S&DS 602: Homework 10

Due Wednesday, November 13 at 2PM, via Gradescope

1. Let $X \in \mathbb{R}^{n \times n}$ be a symmetric random matrix whose entries above the diagonal are i.i.d. $\mathcal{N}(0,1)$, and whose entries on the diagonal are i.i.d. $\mathcal{N}(0,2)$. Let $\lambda_{\max}(X)$ be its largest eigenvalue. Apply the Sudakov-Fernique inequality to show that

$$\mathbb{E}[\lambda_{\max}(X)] \le 2\sqrt{n}$$

Show that this implies, for a universal constant c > 0 and any $t \ge 0$,

$$\mathbb{P}[\lambda_{\max}(X) \ge 2\sqrt{n} + t] \le e^{-ct^2}.$$

[Hint: $\lambda_{\max}(X) = \sup_{u \in S^{n-1}} u^{\top} X u$. Compare with $\sup_{u \in S^{n-1}} 2u^{\top} g$ where $g \sim \mathcal{N}(0, I)$.]

The remaining problems relate to the following extension of the Slepian-Sudakov-Fernique inequalities due to Gordon:

Theorem. (Gordon's inequalities) Let $\{X_{ut}\}_{(u,t)\in U\times T}$ and $\{Y_{ut}\}_{(u,t)\in U\times T}$ be two separable mean-zero Gaussian processes.

(a) Suppose, for all $u \neq v \in U$ and all $t, s \in T$, that

$$\mathbb{E}(X_{ut} - X_{us})^2 \le \mathbb{E}(Y_{ut} - Y_{us})^2, \qquad \mathbb{E}(X_{ut} - X_{vs})^2 \ge \mathbb{E}(Y_{ut} - Y_{vs})^2.$$

Then

$$\mathbb{E}\left[\inf_{u\in U}\sup_{t\in T}X_{ut}\right] \leq \mathbb{E}\left[\inf_{u\in U}\sup_{t\in T}Y_{ut}\right]$$

(b) If, in addition, $\mathbb{E}X_{ut}^2 = \mathbb{E}Y_{ut}^2$ for all $u \in U$ and $t \in T$, then for any $\tau \in \mathbb{R}$,

$$\mathbb{P}\left[\inf_{u\in U}\sup_{t\in T}X_{ut}\geq \tau\right]\leq \mathbb{P}\left[\inf_{u\in U}\sup_{t\in T}Y_{ut}\geq \tau\right].$$

Problem 2 will ask you to show part (b). You may use both parts (a) and (b) in the remaining Problems 3 and 4.

2. (Vershynin 7.2.14) Prove part (b) of the above theorem.[Hint: The proof is similar to that of Slepian's inequality. Consider first the case where U and T are finite sets, and apply Gaussian interpolation to the function

$$\prod_{u \in U} \left(1 - \prod_{t \in T} \varphi(X_{ut}) \right)$$

where $\varphi(x)$ is a smooth approximation to $\mathbf{1}\{x < \tau\}$.]

3. (Vershynin 7.3.4) Suppose $X \in \mathbb{R}^{n \times m}$ has i.i.d. $\mathcal{N}(0, 1)$ entries, and $n \geq m$. Let $s_{\min}(X)$ be its smallest singular value. Apply Gordon's inequality to show

$$\mathbb{E}s_{\min}(X) \ge \sqrt{n} - \sqrt{m}.$$

Show that this implies, for a universal constant c > 0 and any $t \ge 0$,

$$\mathbb{P}[s_{\min}(X) \le \sqrt{n} - \sqrt{m} - t] \le e^{-ct^2}.$$

[Hint: $s_{\min}(X) = \inf_{u \in S^{m-1}} ||Xu||_2 = \inf_{u \in S^{m-1}} \sup_{t \in S^{n-1}} t^\top Xu.$]

4. (Convex Gaussian min-max theorem) (a) Let $X \in \mathbb{R}^{n \times m}$, $g \in \mathbb{R}^n$, and $h \in \mathbb{R}^m$ be independent with i.i.d. $\mathcal{N}(0, 1)$ entries. Let $U \subset \mathbb{R}^n$, $T \subset \mathbb{R}^m$, and define

$$\Phi(X) = \inf_{u \in U} \sup_{t \in T} u^{\top} X t$$
$$\phi(g, h) = \inf_{u \in U} \sup_{t \in T} \|u\|_2 g^{\top} t + \|t\|_2 h^{\top} u.$$

Show that for any $\tau \in \mathbb{R}$,

$$\mathbb{P}[\Phi(X) < \tau] \le 2 \mathbb{P}[\phi(g, h) < \tau].$$

[Hint: Let $z \sim \mathcal{N}(0, 1)$ be independent of X, g, h, and consider

$$\tilde{\Phi}(X,z) = \inf_{u \in U} \sup_{t \in T} u^{\top} X t + z \|u\|_2 \|t\|_2.$$

Show first that $\mathbb{P}[\Phi(X) < \tau] \le \mathbb{P}[\tilde{\Phi}(X, z) < \tau \mid z \le 0] \le 2 \mathbb{P}[\tilde{\Phi}(X, z) < \tau].]$

(b) Suppose, in addition, that U and T are compact and convex sets. Show that also for any $\tau \in \mathbb{R}$,

$$\mathbb{P}[\Phi(X) > \tau] \le 2 \mathbb{P}[\phi(g, h) > \tau],$$

and hence for any $c \in \mathbb{R}$ and $\tau > 0$,

$$\mathbb{P}[|\Phi(X) - c| > \tau] \le 2 \mathbb{P}[|\phi(g, h) - c| > \tau].$$

[Hint: For any continuous $f: U \times T \to \mathbb{R}$ that is convex in u and concave in t,

$$\sup_{t \in T} \inf_{u \in U} f(u, t) = \inf_{u \in U} \sup_{t \in T} f(u, t)$$

(You may use this without proof.) Apply this and part (a) with $-\Phi(X)$.]