

CPSC 303: ( $p \geq 1$ , especially  $p=1, 2, \infty$ )

$\text{cond}_p(A)$  (definition)

$\max_{\hat{b} \neq \vec{b}}$

$$\frac{\text{Rel}_p(\hat{x}, x)}{\text{Rel}_p(\hat{b}, b)}$$

$$\begin{cases} A\vec{x} = \vec{b} \\ A\hat{x} = \hat{b} \end{cases}$$

$$= \max_{\hat{b} \neq \vec{b}} \frac{\text{Rel}_p(A^{-1}\hat{b}, A^{-1}\vec{b})}{\text{Rel}_p(\hat{b}, \vec{b})}$$

$\text{Rel}_p$

also "Black Box"

Magic!

= maximum loss of relative  $p$ -norm error in "solving  $A\vec{x} = \vec{b}$ "

Theorem:  $\text{cond}_p(A) = \|A\|_p \|A^{-1}\|_p$  ( $p$ -norm of matrices)

fairly easy to approximate

Section 5.8 of [A&G]:  $n = 10, 100$ , or  $10^4, 10^5, \dots$

Us:  $n = 2, 3$ , (CL) 4

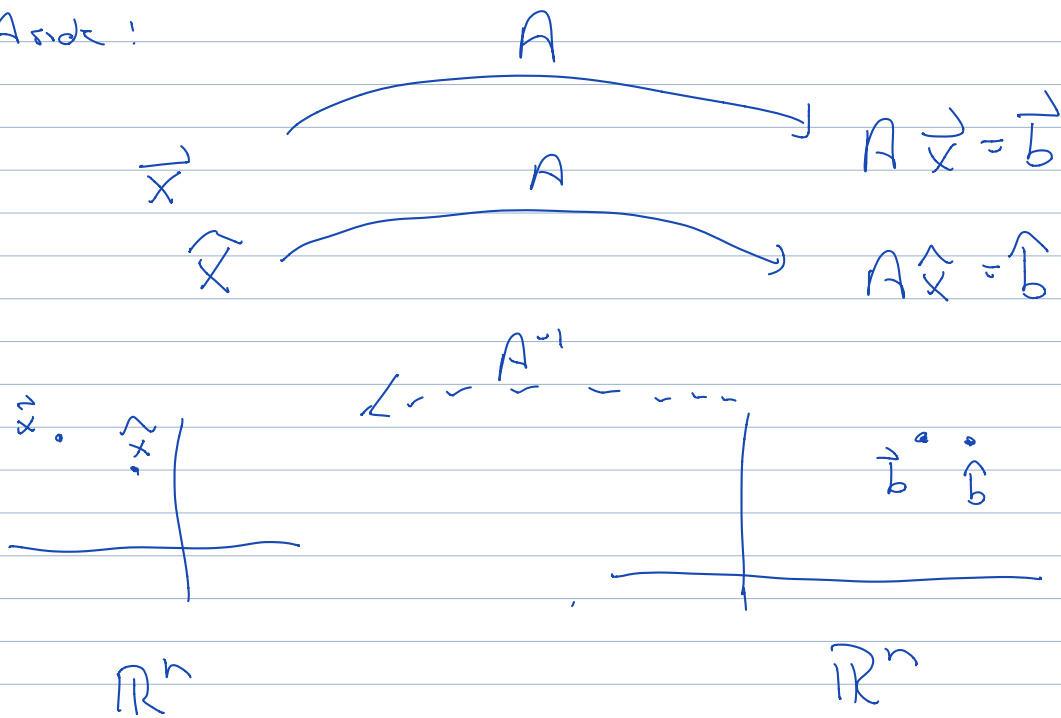
Fact:  $\text{cond}_p(A) = \|A\|_p \|A^{-1}\|_p \geq 1$

We fix  $p \geq 1$  ( $p = \infty, 1, 2$ ),  $A$  is  $n \times n$

$$\|A\|_p = \max_{\vec{v} \neq 0} \frac{\|A\vec{v}\|_p}{\|\vec{v}\|_p}$$

how much  $A$  "stretches a vector" in  $p$ -norm

Aside:



Good news:

Say  $A = \begin{bmatrix} \alpha & \beta \\ \gamma & \delta \end{bmatrix}$ ,  $M = \max(|\alpha|, |\beta|, |\gamma|, |\delta|)$

then

$M \leq \|A\|_p \leq 2M$  (any  $p \geq 1$ )

*block-box magic then*

Last time:

$\begin{bmatrix} ? & ? \\ 27 & ? \end{bmatrix} \underbrace{\begin{bmatrix} 1 \\ 0 \end{bmatrix}}_{p\text{-norm} = 1} = \underbrace{\begin{bmatrix} ? \\ 27 \end{bmatrix}}_{p\text{-norm} \geq 27}$

Exact 1-norm, 2-norm,  $\infty$ -norm?

$\| \begin{bmatrix} \alpha & \beta \\ \gamma & \delta \end{bmatrix} \|_{\infty} = \max(|\alpha| + |\beta|, |\gamma| + |\delta|)$

$\| \quad \quad \quad \|_1 = \max(|\alpha| + |\gamma|, |\beta| + |\delta|)$

||

$$\|_2 = \sqrt{\text{largest eigenvalue of } A^T A}$$



I initially goofed until corrected by student



Aside!

$$\left\| \begin{bmatrix} \alpha & \beta \\ \gamma & \delta \end{bmatrix} \right\|_{\infty} = ?$$

$$\max_{\vec{v} = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \neq \vec{0}}$$

$$\frac{\left\| \begin{bmatrix} \alpha & \beta \\ \gamma & \delta \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right\|_{\infty}}{\left\| \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right\|_{\infty}}$$

True for any norm

$$= \max_{\left\| \vec{v} = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right\|_{\infty} = 1} \left\| \begin{bmatrix} \alpha & \beta \\ \gamma & \delta \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right\|_{\infty}$$

$$\max_{\left\| \vec{v} = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right\|_{\infty} \leq 1} \left\| \begin{bmatrix} \alpha & \beta \\ \gamma & \delta \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right\|_{\infty}$$

$$= \max_{|v_1|, |v_2| \leq 1} \left( \max \left( \alpha v_1 + \beta v_2, \gamma v_1 + \delta v_2 \right) \right)$$

$\uparrow \quad \uparrow$   
 $|v_1| \leq 1 \quad |v_2| \leq 1$

Say  $A = \begin{bmatrix} 5 & -4 \\ 1 & 2 \end{bmatrix}$  ;  $A \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 9 \\ -1 \end{bmatrix} \leftarrow \max_{\infty} 9$

$A \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix} \leftarrow \text{"} < 9$

$A \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -5 \\ 3 \end{bmatrix} \leftarrow 9$

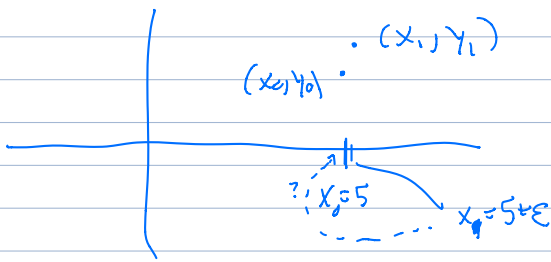
$A \begin{bmatrix} -1 \\ -1 \end{bmatrix} = \begin{bmatrix} -9 \\ -3 \end{bmatrix} \leftarrow 9$

So

$$\|A\|_{\infty} = \max(|5| + |-4|, |1| + |2|)$$

Interpolation: Tangent Line

e.g.  $x_0 = 5$ ,  $x_1 = 5 + \epsilon$ ,  $\epsilon$  small



$$p(x) = c_0 + c_1 x$$

$$c_0 + 5c_1 = y_0$$

$$c_0 + (5 + \epsilon)c_1 = y_1$$

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

$$\begin{bmatrix} 1 & 5 \\ 1 & 5 + \epsilon \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \end{bmatrix} = \begin{bmatrix} y_0 \\ y_1 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 5 \\ 1 & 5+\epsilon \end{bmatrix},$$

$$A^{-1} = \frac{1}{1 \cdot (5+\epsilon) - 1 \cdot 5} \begin{bmatrix} 5+\epsilon & -5 \\ -1 & 1 \end{bmatrix}$$

$$= \frac{1}{\epsilon} \begin{bmatrix} \phantom{5+\epsilon} & \phantom{-5} \\ \phantom{-1} & \phantom{1} \end{bmatrix}$$

$$\text{cond}_p(A) = \underbrace{\left\| \begin{bmatrix} 1 & 5 \\ 1 & 5+\epsilon \end{bmatrix} \right\|_p}_{\substack{\text{max entry} \\ 5 \text{ or } |5+\epsilon|}} \underbrace{\left\| \frac{1}{\epsilon} \begin{bmatrix} 5+\epsilon & -5 \\ -1 & 1 \end{bmatrix} \right\|_p}_{\frac{1}{\epsilon} \text{max}(5, 5+\epsilon) \leq \text{norm}}$$

$$\text{max}(5, 5+\epsilon) \leq \text{norm} \leq 2 \text{max}(5, 5+\epsilon) \leq \frac{2}{\epsilon} \text{max}(5, 5+\epsilon)$$

As  $\epsilon \rightarrow 0$   $\text{cond} \begin{pmatrix} 1 & 5 \\ 1 & 5+\epsilon \end{pmatrix}$  roughly  $\frac{\text{const} \approx 25}{\epsilon}$   $4 \cdot (5.01)^2$   
with a factor of 4

$0 < \epsilon < 0.1$ :

$$5 \leq \text{first norm} \leq 2 \cdot 5.01 \qquad \frac{5}{\epsilon} \leq \text{second norm} \leq 2 \cdot \frac{5.01}{\epsilon}$$

$$\frac{25}{\epsilon} \leq \text{cond}(A) \leq \frac{4 \cdot (5.01)^2}{\epsilon}$$

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$$\begin{cases} | c_0 + 5 c_1 = y_0 \\ | c_0 + (5+\epsilon) c_1 = y_1 \end{cases}$$

$$\left\{ \begin{array}{l} 1 \cdot c_0 + 5c_1 = y_0 \\ \varepsilon c_1 = y_1 - y_0 \end{array} \right. \quad \left. \begin{array}{l} \text{cond \#} \\ \text{good or} \\ \text{bad as} \\ \varepsilon \rightarrow 0 \end{array} \right.$$

$$\left\{ \begin{array}{l} 1 \cdot c_0 + 5c_1 = y_0 \\ c_1 = (y_1 - y_0) / \varepsilon \end{array} \right.$$