

pre-conditioning
= scaling

do page 58 pp

Sparse matrix



→ wiki

space
curves

$$Ax = b$$

answer x_0

$$b - Ax_0 = \underbrace{\text{residual}}_{r_0} \neq 0$$

$$Ax = r_0$$

iteration

answer x_1

$$x_0 + x_1 + \dots$$

$$x_0 = 0$$

$$Ax = b$$

Iterative Solvers of
 $Ax = b$

$$(L\bar{A} + D)x = b - \underbrace{\bar{A}x_0}_{O(n^2)}$$

back solve for $x \rightarrow$ yields x_1

$$(L\bar{A} + D)x_{i+1} = b - \bar{A}x_i$$

Gauss - Seidel

$$A = \begin{bmatrix} \times & \times & \times \\ \times & \times & \times \\ \times & \times & \times \end{bmatrix} \rightarrow \begin{bmatrix} \times & \times & \times \\ \times & \times & \times \\ \times & \times & \times \end{bmatrix}$$

↑ row index > col index

split not
factor

if A is diagonally dominant
then $x_{i+1} \xrightarrow{i=1,2,\dots} \text{solution}$
converges

$$\{x_i\}_{i=0}^{\infty} \rightarrow x = \begin{bmatrix} \text{---} \\ \text{---} \\ \text{---} \end{bmatrix} = \begin{bmatrix} \text{---} \\ \text{---} \\ \text{---} \end{bmatrix} = [A^{-1}b]$$

Gauss-Seidel

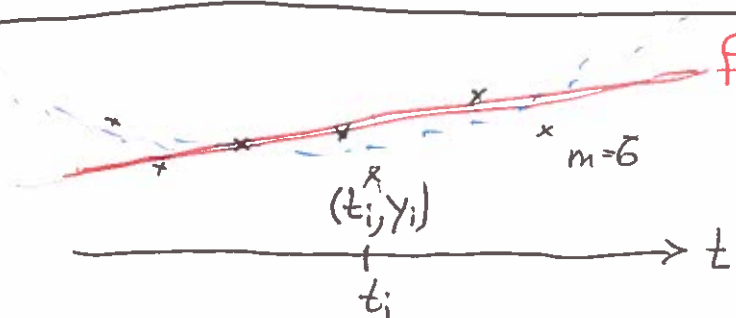
$$\triangle x_{i+1} = b - \triangle x_i$$

Jacobi iteration

$$\parallel x_{i+1} = b - \nabla x_i$$

Successive Over Relaxation

diagonally dominant



$$f(t, x) = \sum_{j=0}^n x_j \phi_j(t)$$

for example

$$f = x_0 t^0 + x_1 t^1 + x_2 t^2 + \dots + x_n t^n$$

$$= \sum_{j=0}^n x_j t^j$$

$n=1$: linear polynomial
 $n=2$: quadratic poly--

min
~~find~~
 $X = (x_0, x_1, \dots, x_n)$

$$\sum_{i=0}^m (y_i - f(t_i, x))^2$$

$$= \min_x \|y - f(x)\|_2^2$$

Linear least squares;

f depends linearly on x

$$f = (x_0 \dots x_n) \begin{pmatrix} \phi_0(t) \\ \vdots \\ \phi_n(t) \end{pmatrix}$$

$$= x_0 \phi_0(t) + \dots + x_n \phi_n(t)$$

NOT

non-linear least squares

$$\sum x_0 e^{x_1 \frac{2\pi}{n} t F_1}$$

$$\min_x \|y - \underbrace{\begin{pmatrix} \phi_0(t_i) \\ \vdots \\ \phi_n(t_i) \end{pmatrix}}_A x\|_2^2$$

Vandermonde

$$\Rightarrow A^T(b - Ax) = 0$$

$$A^T A x = A^T b$$

Normal Equ's