

Matrix Nearness Problem

Definition: Distance

Let $|||\cdot|||$ be a matrix norm on M_n , $S \subseteq M_n$, and $A \notin S$. The *distance* from A to S is given by:

$$d(A) = \inf_{X \in S} \{|||A - X|||\}$$

Questions:

- 1). Compute $d(A)$
- 2). Does $X_A \in S$ exist such that $d(A) = |||X - X_A|||$?
- 3). If X_A exists, is it unique?

Lemma

Let $|||\cdot|||$ be a matrix norm on M_n and $A \in M_n$:

$$|||I - A||| < 1 \implies A \text{ is invertible}$$

Proof

Assume $|||I - A||| < 1$

Let $A = I - (I - A)$

$$A^{-1} = [I - (I - A)]^{-1} = \sum_{k=0}^{\infty} (I - A)^k$$

Check for absolute convergence under the norm:

$$|||\sum_{k=0}^{\infty} (I - A)^k||| \leq \sum_{k=0}^{\infty} |||I - A|||^k$$

which converges for $|||I - A||| < 1$.

Theorem: Best Singular Approximation

Let $|||\cdot|||$ be the operator norm and let S be the collection of all singular matrices. Fix $A \notin S$:

$$d(A) = s_n$$

the smallest singular value.

Proof

Consider the SVD for A :

$$A = V \begin{bmatrix} s_1 & & 0 \\ & \ddots & \\ 0 & & s_n \end{bmatrix} W^*$$

$$A^{-1} = (W^*)^{-1} \begin{bmatrix} \frac{1}{s_1} & & 0 \\ & \ddots & \\ 0 & & \frac{1}{s_n} \end{bmatrix} V^{-1} = W \begin{bmatrix} \frac{1}{s_1} & & 0 \\ & \ddots & \\ 0 & & \frac{1}{s_n} \end{bmatrix} V^*$$

Note that this is the SVD for A^{-1} , so $|||A^{-1}||| = \frac{1}{s_n}$.

Assume $B \in S$

$A^{-1}B \in S$

By the CP of the above lemma:

$$\begin{aligned} |||I - A^{-1}B||| &\geq 1 \\ |||A^{-1}(A - B)||| &\geq 1 \\ |||A^{-1}||| |||A - B||| &\geq 1 \\ \frac{1}{s_n} |||A - B||| &\geq 1 \\ |||A - B||| &\geq s_n \end{aligned}$$

$$\text{Now, let } X_A = V \begin{bmatrix} s_1 & & 0 \\ & \ddots & \\ & & s_n \\ 0 & & 0 \end{bmatrix} W^* \in S$$

$$|||A - X_A||| = \left\| \left\| V \begin{bmatrix} 0 & & 0 \\ & \ddots & \\ & & 0 \\ 0 & & s_n \end{bmatrix} W^* \right\| \right\| = s_n$$

$$\therefore d(A) = s_n$$