

A

LU decomposition

$$A \rightarrow \begin{matrix} M_1 & a_{11} \\ & M_2 & \tilde{a}_{22} \\ & & \vdots \end{matrix}$$

$$M_1 = \begin{pmatrix} 1 & & & & \\ -\frac{a_{21}}{a_{11}} & 1 & & & \\ & & \ddots & & \\ -\frac{a_{31}}{a_{11}} & & & 1 & \\ & & & & \ddots \\ -\frac{a_{n1}}{a_{11}} & & & & & 1 \end{pmatrix}$$

$$a_{11} \neq 0$$

$$A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

B

$$A = \left(\begin{array}{ccc|c} 0 & 1 & 5 & 5 \\ 1 & 0 & 2 & 25 \end{array} \right) \Rightarrow A' = \left(\begin{array}{cc} 1 & 0 \\ 0 & 1 \end{array} \right)$$

$$\begin{cases} x + 2y + 3z = 10 \\ 2x + 3y + 2z = 25 \\ 3x + 5y + 7z = 7 \end{cases}$$

$$\begin{pmatrix} 1 & 2 & 3 \\ 2 & 3 & 2 \\ 3 & 5 & 7 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 10 \\ 25 \\ 7 \end{pmatrix}$$

⇒ changing rows does NOT affect a solution
 partial pivoting

is a process of changing order of rows to have the biggest possible a_{ii}

Theorem 3.1

Any $n \times n$ matrix can be
pivoted, so that LU
decomposition of
matrix exists.

Norm of a vector

D

$$\underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

$$\|x\|_p = \left[\sum_{k=1}^n |x_k|^p \right]^{1/p}$$

$$\|A\| = \max_{\underline{x} \neq 0} \frac{\|A \cdot x\|}{\|x\|}$$

$$\text{If } \|x\| = 1, \|A\| = \max_{x \neq 0} \|Ax\|$$

E

Condition number
of a matrix.

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

Cond(A)

Cond(A, p) ← norm

- $A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \Rightarrow A^{-1} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$
 $\text{Cond}_1(A) = 1$

- $A = \begin{pmatrix} d_1 & 0 \\ 0 & d_2 \end{pmatrix}; A^{-1} = \begin{pmatrix} 1/d_1 & 0 \\ 0 & 1/d_2 \end{pmatrix}$

$$\|A\|_\infty = \max(|d_1|, |d_2|)$$

$$\|A^{-1}\|_\infty = \max(1/|d_1|, 1/|d_2|)$$

$$\|A\|_{\infty} = \max(|d_1|, |d_2|)$$

$$\|A^{-1}\|_{\infty} = \max\left(\frac{1}{|d_1|}, \frac{1}{|d_2|}\right)$$

$$= \frac{1}{\min(|d_1|, |d_2|)}$$

$$\kappa(A) = \frac{\max(|d_1|, |d_2|)}{\min(|d_1|, |d_2|)}$$

$|d_1|$ is small

$$0 < d_1 \ll 1$$

$$d_2 \approx 1$$

$$A = \begin{pmatrix} d_1 & 0 \\ 0 & d_2 \end{pmatrix}$$

- $$A = \begin{pmatrix} d_1 & & & & 0 \\ & d_2 & & & \\ & & d_3 & & \\ & & & \dots & \\ & & & & d_n \end{pmatrix}$$

G

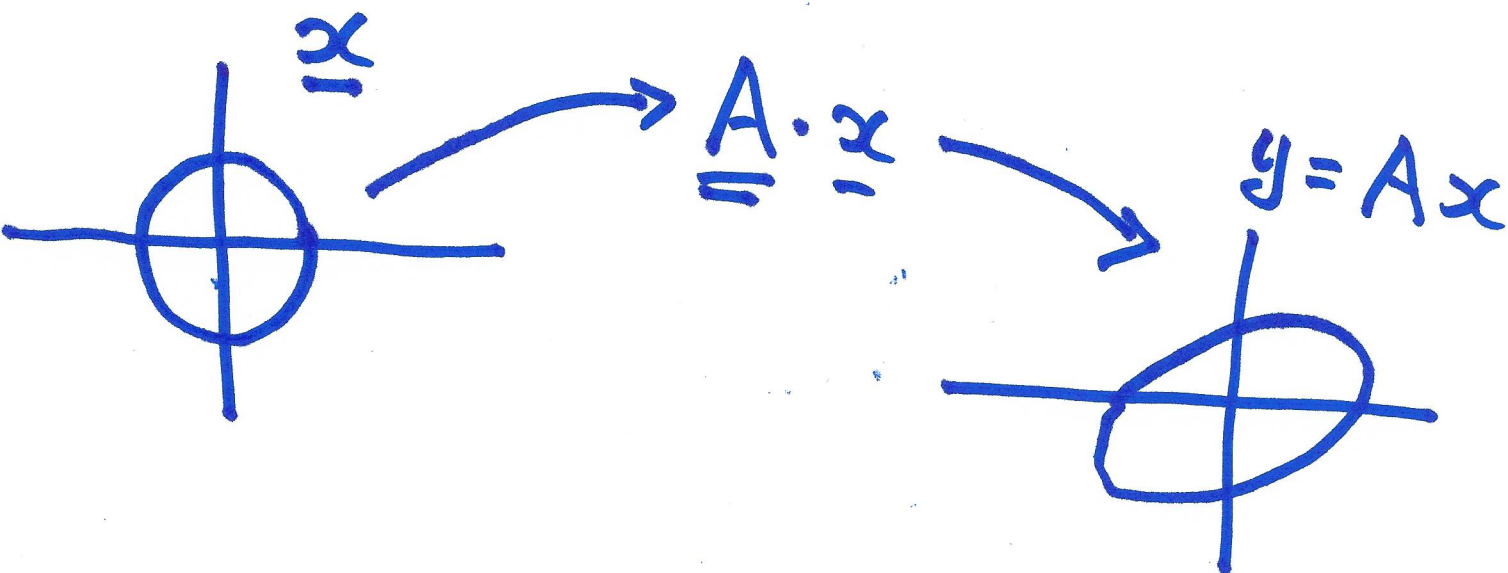
$$A^{-1} = \begin{pmatrix} 1/d_1 & & & & 0 \\ & & & & \\ & & 1/d_2 & & \\ & & & \dots & \\ & & & & 1/d_n \end{pmatrix}$$

$$\kappa(A) = \frac{\max_{i=1..n} |d_i|}{\min_{i=1..n} |d_i|}$$

$$A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

H

$$\|x\|_2 = 1$$



$\|x\|_2$

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

$$\kappa_2(A) = \frac{\max |y|}{\min |y|}$$

I

$$\|A\|_1 = \max_j \sum_{i=1}^n a_{ij} \quad \text{- max column sum}$$

$$\|A\|_\infty = \max_j \sum_{i=1}^n a_{ij} \quad \text{- max row sum}$$

① $\forall A, \rho(A) \geq 1$

② IF $A = I, \rho(A) = 1$

③ IF A is a matrix,
 γ is a scalar

$$\rho(\gamma A) = \gamma \rho(A)$$

④ IF A is singular
($\det(A) = 0$, or A^{-1} does not exist)

then $\rho(A) = \infty$

⑤ IF $A = d_i \delta_{ij}$ - diagonal, $\rho = \max |d_i|$

9

$$\text{solve } Ax = b,$$

$$\text{solve } A(\hat{x}) = b + \Delta b$$

$$\hat{x} = x + \Delta x$$

$$\Delta x = \hat{x} - x$$

$$\frac{\|\Delta x\|}{\|x\|} \leq \frac{\|A^{-1}\| \cdot \|\Delta b\|}{\|b\|}$$

$$A \hat{x} = A \cdot (x + \Delta x) =$$

$$= \underline{A \cdot x} + A \cdot \Delta x = \underline{b + \Delta b}$$

$$A \cdot \Delta x = \Delta b, \Delta x = A^{-1} \cdot \Delta b$$

$$\|\Delta x\| = \|A^{-1} \Delta b\| \leq \|A^{-1}\| \|\Delta b\|$$

$$\|b\| = \|A \cdot x\| \leq \|A\| \cdot \|x\|$$

$$\frac{1}{\|x\|} \leq \frac{\|A\|}{\|b\|}$$

$$\frac{\|\Delta x\|}{\|x\|} \leq \kappa(A) \frac{\|\Delta b\|}{\|b\|}$$

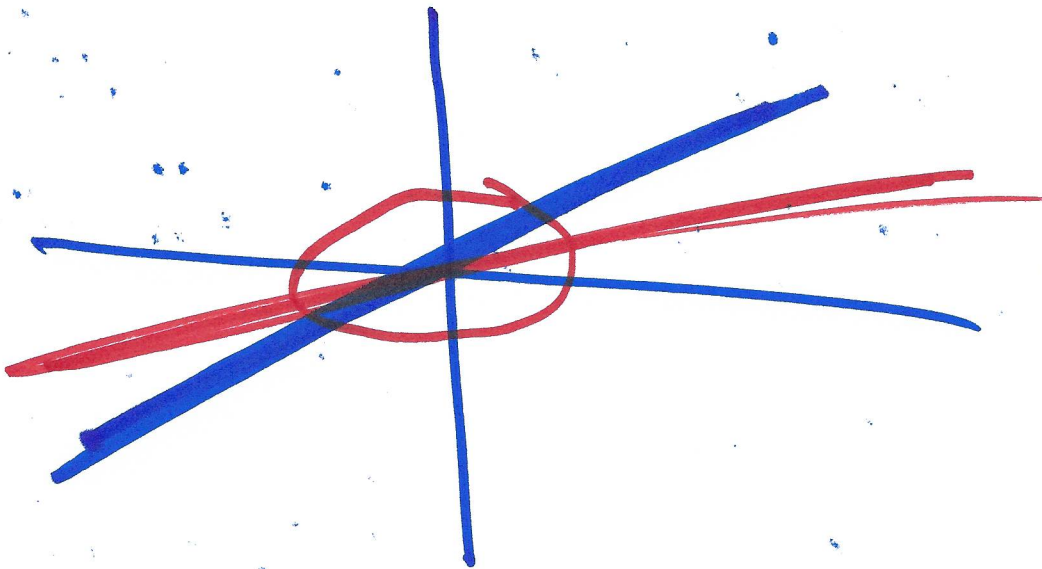
$$Ax = b$$

K

$$(A + \Delta A)\hat{x} = b$$

$$\frac{\|b - \hat{x}\|}{\|\hat{x}\|} \leq \kappa(A) \frac{\|\Delta A\|}{\|A\|}$$

$$\frac{\|\Delta x\|}{\|x\|} \leq \kappa(A) \epsilon_{\text{mach}}$$



Residual

$$\frac{\|\Delta x\|}{\|x\|} \leq \kappa(A) \frac{\|r\|}{\|b\|}$$

Solve $Ax = b$

computer gives \hat{x}

$$\Delta x = -\hat{x} + x \text{ - error}$$

$$r = b - A\hat{x}; \quad A\hat{x} = b - r$$

$$\hookrightarrow \text{residual} \quad \hat{x} = A^{-1}b - A^{-1}r$$

$$\Delta x = x - \hat{x} = A^{-1}b - A^{-1}b + A^{-1}r$$

$$\Delta x = A^{-1}r = A^{-1}r$$

$$\|\Delta x\| = \|A^{-1} \cdot r\| \leq \|A^{-1}\| \cdot \|r\|$$

$$\frac{1}{\|x\|} \leq \|A\| / \|b\|$$

$$\frac{\|\Delta x\|}{\|x\|} \leq \frac{\|A^{-1}\| \cdot \|A\| \|r\|}{\|b\|}$$

4

Solving nonlinear
equations.

$$f(x, y) = 0$$

$$g(x, y) = 0$$

Simplistic way

$$x_0, y_0$$

1st equation

$$x_1 = x_0 - \frac{f(x_0, y_0)}{\partial f(x_0, y_0) / \partial x}$$

$$y_1 = y_0 - \frac{g(x_0, y_0)}{\partial g(x_0, y_0) / \partial y}$$

N

$$x_{k+1} = x_k - \frac{f(x_k, y_k)}{\partial f(x_k, y_k) / \partial x}$$

$$y_{k+1} = y_k - \frac{g(x_k, y_k)}{\partial g(x_k, y_k) / \partial y}$$