

$$A \in \mathbb{R}^{m \times n}.$$

$$m > n.$$

$$A_k = \sum_{i=1}^k \sigma_i \vec{u}_i \vec{v}_i^T$$

$$A_k = \operatorname{argmin}_{\substack{B \in \mathbb{R}^{m \times n} \\ \operatorname{Rank}(B) = k}} \|A - B\|_F^2$$

$$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n > 0.$$

Proof: Consider: $\|A - A_k\|_F^2 = \left\| \sum_{i=1}^n \sigma_i \vec{u}_i \vec{v}_i^T - \sum_{i=1}^k \sigma_i \vec{u}_i \vec{v}_i^T \right\|_F^2$

$$= \left\| \sum_{i=k+1}^n \sigma_i \vec{u}_i \vec{v}_i^T \right\|_F^2 = \sum_{i=k+1}^n \sigma_i^2$$

Recall: Frob. Norm

is invariant to orthogonal ^{normal} transformations.

$$\|A\|_F^2 = \|U \Sigma V^T\|_F^2 = \|\Sigma\|_F^2 = \sum_{i=1}^n \sigma_i^2$$

$$\|A\|_F = \|AU\|_F$$

So we want to show:

For any B , that is rank (k) :

$$\|A - B\|_F^2 \geq \|A - A_k\|_F^2 = \sum_{i=k+1}^n \sigma_i^2$$

Notation: $\sigma_i(A)$: i th singular value of A
 $\sigma_i(A-B)$: " " " " " " $A-B$

$$\|A-B\|_F^2 = \sum_{i=1}^n \sigma_i^2(A-B)$$

all sing.
values of $A-B$.

≥

WANT

$$\|A - A_k\|_F^2 = \sum_{i=k+1}^n \sigma_i^2(A)$$

(n-k) smallest SV of A .

So if we could show:

$$\sigma_i^2(A-B) \geq \sigma_{k+i}^2(A)$$

How to show this?

Rank constraint

Ingredients:

- Spectral norm.
- B has rank k . $\Rightarrow \sigma_{k+i}(B) = 0$.
- Triangle ineq.

$\sigma_{k+i}(A)$ = $k+i$ th largest singular value of A .

= largest SV after removing the top $k+i-1$ singular values.

$$= \|A - A_{k+i-1}\|_2$$

A_{k+i-1} is the best rank $(k+i-1)$ approximation to A in the spectral norm sense.

$$\sigma_i(A-B)$$

$$A-B=C$$

$$\sigma_i(C) = \|C - C_{i-1}\|_2 \quad : \quad C_{i-1} : \text{Best rank } i-1 \text{ approx to } C$$

$$\begin{aligned} \sigma_{k+1}(B) &= 0 \\ &= \|B - B_k\|_2 \end{aligned} \quad B_k : \text{Best rank } k \text{ approx. to } B \text{ in spectral norm sense.}$$

$$\begin{aligned} \sigma_i(A-B) &= \sigma_i(C) \\ &= \|C - C_{i-1}\|_2 + \|B - B_k\|_2 \end{aligned}$$

$$\geq \|C + B - C_{i-1} - B_k\|_2 \quad \left(\begin{array}{l} \text{Triangle} \\ \text{inequality} \\ \text{for spectral norm} \end{array} \right)$$

$$= \|A - C_{i-1} - B_k\|_2$$

$$= \|A - D\|_2$$

$$\geq \|A - A_{k+i-1}\|_2$$

$$= \sigma_{k+i}(A)$$

$$A_{k+i-1} = \underset{\text{Rank}(D) \leq k+i-1}{\text{argmin}} \|A - D\|_2$$

$$\begin{aligned} \sigma_{k+i}(A) &= \|A - A_{k+i-1}\|_2 \\ &= \|B + C - A_{k+i-1}\|_2 \end{aligned}$$

$$\text{Let } \underbrace{C_{i-1}}_{\text{rank } i-1} + \underbrace{B_k}_{\text{rank } k} = D$$

$$\begin{aligned} \text{Rank}(D) &\leq \text{Rank}(C_{i-1}) + \text{Rank}(B_k) \\ &\leq k+i-1 \end{aligned}$$