

EE-546: Assignment # 2

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Problem # 1

Problem 1a):

$$\begin{aligned}
 Z &= \frac{1}{m} \sum_{r=1}^m X_r |X_r| \operatorname{sign}(X_r + \mathbf{a}_r^T \mathbf{y}) \\
 &= \frac{1}{m} \sum_{r=1}^m X_r^2 \operatorname{sign}(X_r) \operatorname{sign}(X_r + \mathbf{a}_r^T \mathbf{y}) \\
 \text{Let } Z_r &= X_r^2 \operatorname{sign}(X_r) \operatorname{sign}(X_r + \mathbf{a}_r^T \mathbf{y}) \\
 \mathbf{a}_r^T \mathbf{y} &\sim \mathcal{N}(0, \mathbf{y}^T \mathbf{y}) \\
 \Rightarrow \frac{\mathbf{a}_r^T \mathbf{y}}{\sqrt{\mathbf{y}^T \mathbf{y}}} &\sim \mathcal{N}(0, 1) \\
 \mathbb{E}[Z_r] &= \mathbb{E}[X_r^2 \operatorname{sign}(X_r) \operatorname{sign}(X_r + \mathbf{a}_r^T \mathbf{y})] \\
 &= \mathbb{E}[X_r^2 \operatorname{sign}(X_r) \operatorname{sign}(X_r + \sqrt{\mathbf{y}^T \mathbf{y}} \frac{\mathbf{a}_r^T \mathbf{y}}{\sqrt{\mathbf{y}^T \mathbf{y}}})] \\
 &= \frac{2}{\pi} \tan^{-1} \frac{1}{\sqrt{\mathbf{y}^T \mathbf{y}}} + \frac{2}{\pi} \frac{\sqrt{\mathbf{y}^T \mathbf{y}}}{1 + \mathbf{y}^T \mathbf{y}} \\
 \Rightarrow \mathbb{E}[Z] &= \frac{1}{m} \sum_{r=1}^m \mathbb{E}[Z_r] \\
 &= \frac{2}{\pi} \tan^{-1} \frac{1}{\sqrt{\mathbf{y}^T \mathbf{y}}} + \frac{2}{\pi} \frac{\sqrt{\mathbf{y}^T \mathbf{y}}}{1 + \mathbf{y}^T \mathbf{y}}
 \end{aligned}$$

Problem 1b):

$$\begin{aligned}
 \mathbb{P}(|X_r^2 \operatorname{sign}(X_r) \operatorname{sign}(X_r + \mathbf{a}_r^T \mathbf{y})| \geq t) &= \mathbb{P}(|X_r^2| \geq t) && \because \operatorname{sign} \text{ is immaterial} \\
 &= P(\chi_1^2 \geq t)
 \end{aligned}$$

$P(\chi^2 \geq t)$ is easily obtained as χ^2 is sub-exponential,
 $Z \sim \chi_m^2$ and hence:

$$P(|Z - \mathbb{E}[Z]| \geq t) \leq \exp(1 - \frac{t}{k_1})$$

Problem # 2

Problem 2 (i):

From Cauchy Schwartz inequality we have:

$$|\langle x, y \rangle| \leq \|x\| \|y\|$$

Thus,

$$\begin{aligned} \langle x, Ay \rangle &\leq \|x\| \|Ay\| \\ \max_{\|x\|_{l_2}=1, \|y\|_{l_2}=1} \langle x, Ay \rangle &\leq \max_{\|x\|_{l_2}=1, \|y\|_{l_2}=1} \|x\| \|Ay\| \\ &= \max_{\|y\|_{l_2}=1} \|Ay\| && \because \|x\| = 1 \\ &= \|A\| && \because \text{from the definition of } \|A\| \end{aligned}$$

Fact: For unitary U , $U^T U = I \implies \|Ux\|_{l_2}^2 = x^T U^T U x = \|x\|_{l_2}^2$

Now consider $\|Ax\|$:

$$\begin{aligned} \sup_{\|x\|=1} \|Ax\| &= \sup_{\|x\|=1} \|U\Sigma V^T x\| && \because \text{spectral decomposition of } A \\ &= \sup_{\|x\|=1} \|\Sigma V^T x\| && \because \|Ux\| = \|x\| \text{ for unitary } U \\ &= \sup_{\|y\|=1} \|\Sigma y\| && \because y = V^T x \text{ and } y^T y = 1 \end{aligned}$$

Since Σ is a diagonal matrix $\sup_{\|y\|=1} \|\Sigma y\|$ can easily be obtained when $y = (1, 0, \dots, 0)$ such that the maximum will be $\sigma_1(A)$

Problem 2 (ii):

$$\begin{aligned} \text{trace}(U^T AV) &= \sum_{i=1}^r u_i^T a_{ii} v_i \\ \max \text{trace}(U^T AV) &= \max \sum_{i=1}^r u_i^T a_{ii} v_i \\ &= \sum_{i=1}^r \sigma_i(A) \end{aligned}$$

Problem # 3

Upper bound:

$$\begin{aligned}
\|A\| &= \sup_{\|x\|=1} \|Ax\| = \sup_{\|x\|=1} \sqrt{\sum_{i=1}^m \left(\sum_{j=1}^n A_{ij}x_j\right)^2} \\
&\leq \sup_{\|x\|=1} \sqrt{\sum_{i=1}^m \left(\sum_{j=1}^n A_{ij}^2\right) \left(\sum_{k=1}^n x_k^2\right)} && \because \text{using Cauchy-Schwartz inequality} \\
\Rightarrow \|A\| &\leq \sqrt{m} \max_{i \in \{1, 2, \dots, m\}} \left(\sum_{j=1}^n A_{ij}^2\right)^{1/2}
\end{aligned}$$

For equality $A_{ij}x_j = \lambda x_j$ and hence $A = I$ is one such matrix.

Lower bound:

Consider $x = \left(\frac{1}{\sqrt{n}}, \frac{1}{\sqrt{n}}, \dots, \frac{1}{\sqrt{n}}\right)$

them

$$\begin{aligned}
\|A\| &= \sup_{\|x\|=1} \|Ax\| \geq \sqrt{\sum_{i=1}^m \left(\sum_{j=1}^n A_{ij} \frac{1}{\sqrt{n}}\right)^2} \\
&= \frac{1}{\sqrt{n}} \sqrt{\sum_{i=1}^m \left(\sum_{j=1}^n A_{ij} \times 1\right)^2} \\
&\geq \frac{1}{\sqrt{n}} \sqrt{\frac{1}{m} \left(\sum_{i=1}^m \left|\sum_{j=1}^n A_{ij}\right|\right)^2} \\
&= \frac{1}{\sqrt{mn}} \sum_{i=1}^m \left|\sum_{j=1}^n A_{ij}\right|
\end{aligned}$$

Problem # 4

$$\begin{aligned}
\|A\|_F &= \left(\sum_{i=1}^m \sum_{j=1}^n |a_{ij}|^2 \right)^{\frac{1}{2}} \\
&= (\text{Tr}(A^T A))^{\frac{1}{2}} \\
&= (\text{Tr}(V \Sigma U^T U \Sigma V^T))^{\frac{1}{2}} && \because \text{Spectral decomposition of } A \\
&= (\text{Tr}(V \Sigma^2 V^T))^{\frac{1}{2}} \\
&= (\text{Tr}(V^T V \Sigma^2))^{\frac{1}{2}} && \because \text{Tr}(AB) = \text{Tr}(BA) \\
&= (\text{Tr}(\Sigma^2))^{\frac{1}{2}} \\
&= \left(\sum_{i=1}^{\min(m,n)} \sigma_i^2(A) \right)^{\frac{1}{2}}
\end{aligned}$$

From Problem 2 $\|A\| = \sigma_1(A)$. Now, given $\sigma_1(A) \geq \sigma_2(A) \geq \dots \geq \sigma_{\min(m,n)}(A)$

$$\begin{aligned}
\sigma_1(A) &\leq \sqrt{\sigma_1(A)^2 + \sigma_2(A)^2 + \dots + \sigma_{\min(m,n)}(A)^2} \\
\sqrt{\sigma_1(A)^2 + \sigma_2(A)^2 + \dots + \sigma_{\min(m,n)}(A)^2} &\leq \sqrt{\sigma_1(A)^2 + \sigma_1(A)^2 + \dots + \sigma_1(A)^2} \\
&= \sqrt{\text{rank}(A)} \sigma_1(A)
\end{aligned}$$

where the last equality follows from the fact that only $\text{rank}(A)$ singular vectors are non zero. Thus,

$$\sigma_1(A) \leq \sqrt{\sigma_1(A)^2 + \sigma_2(A)^2 + \dots + \sigma_{\min(m,n)}(A)^2} \leq \sqrt{\text{rank}(A)} \sigma_1(A)$$

\Rightarrow

$$\sigma_1(A) \leq \|A\| \leq \sqrt{\text{rank}(A)} \|A\|$$