

Topic: Fast Randomized SVD

tall Gaussian matrices:

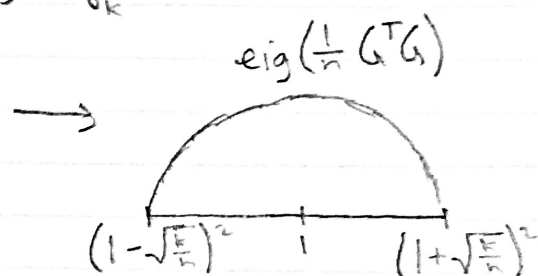
$$\begin{matrix} k \\ m \end{matrix} \boxed{G}$$

$G \sim N(0,1)$, not symmetric

$k \leq m$, denote k singular values as $\sigma_1 \dots \sigma_k$

defn: condition number $\kappa(G) = \frac{\sigma_1}{\sigma_k}$

from M-P: $\begin{cases} \sigma_1 \approx \sqrt{m} + \sqrt{k} \\ \sigma_k \approx \sqrt{m} - \sqrt{k} \end{cases}$



Thm (Gordon's): $X_{m \times n}$, $X_{ij} \sim \text{i.i.d. } N(0,1)$, $m \geq n$

then $\sqrt{m} - \sqrt{n} \leq E \sigma_n \leq E \sigma_1 \leq \sqrt{m} + \sqrt{n}$

— stronger than last lecture's result, non-asymptotic

Recall Gaussian concentration:

σ_i or σ_k is 1-Lip function of G

\rightarrow w.h.p. $|\sigma_i - E \sigma_i| \sim O(1)$ i.e. $\Pr(|\sigma_i - E \sigma_i| > \alpha) \leq e^{-\alpha^2/2}$

\rightarrow w.h.p. $-\sqrt{m} - \sqrt{k} - t_2 \leq \sigma_k \leq \sigma_1 \leq \sqrt{m} + \sqrt{k} + t_1$, where $t_1, t_2 \sim O(1)$

Thus, $\kappa(G) \leq \frac{\sqrt{m} + \sqrt{k} + t_1}{\sqrt{m} - \sqrt{k} - t_2}$, where $m \geq k \gg t_1, t_2$

\rightarrow if $m \gg k$, $\kappa(G)$ will be close to 1

Randomized SVD:

Goal: given $A_{m \times n}$, target k -rank factorization

$$A_{m \times n} = B_{m \times k} \cdot C_{k \times n} \quad \text{e.g. } \min_{B, C} \|BC - A\|_{Fro} \rightarrow \text{"k-SVD"}$$

assume $\sigma_1(A) \geq \dots \geq \sigma_r(A)$, $\sigma_{r+1}(A) = \epsilon$

Note: classical complexity of k -SVD is $O(mnk)$

Idea: draw $\Omega_{n \times k'}$, Ω_{ij} i.i.d. $N(0, 1)$
 \rightarrow compute SVD of $A\Omega$ (can be done $O(mk^2)$)

Note: Ω is well-conditioned, $k' \approx k$

Note: $\text{Range}_k(A\Omega) \approx \text{Range}_k(A)$

Approximate k-SVD algorithm:

Input: $A_{m \times n}$, k

Step 1: draw $\Omega_{n \times l}$, $l = 2k$

$$Y \leftarrow A\Omega$$

$$Y = QR, \quad Q_{m \times l}, \quad QQ^T \approx I_{l \times l}$$

Step 2: compute SVD of $Q^T A$
i.e. $Q^T A = \tilde{U} S V^T$

Step 3: $U \leftarrow Q\tilde{U}$

Output: $U_{m \times l}$, $S_{l \times l}$, $V_{n \times l}$

Then: (U, S, V) approximates k -SVD of A

$$\text{i.e. } E\|A - USV^T\| \leq \left(1 + 4\sqrt{\frac{2n}{k-1}}\right) \sigma_{k+1}(A)$$

$$\text{Recall: } \sigma_{k+1}(A) = \varepsilon$$

Applying concentration argument, we obtain:

$$\|A - USV^T\|_{\text{op}} \leq C \cdot \sigma_{k+1}(A) \text{ w.h.p.}$$