

Revision

System of Linear Equations $Ax = b$

Direct Methods

1. Gauss Elimination (GE)
2. Gauss Jordan
3. LU Decomposition

Computational / Algorithmic complexity

$$A = LU$$

Unique if l_{ii} or $u_{ii} = 1$

- Computational -
- GE \rightarrow All steps including scaling and pivoting
 - Doolittle $\left\{ \begin{array}{l} l_{ii} = 1 \end{array} \right.$
 - Crout - $u_{ii} = 1$
 - Thomas algorithm - tridiagonal
 - Cholesky method \rightarrow symmetric positive definite

Doolittle / Crout Method

Example

Doolittle

$$\begin{matrix} & & \rightarrow 1 \\ 2 & \begin{bmatrix} 3 & 1 \\ 2 & 4 \end{bmatrix} \end{matrix} = \begin{matrix} L & & U \\ \begin{bmatrix} 1 & 0 \\ 2/3 & 1 \end{bmatrix} & & \begin{bmatrix} 3 & 1 \\ 0 & 4 - \frac{2}{3} \end{bmatrix} \\ & & = 10/3 \end{matrix}$$

Crout

$$\begin{matrix} & & \rightarrow 2 \\ 1 & \begin{bmatrix} 3 & 1 \\ 2 & 4 \end{bmatrix} \end{matrix} = \begin{matrix} L & & U \\ \begin{bmatrix} 3 & 0 \\ 2 & 10/3 \end{bmatrix} & & \begin{bmatrix} 1 & 1/3 \\ 0 & 1 \end{bmatrix} \end{matrix}$$

Thomas Algorithm - Tri diagonal

$$\begin{bmatrix} & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \end{bmatrix}$$

$$\begin{matrix} \rightarrow \\ \rightarrow \\ \rightarrow \end{matrix} \begin{bmatrix} d_1 & u_1 & & & \\ l_2 & d_2 & u_2 & & \\ & l_3 & d_3 & u_3 & \\ & & \ddots & \ddots & \ddots \\ & & & l_n & d_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

Define

$$\alpha_1 = d_1$$

$$\beta_1 = b_1$$

$$\lambda_2 = \frac{l_2}{\alpha_1}$$

$$R_1: \alpha_1 x_1 + u_1 x_2 = \beta_1$$

$$R_2: l_2 x_1 + d_2 x_2 + u_2 x_3 = b_2$$

$$R_2 \rightarrow R_2 - \lambda_2 R_1$$

$$x_2 (d_2 - \lambda_2 u_1) + u_2 x_3 = b_2 - \lambda_2 \beta_1$$

Define

$$\alpha_2 = d_2 - \lambda_2 u_1$$

$$\beta_2 = b_2 - \lambda_2 \beta_1$$

$$\Rightarrow \alpha_2 x_2 + u_2 x_3 = \beta_2$$

$$\lambda_3 = \frac{l_3}{\alpha_2} \quad l_3 x_2 + d_3 x_3 + u_3 x_4 = b_3$$

$$\lambda_i = \frac{l_i}{\alpha_{i-1}}$$

$$\alpha_i = d_i - \lambda_i u_{i-1}$$

$$\beta_i = b_i - \lambda_i \beta_{i-1}$$

$$x_i = \frac{\beta_i - u_i x_{i+1}}{\alpha_i}$$

$$i = n-1 \leftarrow \alpha_i x_i + u_i x_{i+1} = \beta_i$$

$$i = n$$

$$\alpha_n x_n = \beta_n \Rightarrow x_n = \beta_n / \alpha_n$$

$$GE : O(2/3n^3)$$

$$Thomas : O(n)$$

Example

$$\begin{bmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 8 \\ 8 \end{bmatrix}$$

$$\alpha_1 = a_1 \quad \beta_1 = b_1 \quad \lambda_i = \frac{d_i}{\alpha_{i-1}}$$

$$\alpha_i = d_i - \lambda_i u_{i-1}$$

$$\beta_i = b_i - \lambda_i \beta_{i-1}$$

$$x_n = \beta_n / \alpha_n$$

$$x_i = \frac{\beta_i - u_i x_{i+1}}{\alpha_i}$$

i	l	d	u	b	λ	α	β	x
1		2	1	4		2	4	$x_1 = 1$
2	1	2	1	8	$\frac{d_2}{\alpha_1} = 0.5$	$2 - 0.5 \times 1 = 1.5$	$8 - 0.5 \times 4 = 6$	$x_2 = \frac{6 - 1 \times 3}{3/2} = 2$
3	1	2		8	$\frac{d_3}{\alpha_2} = \frac{1}{1.5} = 2/3$	$4/3$	4	$x_3 = 4 / (4/3) = 3$

Cholesky Decomposition

Remark 1

$$A = LU$$

If l_{ii} or $u_{ii} = 1 \rightarrow$ decomposition is unique

Can we write U such that its diagonal elements are unity

$$\begin{bmatrix} d_1 & u_{12} & u_{13} & \dots & u_{1n} \\ & d_2 & u_{23} & \dots & u_{2n} \\ & & d_3 & \dots & u_{3n} \\ & & & \ddots & \\ & & & & d_n \end{bmatrix} = \begin{bmatrix} d_1 & & & & \\ & d_2 & & & \\ & & \ddots & & \\ & & & d_n & \end{bmatrix} \begin{bmatrix} 1 & u_{12}/d_1 & u_{13}/d_1 & \dots & u_{1n}/d_1 \\ & 1 & u_{23}/d_2 & \dots & u_{2n}/d_2 \\ & & \ddots & \ddots & \\ & & & 1 & \\ & & & & 1 \end{bmatrix}$$

$$U = D \tilde{U}$$

$\tilde{U}_{ii} = 1$

$$A = LDU$$

$$l_{ii} = u_{ii} = 1$$

Remark 2 - Symmetric Positive Definite Matrix

Problem \Rightarrow minimization

$$f(x)$$

Minimum
necessary — $\frac{df}{dx} = 0$

sufficient $\frac{d^2f}{dx^2} > 0$

$f(x_1, x_2)$
necessary $\frac{\partial f}{\partial x_1} = \frac{\partial f}{\partial x_2} = 0$

positive definite system $\Leftrightarrow \begin{vmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_2^2} \end{vmatrix} > 0$

$$(x_1, x_2) \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$x^T A x > 0$ — positive definite mds

$$a x_1^2 + 2b x_1 x_2 + c x_2^2 > 0$$

It is possible : $a > 0, c > 0$
 $ac > b^2$

For n dimensional function $f(x_1, x_2, \dots, x_n)$

$$a_{ij} = \frac{\partial^2 f}{\partial x_i \partial x_j} \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \text{ symmetric}$$

A should be positive definite

Symmetric matrix

$$A^T = A$$

$$A = LDU$$

$$A = A^T = U^T D L^T$$

$$\Rightarrow L = U^T$$

$$U = L^T$$

$$\Rightarrow \boxed{A = L D L^T} \quad - \text{symmetric matrix}$$

If A is positive definite $\therefore D$ will be positive

$$A = L \tilde{D}^{\frac{1}{2}} \tilde{D}^{\frac{1}{2}} L^T$$

$$\boxed{A = L_c L_c^T} \quad - \text{Cholesky decomposition}$$

$$\boxed{L_c = L \tilde{D}^{\frac{1}{2}}}$$

cholesky factor

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix} = \begin{bmatrix} l_{11} & & & & \\ l_{21} & l_{22} & & & \\ l_{31} & l_{32} & l_{33} & & \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & \dots & l_{nn} \end{bmatrix} \begin{bmatrix} l_{11} & l_{21} & \dots & l_{n1} \\ & l_{22} & l_{32} & \dots & l_{n2} \\ & & \ddots & \ddots & \vdots \\ & & & l_{nn} \end{bmatrix}$$

$$a_{ij} = a_{ji}$$

$$l_{11}^2 = a_{11}$$

$$l_{11} = \sqrt{a_{11}}$$

$$l_{j1} = \frac{a_{j1}}{l_{11}}$$

$$l_{jj} = \sqrt{a_{jj} - \sum_{k=1}^{j-1} l_{jk}^2} \quad j=2, \dots, n$$

$$l_{kj} = \left(a_{kj} - \sum_{s=1}^{j-1} l_{js} l_{ks} \right) / l_{jj} \quad k=j+1, \dots, n$$

Example

$$\begin{bmatrix} 4 & 2 & 14 \\ 2 & 17 & -5 \\ 14 & -5 & 83 \end{bmatrix} \rightarrow \begin{bmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{bmatrix} \rightarrow \begin{bmatrix} l_{11} & l_{21} & l_{31} \\ 0 & l_{22} & l_{32} \\ 0 & 0 & l_{33} \end{bmatrix}$$

$$\Rightarrow l_{11}^2 = a_{11} \Rightarrow \boxed{l_{11} = \sqrt{a_{11}}} = \underline{\underline{2}}$$

$$l_{21} l_{11} = a_{21}$$

$$\Rightarrow l_{21} = \frac{a_{21}}{l_{11}} \Rightarrow l_{21} = \frac{2}{2} = \underline{\underline{1}}$$

$$\boxed{l_{j1} = \frac{a_{j1}}{l_{11}}}$$

$$l_{31} = \frac{14}{2} = 7$$

$$\rightarrow L = \begin{bmatrix} 2 & 0 & 0 \\ 1 & 4 & 0 \\ 7 & -3 & 5 \end{bmatrix}$$

$$L^T = \begin{bmatrix} 2 & 1 & 7 \\ 0 & 4 & -3 \\ 0 & 0 & 5 \end{bmatrix}$$

$$l_{21}^2 + l_{22}^2 = a_{22}$$

$$\Rightarrow l_{22} = \sqrt{a_{22} - l_{21}^2} = \sqrt{17 - 1^2} = 4$$

$$\boxed{l_{ij} = \sqrt{a_{ij} - \sum_{k=1}^{j-1} l_{ik}^2}}$$

$$l_{31} l_{21} + l_{32} l_{22} = a_{32}$$

$$\Rightarrow l_{32} = \frac{a_{32} - l_{31} l_{21}}{l_{22}} = \frac{-5 - 7 \times 1}{4} = -3$$

$$l_{kj} = \frac{a_{kj} - \sum_{s=1}^{j-1} l_{js} l_{ks}}{l_{jj}} \quad k=j+1, \dots, n$$

$$l_{33} = \sqrt{83 - 7^2 - 3^2} = 5$$

Error Analysis

Condition number of the problem

$$\left. \begin{matrix} f(x) \\ x + \delta x \end{matrix} \right\} C_p = \frac{|\Delta f / f|}{|\Delta x / x|}$$

System of linear equations

$$A x = b$$

To find x .

$$A + \delta \underline{\underline{A}} \rightarrow x + \delta x$$

$$b + \delta \underline{\underline{b}} \rightarrow x + \delta x$$

Norms