

$$\|x\|_p = \left( \sum_{i=1}^n |x_i|^p \right)^{1/p} \quad \text{"EASY" TO COMPUTE}$$

$$\|A\|_p = \max_{\|x\|_p=1} \|Ax\|_p \quad \text{NP-HARD}$$

Exceptions

$$\|A\|_1, \|A\|_2, \|A\|_\infty$$

$$\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^m |A_{ij}|$$

$$\|A\|_\infty = \max_{1 \leq i \leq m} \sum_{j=1}^n |A_{ij}|$$

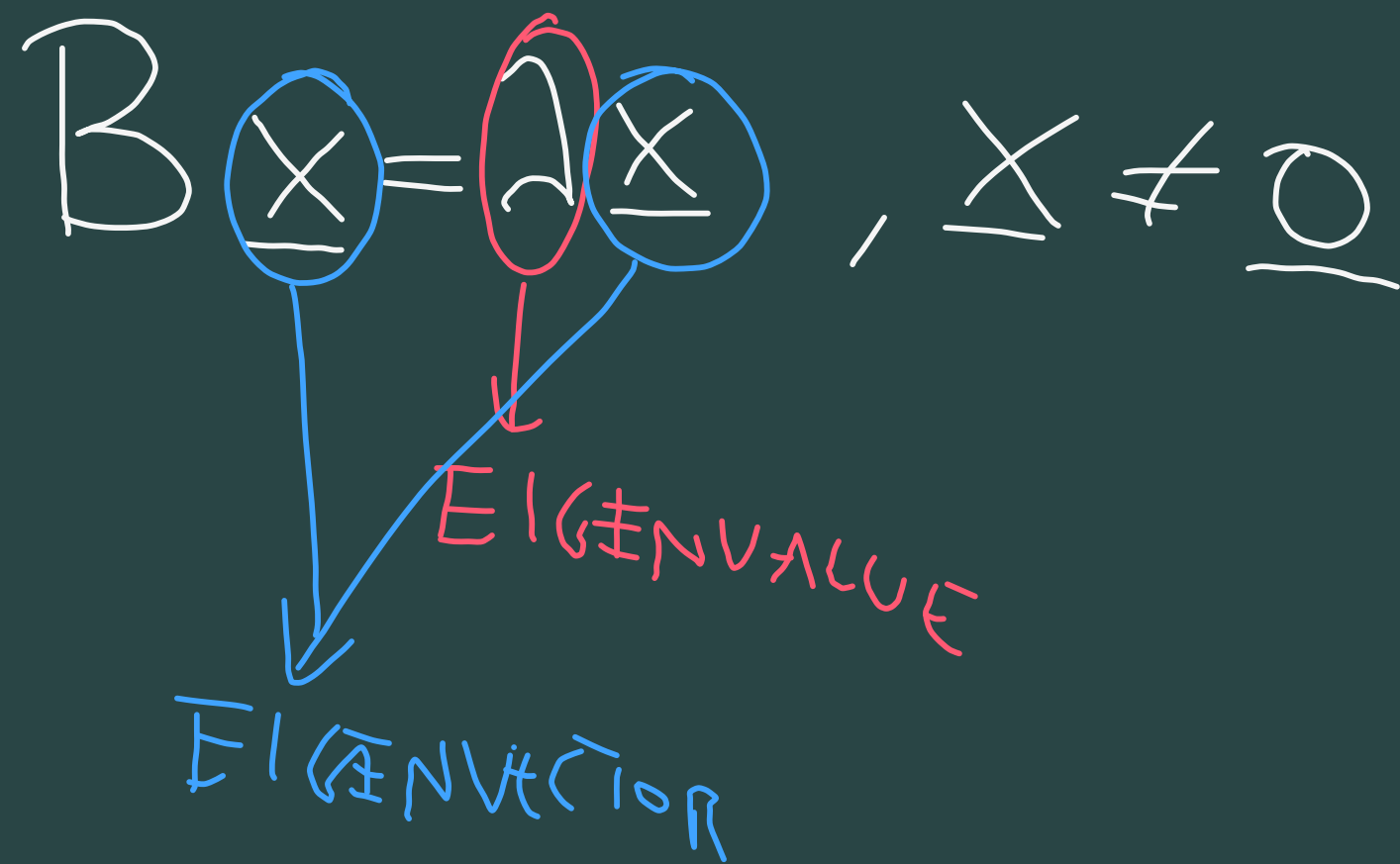
$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

$$\|A\|_1 = \max\{4, 6\} = 6$$

$$\|A\|_\infty = \max\{3, 7\} = 7$$

## LEMMA

Let  $A \in \mathbb{R}^{m \times n}$  and let  $\lambda$  be the largest eigenvalue of  $A^T A$ . Then,  $\|A\|_2 = \sqrt{\lambda}$ .



$A^T A$  is positive semidefinite

$$A^T A \underline{x} = \lambda \underline{x} \quad \text{CLAIM}$$

$$\lambda \geq 0$$

$$\underline{x}^T A^T A \underline{x} = \lambda \underline{x}^T \underline{x}$$

$$\|A \underline{x}\|_2^2 = \lambda \|\underline{x}\|_2^2 \Rightarrow \lambda = \frac{\|A \underline{x}\|_2^2}{\|\underline{x}\|_2^2} \geq 0$$

In the statement,  $B = A^T A$

$$B^T = (A^T A)^T = A^T (A^T)^T = A^T A = B$$

$$\underline{y}^T \underline{y} = \sum_{i=1}^n y_i^2 = \|\underline{y}\|_2^2$$

Proof

$$\|A\|_2 = \max_{x \neq 0} \frac{\|Ax\|_2}{\|x\|_2}$$

$$\|A\|_2^2 = \max_{x \neq 0} \frac{\|Ax\|_2^2}{\|x\|_2^2} =$$

$$\max_{x \neq 0} \frac{x^T A^T A x}{x^T x}$$

By the "spectral theorem", every

symmetric  $B = Q \Lambda Q^T$  where  $\Lambda \in \mathbb{R}^{n \times n}$  that contain the e-values

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and  $Q \in \mathbb{R}^{n \times n}$  is orthogonal,  $Q^T Q = I = Q Q^T$

Hence, there are such  $Q, \Lambda$  st.  $A^T A = Q \Lambda Q^T$

$$\|A\|_2^2 = \max_{x \neq 0} \frac{x^T Q \Lambda Q^T x}{x^T x} \quad \text{let } y = Q^T x, \text{ then}$$

$$\|A\|_2^2 = \max_{x \neq 0} \frac{x^T Q \Lambda Q^T x}{x^T Q Q^T x} = \max_{y \neq 0} \frac{y^T \Lambda y}{y^T y} = \max_{y \neq 0} \frac{\sum_{i=1}^n \lambda_i y_i^2}{\sum_{i=1}^n y_i^2}$$

$$= \max_{\|y\|_2 = 1} \sum_{i=1}^n \lambda_i y_i^2 \leq \lambda$$

and  $\lambda$  is obtainable by some choice of  $y_i$ .

$\lambda_i = \Lambda_{ii} =$   
eigenvalue  
of  $A^T A$

Def

A function  $\|\cdot\|: \mathbb{R}^{m \times n} \rightarrow \mathbb{R}$  is a norm if

①  $\|A\| \geq 0$  and  $\|A\| = 0 \Leftrightarrow A = 0$  for all  $A \in \mathbb{R}^{m \times n}$

②  $\|A+B\| \leq \|A\| + \|B\|$  for all  $A, B \in \mathbb{R}^{m \times n}$

③  $\|\alpha A\| = |\alpha| \|A\|$  for all  $\alpha \in \mathbb{R}, A \in \mathbb{R}^{m \times n}$

# THEOREM

Every operator norm is a norm.

Proof

①  $\|A\| = \max_{\|x\|=1} \|Ax\| \geq 0$ . Moreover,  $\|A\|=0 \Leftrightarrow Ax=0$  for all  $x$

$\Leftrightarrow A=0$

③  $\|\alpha A\| = \max_{\|x\|=1} \|\alpha Ax\| = \max_{\|x\|=1} |\alpha| \|Ax\| = |\alpha| \max_{\|x\|=1} \|Ax\| = |\alpha| \|A\|$



$$\textcircled{2} \quad \|A+B\| = \max_{\|x\|=1} \|(A+B)x\| = \max_{\|x\|=1} \|Ax+Bx\|$$

$$\leq \max_{\|x\|=1} (\|Ax\| + \|Bx\|) \leq \max_{\|x\|=1} \|Ax\| + \max_{\|x\|=1} \|Bx\|$$

$$= \|A\| + \|B\|$$

Example of norms that are not operator norms

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$$\|A\|_F^2 = \sum_{i=1}^m \sum_{j=1}^n |A_{ij}|^2$$

$$\|A\|_{\max} = \max_{i,j} |A_{ij}|$$

$$A \underline{x} = \underline{b}$$

$$\underline{y} = \underline{x} + (\delta x)$$

$$(A + (\delta A)) \underline{y} = (\underline{b} + (\delta b))$$

$$(A + (\delta A)) (\underline{x} + (\delta x)) = \underline{b} + (\delta b)$$

$$\cancel{A} \underline{x} + (\delta A) \underline{x} + A (\delta x) + (\delta A) (\delta x) = \cancel{\underline{b}} + \delta \underline{b}$$

$$(A + \delta A) (\delta x) = \delta \underline{b} - (\delta A) \underline{x}$$

# THEOREM

Let  $\|\cdot\|$  be an operator norm. Then, if  $A$  is invertible and

$$\|\delta A\| < \frac{1}{\|A^{-1}\|}$$

then  $A + \delta A$  is invertible.

$N(A + \delta A) = \{0\} \iff A + \delta A$  is invertible

$$\frac{\|(A + \delta A)x\|}{\|x\|} \geq \frac{\|Ax\| - \|(\delta A)x\|}{\|x\|} = \frac{\|Ax\|}{\|x\|} - \frac{\|(\delta A)x\|}{\|x\|}$$

$$\geq \min_{x \neq 0} \frac{\|Ax\|}{\|x\|} - \max_{x \neq 0} \frac{\|(\delta A)x\|}{\|x\|}$$

$$A \underline{x} = \underline{y} \Leftrightarrow \underline{x} = A^{-1} \underline{y}$$

$$\min_{\underline{x} \neq 0} \frac{\|A \underline{x}\|}{\|\underline{x}\|} = \min_{\underline{y} \neq 0} \frac{\|\underline{y}\|}{\|A^{-1} \underline{y}\|} = \frac{1}{\max_{\underline{y} \neq 0} \frac{\|A^{-1} \underline{y}\|}{\|\underline{y}\|}} = \frac{1}{\|A^{-1}\|}$$

$$\frac{\|(A + \delta A) \underline{x}\|}{\|\underline{x}\|} \geq \frac{1}{\|A^{-1}\|} - \|\delta A\| > 0$$

$$\Rightarrow \forall \underline{x} \neq 0 \text{ then } (A + \delta A) \underline{x} \neq 0 \Rightarrow N(A + \delta A) = \{0\}$$

## COROLLARY

Under the same assumptions,

$$\|(A + \delta A)^{-1}\| \leq \frac{\|A^{-1}\|}{1 - \|\delta A\| \|A^{-1}\|}$$

Proof

$$\|(A + \delta A)^{-1}\| = \| \bar{A}^{-1} (I + (\delta A) \bar{A}^{-1})^{-1} \|$$

$$\leq \| \bar{A}^{-1} \| \cdot \| (I + (\delta A) \bar{A}^{-1})^{-1} \|$$

$$\max_{\underline{x} \neq 0} \frac{\| (I + (\delta A) \bar{A}^{-1}) \underline{x} \|}{\| \underline{x} \|} = \max_{\underline{y} \neq 0} \frac{\| \underline{y} \|}{\| \underline{y} + \delta A \bar{A}^{-1} \underline{y} \|}$$

$$(I + \delta A \cdot \bar{A}^{-1}) \underline{x} = \underline{y}$$

$$\underline{x} = (I + \delta A \cdot \bar{A}^{-1})^{-1} \underline{y}$$

$$(XY)^{-1} = Y^{-1} X^{-1}$$

$$\begin{aligned} (I + \delta A) \bar{A}^{-1} A &= \\ &= A + \delta A \cdot \bar{A}^{-1} A = A + \delta A \end{aligned}$$



$$\|y + \delta A A^{-1} y\| \geq \|y\| - \|\delta A \cdot A^{-1} y\|$$

$$\geq \|y\| - \|\delta A\| \cdot \|A^{-1}\| \cdot \|y\|$$

$$= \|y\| (1 - \|\delta A\| \cdot \|A^{-1}\|)$$

$$\begin{aligned} \left\| \left( I + (\delta A) A^{-1} \right)^{-1} \right\| &\leq \max_{y \neq 0} \frac{\|y\|}{\|y\| (1 - \|\delta A\| \|A^{-1}\|)} \\ &= \frac{1}{1 - \|\delta A\| \|A^{-1}\|} \end{aligned}$$

