

I Error vs Residual: Consider the Linear System

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

- A. Let:
- ① \vec{x} be an exact solution
 - ② $\tilde{\vec{x}}$ be a solution "infused" with error

- B. DEFINITION:
- ① $\vec{e} = \tilde{\vec{x}} - \vec{x}$ "error"
 - ② $\vec{r} = A\tilde{\vec{x}} - \vec{b}$ "residual"

C. Example: Consider THE SYSTEM

$$\begin{bmatrix} 1 & -2 \\ -0.99 & 1.99 \end{bmatrix} \vec{x} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\vec{x} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \quad \text{let } \tilde{\vec{x}} = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

$$\vec{e} = \tilde{\vec{x}} - \vec{x} = \begin{bmatrix} -2 \\ -1 \end{bmatrix} \quad \text{"big error"} \Rightarrow \|\vec{e}\|_{\infty} = 2$$

$$\vec{r} = A\tilde{\vec{x}} - \vec{b} = \begin{bmatrix} -1 \\ -0.99 \end{bmatrix} - \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ -0.01 \end{bmatrix} \Rightarrow \|\vec{r}\|_{\infty} = 0.01$$

small residual

$$\Rightarrow \frac{\|\vec{e}\|_{\infty}}{\|\vec{r}\|_{\infty}} = 200!! \quad \text{"error is 200 times bigger than residual"}$$

II. Theorem: Let: ① A be non singular matrix
 ② \tilde{x} be approximate soln to $Ax = b$
 ③ $\vec{r} = A\tilde{x} - b$
 ④ $\vec{e} = \tilde{x} - \hat{x}$ ⑤ $\|\cdot\|$ be any natural matrix norm

Then:

$$\textcircled{1} \frac{1}{\|A\|} \|\vec{r}\| \leq \|\vec{e}\| \leq \|A\| \|\vec{r}\|$$

$$\textcircled{2} \frac{1}{\|A\| \|A^{-1}\|} \frac{\|\vec{r}\|}{\|b\|} \leq \frac{\|\vec{e}\|}{\|\hat{x}\|} \leq \|A\| \|A^{-1}\| \frac{\|\vec{r}\|}{\|b\|}$$

note: $\|b\| \neq 0, \|\hat{x}\| \neq 0$

A. Proof \rightarrow

(as part 1)

① $\vec{r} = A\tilde{x} - b = A\tilde{x} - A\hat{x} = A(\tilde{x} - \hat{x}) = A\vec{e}$

② $\therefore \vec{e} = A^{-1}\vec{r}$

③ $\therefore \|\vec{e}\| = \|A^{-1}\vec{r}\| \leq \|A^{-1}\| \|\vec{r}\|$ (prop 5 mat norm)

④ $\|\vec{r}\| = \|A\vec{e}\| \leq \|A\| \|\vec{e}\| \Rightarrow \frac{1}{\|A\|} \|\vec{r}\| \leq \|\vec{e}\|$

⑤ $\frac{1}{\|A\|} \|\vec{r}\| \leq \|\vec{e}\| \leq \|A^{-1}\| \|\vec{r}\|$ QED

III. Definition: Condition Number \rightarrow $\boxed{K(A) = \|A\| \|A^{-1}\|}$

Example: $A = \begin{bmatrix} 1 & -2 \\ -99 & 199 \end{bmatrix}$ $A^{-1} = \begin{bmatrix} 199 & 200 \\ 99 & 100 \end{bmatrix}$

$$\|A\|_{\infty} = \max(1+2, 99+199) = 3$$

$$\|A^{-1}\|_{\infty} = \max(199+200, 99+100) = 399$$

$$K(A) = (3)(399) = \boxed{1197}$$

IV. Rewrite Part I of Theorem

$$\frac{1}{k(A)} \frac{\|\vec{r}\|}{\|\vec{b}\|} \leq \frac{\|\vec{e}\|}{\|\vec{x}\|} \leq k(A) \frac{\|\vec{r}\|}{\|\vec{b}\|}$$

Therefore ① Small $k(A)$ provides a good measure of error in an approximate solution, ② large $k(A)$ provides bad.

A) I.E. in above example

$$\frac{1}{1197} \frac{\|\vec{r}\|}{\|\vec{b}\|} \leq \frac{\|\vec{e}\|}{\|\vec{x}\|} \leq 1197 \frac{\|\vec{r}\|}{\|\vec{b}\|}$$

↑ i.e. very bad indicator

B) Lower Bound For $k(A)$

$$1 = \|I\| = \|AA^{-1}\| \leq \|A\| \|A^{-1}\| = k(A)$$