q far from $0: G_d(q) \leq -\frac{1}{n} \log \mathbb{P}[\|\mathbf{W}\|_{\text{op}} \leq \kappa].$
- ★ For small q, upper bounding sup $G''_{d}(q) \leq \frac{\tau_{2}(\kappa)}{\tau}$ ↓ Log-Sobolev inequality for $\langle \cdot \rangle_{q,\kappa}$ Approximation of $\mathbb{1}[|x| \leq \kappa]$ by smooth functions
 ◆ Bakry-Emery condition for smooth and strongly log-concave measures
 □ Concentration of moments $\operatorname{Tr}[\mathbf{W}^{a}\mathbf{Z}^{b}]$ under $\langle \cdot \rangle_{q,\kappa}$.
 - $\blacktriangleright \quad \text{Discrete Laplace's method over the overlap } q \in [-1,1] \text{ in } \frac{1}{2^n} \sum_{l=0}^n \binom{n}{l} \exp\{nG_d(q_l)\}. \quad \Box \quad \underbrace{\mathbb{E}[Z_{\kappa}]^2}_{\mathbb{E}[Z_{\kappa}]^2} \lesssim \left[1 \frac{\tau_2(\kappa)}{\tau}\right]^{-1/2}$



Results III: failure of the second moment method

$$G_d(q) \coloneqq \frac{1}{n} \log \frac{\mathbb{P}\left[\|\mathbf{W}\|_{\text{op}} \le \kappa \text{ and } \|q\mathbf{W} + \sqrt{1 - q^2} \mathbf{Z}\|_{\text{op}} \le \kappa \right]}{\mathbb{P}[\|\mathbf{W}\|_{\text{op}} \le \kappa]^2}$$

 $\blacktriangleright \quad \text{Explicit computation:} \quad \lim_{d \to \infty} G_d''(0) = \frac{\tau_{\text{fail.}}(\kappa)}{\tau}$

$$\tau_{\text{fail.}}(\kappa) \coloneqq \frac{1}{2} \left(\frac{\kappa^2}{4} - 1\right)^4$$

Theorem

Non-concentration of Z_{κ} on $\mathbb{E}[Z_{\kappa}]$.

 $\succ \text{ Lower bound in Laplace's method: } \left(n/d^2 \to \tau < \tau_{\text{fail.}}(\kappa) \ \bigsqcup _{d \to \infty} \lim _{d \to \infty} \frac{1}{d^2} \log \frac{\mathbb{E}[Z_{\kappa}^2]}{\mathbb{E}[Z_{\kappa}]^2} > 0 \right)$

Technical assumption: uniform continuity of $G_d''(q)$ in q=0 as $d\to\infty$

CAUTION WORK IN PROGRESS

Failure of the second moment method

$$n/d^2 \to \tau < \tau_{\text{fail.}}(\kappa) \implies \liminf_{d \to \infty} \frac{1}{d^2} \log \frac{\mathbb{E}[Z_{\kappa}^2]}{\mathbb{E}[Z_{\kappa}]^2} > 0$$



• The phase diagram is more complex than in the Symmetric Binary Perceptron !

Summary: average-case matrix discrepancy



