Improved bounds for random projection

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Notation. We say that an event A occurs "with high probability" if $\mathbb{P}(A)$ tends to 1 as the dimension $d_{\min} = \min\{d_1, d_2\}$ tends to infinity. We say that A occurs "with very high probability" if $\mathbb{P}(A)$ tends to 1 faster than any polynomial of d_{\min} .

Let $E \in \mathbb{R}^{d_1 \times d_2}$ be a real matrix. The matrix spectral norm is defined as

$$\|\boldsymbol{E}\|_{\sigma} = \max_{(\boldsymbol{a}, \boldsymbol{b}) \in \mathbb{S}^{d_1 - 1} \times \mathbb{S}^{d_2 - 1}} \boldsymbol{a}^T \boldsymbol{E} \boldsymbol{b},$$

where $\mathbb{S}^{d-1} = \{ \boldsymbol{a} \in \mathbb{R}^d : \|\boldsymbol{a}\|_{\sigma} = 1 \}$ denotes the unit sphere in \mathbb{R}^d .

Theorem 0.1. Consider a noisy rank-1 matrix model $\mathbf{D} = \lambda \mathbf{a} \otimes \mathbf{b} + \mathbf{E}$, where $\lambda \in \mathbb{R}_+$ is a scalar, $\mathbf{a} \in \mathbb{R}^{d_1}, \mathbf{b} \in \mathbb{R}^{d_2}$ are unit-1 vectors, and $\mathbf{E} \in \mathbb{R}^{d_1 \times d_2}$ is a Gaussian matrix with i.i.d. $N(0, \sigma^2)$ entries. Define a random projection

$$\hat{\boldsymbol{a}} = \boldsymbol{D}\boldsymbol{\Omega}, \text{ where } \boldsymbol{\Omega} = (z_1, \dots, z_{d_2})^T \overset{i.i.d.}{\sim} N(0, 1).$$

Suppose $\lambda \gg \sigma \sqrt{d_2 \max(d_1, d_2)}$ as $d_1, d_2 \to \infty$. Then, $\cos \Theta(\boldsymbol{a}, \hat{\boldsymbol{a}}) \to 1$ in probability.

Proof. The perturbed rank-1 model $D = \lambda a \otimes b + E$ implies that

$$\hat{a} = D\Omega = \lambda Z a + E\Omega, \tag{1}$$

where $Z \stackrel{\text{def}}{=} \langle \boldsymbol{b}, \boldsymbol{\Omega} \rangle$ is an N(0,1) random variable. Note that $\boldsymbol{E}\boldsymbol{\Omega} \in \mathbb{R}^{d_1}$ is a random vector with length $\|\boldsymbol{E}\boldsymbol{\Omega}\|_2 \leq \|\boldsymbol{E}\|_{\sigma} \|\boldsymbol{\Omega}\|_2$.

To show $\cos(a, \hat{a}) \to 1$, it suffices to show that $\cot(a, \hat{a}) \to \infty$. Based on (1), we have

$$\cot \Theta(\boldsymbol{a}, \hat{\boldsymbol{a}}) \ge \frac{\|\lambda Z \boldsymbol{a}\|_2}{\|\boldsymbol{E}\boldsymbol{\Omega}\|_2} \ge \frac{\lambda |Z|}{\|\boldsymbol{E}\|_{\sigma} \|\boldsymbol{\Omega}\|_2}.$$
 (2)

Now consider the asymptotical property of (2) as $d_1, d_2 \to \infty$. The fact $\|\Omega\|_2^2 \sim \chi^2(d_2)$ implies that $\|\Omega\|_2 \approx (1 + o(1))\sqrt{d_2}$ with very high probability, for some constant C > 0. Furthermore, we have that $\|E\|_{\sigma} \approx (2 + o(1))\sigma\sqrt{\max(d_1, d_2)}$ by Lemma 1. Therefore, for any fixed L > 0,

$$\mathbb{P}(\cot(\boldsymbol{a}, \hat{\boldsymbol{a}}) \ge L) \ge \mathbb{P}\left(\frac{\lambda |Z|}{\|\boldsymbol{E}\|_{\sigma} \|\boldsymbol{\Omega}\|_{2}} \ge L\right) \\
\ge \mathbb{P}\left(|Z| \ge \frac{2L\sigma\sqrt{d_{2} \max(d_{1}, d_{2})}}{\lambda}\right) \\
\ge 1 - \frac{4\lambda_{*}}{\sqrt{2\pi}}, \tag{3}$$

where $\lambda_* \stackrel{\text{def}}{=} L\sigma\sqrt{d_2\max(d_1,d_2)}/\lambda$. Here the last line of (3) uses the fact that $\mathbb{P}(|N(0,1)| \geq t) \geq t$

 $1-2t\phi(0)$ for any $t \geq 0$, where $\phi(\cdot)$ is the cdf for standard normal. Based on assumption $\lambda_* \to 0$ we conclude that $\cot(\boldsymbol{a}, \hat{\boldsymbol{a}}) \geq L$ with high probability. Sending $L \to \infty$ gives the desired result. \square Lemma 1 (Spectral norm of Gaussian matrix [1]). Let $\boldsymbol{E} \in \mathbb{R}^{d_1 \times d_2}$ be a random matrix with i.i.d. N(0,1) entries. Then, we have, with very high probability,

$$\|\boldsymbol{E}\|_{\sigma} \approx (2 + o(1))\sqrt{\max(d_1, d_2)}.$$

References

[1] Mark Rudelson and Roman Vershynin. Non-asymptotic theory of random matrices: extreme singular values. In *Proceedings of the International Congress of Mathematicians 2010 (ICM 2010) (In 4 Volumes) Vol. I: Plenary Lectures and Ceremonies Vols. II–IV: Invited Lectures*, pages 1576–1602. World Scientific, 2010.