# 线性方程组的直接解法

- ◆Gauss消元法
- ◆LU三角分解
- ◆范数/误差
- ◆部分主元Gauss消元法
- ◆LU分解与Gauss消元法的联系

# 理论基础

 求解n×n线性方程组  $\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \vdots \\ a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n \end{cases}$ 定义:  $x = [x_1, x_2, ..., x_n]^T$ ,  $b = [b_1, b_2, ..., b_n]^T$  $A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$ 

方程组可表示为Ax = b

# 理论基础

• 如何求解Ax = b?

1. 如果
$$|A| \neq 0$$
,  $x = A^{-1}b$ 

2. 如果 $|A| \neq 0$ , 由Cramer法则知:

$$x_k = \frac{|A_k|}{|A|}$$

• 有没有更简单实用的求解方法?

#### 理论基础

• 容易求解的情况

$$A = I = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \ddots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

$$A = L = \begin{bmatrix} l_{11} & 0 & \dots & 0 \\ l_{21} & l_{22} & \ddots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & \dots & l_{nn} \end{bmatrix} \qquad A = U = \begin{bmatrix} u_{11} & u_{12} & \dots & u_{1n} \\ 0 & u_{22} & \ddots & u_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

$$A = L^* = \begin{bmatrix} l_{11}^* & l_{12}^* & \dots & l_{1n}^* \\ l_{21}^* & l_{22}^* & \ddots & 0 \\ \vdots & \ddots & \vdots & \vdots \\ l_{n1}^* & 0 & \dots & 0 \end{bmatrix}$$

$$A = I = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \ddots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix} \qquad A = D = \begin{bmatrix} d_1 & 0 & \dots & 0 \\ 0 & d_2 & \ddots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & d_n \end{bmatrix}$$

$$A = U = \begin{bmatrix} u_{11} & u_{12} & \dots & u_{1n} \\ 0 & u_{22} & \ddots & u_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

$$A = L^* = \begin{bmatrix} l_{11}^* & l_{12}^* & \dots & l_{1n}^* \\ l_{21}^* & l_{22}^* & \ddots & 0 \\ \vdots & \ddots & \vdots & \vdots \\ l_{n1}^* & 0 & \dots & 0 \end{bmatrix} \qquad A = U^* = \begin{bmatrix} 0 & 0 & \dots & u_{1n}^* \\ 0 & 0 & \ddots & u_{2n}^* \\ \vdots & \ddots & \vdots & \vdots \\ u_{n1}^* & u_{n2}^* & \dots & u_{nn}^* \end{bmatrix}$$

• 核心思想:将方程组化简为三角形方程组



高斯1777-1855



九章算术成书于约公元前150年

- 理论支撑: 对一个方程组做有限次行初等运算得到的方程与原方程同解
  - ▶交换方程组中的两个方程
  - ▶用一个非零数乘一个方程
  - ▶一个方程加上某个其它方程的倍数

• 具体操作

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$a_{ij} \leftarrow a_{ij} - \left(\frac{a_{i1}}{a_{11}}\right) a_{1j}$$
  $b_i \leftarrow b_i - \left(\frac{a_{i1}}{a_{11}}\right) b_1, a_{11} \neq 0$ 

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \mathbf{0} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ \mathbf{0} & a_{n2} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

• 经过n-1轮操作后

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + & \dots + a_{1n}x_n = b_1 \\ a_{22}x_2 + a_{23}x_3 + & \dots + a_{2n}x_n = b_2 \\ \vdots & \vdots & \vdots \\ a_{n-1,n-1}x_{n-1} + a_{n-1,n}x_n = b_{n-1} \\ a_{nn}x_n = b_n \end{cases}$$

• 回代求解

$$x_n = b_n/a_{nn}$$

$$x_{n-1} = (b_{n-1} - a_{n-1,n}x_n)/a_{n-1,n-1}$$

$$x_i = \left(b_i - \sum_{i=i+1}^n a_{ij}x_i\right)/a_{ii}, i = n-1, n-2, ..., 1$$

• 例: 求解线性方程组

$$\begin{cases} x + 2y - z = 3 \\ 2x + y - 2z = 3 \\ -3x + y + z = -6 \end{cases}$$

• 伪代码(pseudocode)

**procedure**  $Naive\_Gauss(n, (a_{ij}), (b_i), (x_i))$ **integer** i, j, k, n; **real** sum, xmult**real array**  $(a_{ij})_{1:n\times 1:n}$ ,  $(b_i)_{1:n}$ ,  $(x_i)_{1:n}$ • for k = 1 to n - 1 do for i = k + 1 to n do  $xmult \leftarrow a_{ik}/a_{kk}$  $a_{ik} \leftarrow xmult$ for j = k + 1 to n do 消元  $a_{ij} \leftarrow a_{ij} - (xmult)a_{kj}$ end for  $b_i \leftarrow b_i - (xmult)b_k$ end for end for  $x_n \leftarrow b_n/a_{nn}$ for i = n - 1 to 1 step -1 do  $sum \leftarrow b_i$ for j = i + 1 to n do  $sum \leftarrow sum - a_{ij}x_j$  end for end procedure Naive\_Gauss

$$\tilde{a}_{ij} = a_{ij} - \left(\frac{a_{ik}}{a_{kk}}\right) a_{kj}$$

$$\tilde{b}_i = b_i - \left(\frac{a_{ik}}{a_{kk}}\right) b_k$$

$$x_n = b_n / a_{nn}$$

$$x_i = \left(b_i - \sum_{j=i+1}^n a_{ij} x_j\right) / a_{ii}$$

• 第一轮消元计算量

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$a_{ij} \leftarrow a_{ij} - \left(\frac{a_{i1}}{a_{11}}\right) a_{1j}$$
  $b_i \leftarrow b_i - \left(\frac{a_{i1}}{a_{11}}\right) b_1, a_{11} \neq 0$ 

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \mathbf{0} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ \mathbf{0} & a_{n2} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

消去一个元的计算量: 1次除法,n次乘法,n次减法(加法)

• 消元总体计算量 $(N_e)$ 

$$\begin{bmatrix} 0 \\ 2n+1 & 0 \\ 2n+1 & 2(n-1)+1 & \ddots \\ \vdots & \vdots & \ddots & 0 \\ 2n+1 & 2(n-1)+1 & \dots & 2(2)+1 & 0 \end{bmatrix}$$

$$N_e = \sum_{j=1}^{n} (2j+1)(j-1) = \frac{2}{3}n^3 + \frac{1}{2}n^2 - \frac{7}{6}n = O(n^3)$$

• 回代计算量 $(N_b)$ 

$$x_n = b_n/a_{nn}$$

$$x_i = \left(b_i - \sum_{j=i+1}^n a_{ij} x_j\right) / a_{ii}, i = n-1, n-2, ..., 1$$

$$N_b = \sum_{i=1}^{n} [2(n-i) + 1] = n^2 = O(n^2)$$

•例:估算在求解一个规模为500×500个未知量的系统时,计算机系统消元过程和回代过程所花的时间之比。

$$\frac{t_b}{t_e} \approx \frac{N_b}{N_e} \approx \frac{500^2}{\frac{2}{3} \times 500^3} = 0.003$$

- 什么是矩阵A的LU分解?
- · LU分解有什么用?
- 怎么分解?

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 & \dots & 0 \\ l_{21} & l_{22} & 0 & \dots & 0 \\ l_{31} & l_{32} & l_{33} & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & l_{n3} & \dots & l_{nn} \end{bmatrix} \cdot \begin{bmatrix} u_{11} & u_{12} & u_{13} & \dots & u_{1n} \\ 0 & u_{22} & u_{23} & \dots & u_{2n} \\ 0 & 0 & u_{33} & \dots & u_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

$$a_{ij} = \sum_{s=1}^{s=n} l_{is} u_{sj} = \sum_{s=1}^{s=\min(i,j)} l_{is} u_{sj}$$
  $n^2$ 个方程, $n^2 + n$ 个未知数

• Doolittle分解

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ l_{21} & 1 & 0 & \dots & 0 \\ l_{31} & l_{32} & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & l_{n3} & \dots & 1 \end{bmatrix} \cdot \begin{bmatrix} u_{11} & u_{12} & u_{13} & \dots & u_{1n} \\ 0 & u_{22} & u_{23} & \dots & u_{2n} \\ 0 & 0 & u_{33} & \dots & u_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

• Crout分解

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 & \dots & 0 \\ l_{21} & l_{22} & 0 & \dots & 0 \\ l_{31} & l_{32} & l_{33} & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & l_{n3} & \dots & l_{nn} \end{bmatrix} \cdot \begin{bmatrix} 1 & u_{12} & u_{13} & \dots & u_{1n} \\ 0 & 1 & u_{23} & \dots & u_{2n} \\ 0 & 0 & 1 & \dots & u_{3n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 & \dots & 0 \\ l_{21} & l_{22} & 0 & \dots & 0 \\ l_{31} & l_{32} & l_{33} & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & l_{n3} & \dots & l_{nn} \end{bmatrix} \cdot \begin{bmatrix} u_{11} & u_{12} & u_{13} & \dots & u_{1n} \\ 0 & u_{22} & u_{23} & \dots & u_{2n} \\ 0 & 0 & u_{33} & \dots & u_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

$$a_{11} = \sum_{s=1}^{s=\min(1,1)} l_{1s} u_{s1} = l_{11} u_{11} \to l_{11} \setminus u_{11}$$

$$a_{1j} = \sum_{s=1}^{s=\min(1,j)} l_{1s} u_{sj} = l_{11} u_{1j} \to u_{1j} \ (2 \le j \le n)$$

$$a_{i1} = \sum_{s=1}^{s=\min(i,1)} l_{is} u_{s1} = l_{i1} u_{11} \to l_{i1} \ (2 \le i \le n)$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 & \dots & 0 \\ l_{21} & l_{22} & 0 & \dots & 0 \\ l_{31} & l_{32} & l_{33} & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & l_{n3} & \dots & l_{nn} \end{bmatrix} \cdot \begin{bmatrix} u_{11} & u_{12} & u_{13} & \dots & u_{1n} \\ 0 & u_{22} & u_{23} & \dots & u_{2n} \\ 0 & 0 & u_{33} & \dots & u_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

$$a_{22} = \sum_{s=1}^{s=\min(2,2)} l_{2s} u_{s2} = l_{21} u_{12} + l_{22} u_{22} \to l_{22} \setminus u_{22}$$

$$a_{2j} = \sum_{s=1}^{s=\min(2,j)} l_{2s} u_{sj} = l_{21} u_{1j} + l_{22} u_{2j} \to u_{2j} \quad (3 \le j \le n)$$

$$a_{i2} = \sum_{s=1}^{s=\min(i,2)} l_{is} u_{s2} = l_{i1} u_{12} + l_{i2} u_{22} \to l_{i2} \quad (3 \le i \le n)$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 & \dots & 0 \\ l_{i1} & \ddots & 0 & \dots & 0 \\ l_{k-1,1} & \dots & l_{k-1,k-1} & \dots & 0 \\ l_{k-1,1} & \dots & l_{nk-1} & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ l_{n1} & \dots & l_{nk-1} & \dots & l_{nn} \end{bmatrix} \cdot \begin{bmatrix} u_{11} & \dots & u_{1,k-1} & \dots & u_{1n} \\ 0 & \ddots & \vdots & \dots & \vdots \\ 0 & 0 & u_{k-1,k-1} & \dots & u_{k-1,n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

$$a_{k,k} = \sum_{s=1}^{s=\min(k,k)} l_{ks} u_{sk} = \sum_{s=1}^{k-1} l_{ks} u_{sk} + l_{kk} u_{kk} \to l_{kk} \setminus u_{kk}$$

$$a_{kj} = \sum_{s=1}^{s=\min(k,j)} l_{ks} u_{sj} = \sum_{s=1}^{k-1} l_{ks} u_{sj} + l_{kk} u_{k,j} \to u_{kj} \ (k < j \le n)$$

$$a_{ik} = \sum_{s=1}^{s=\min(i,k)} l_{is} u_{sk} = \sum_{s=1}^{k-1} l_{is} u_{sk} + l_{ik} u_{kk} \to l_{ik} \ (k < i \le n)$$

• A = LU分解过程

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 & \dots & 0 \\ l_{21} & l_{22} & 0 & \dots & 0 \\ l_{31} & l_{32} & l_{33} & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ l_{n1} & l_{n2} & l_{n3} & \dots & l_{nn} \end{bmatrix} \cdot \begin{bmatrix} u_{11} & u_{12} & u_{13} & \dots & u_{1n} \\ 0 & u_{22} & u_{23} & \dots & u_{2n} \\ 0 & 0 & u_{33} & \dots & u_{3n} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & u_{nn} \end{bmatrix}$$

求解顺序: 
$$u_{11} \to l_{i1}, u_{1j} \to u_{22} \to l_{i2}, u_{2j} \to \cdots \to u_{nn}$$

显式求解:

$$u_{kj} = \left(a_{kj} - \sum_{s=1}^{k-1} l_{ks} u_{sj}\right) / l_{kk}$$

$$l_{ik} = \left(a_{ik} - \sum_{s=1}^{k-1} l_{is} u_{sk}\right) / u_{kk}$$

• 例: 3×3矩阵的Doolittle分解

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & -2 \\ -3 & 1 & 1 \end{bmatrix} = LU = \begin{bmatrix} 1 & 0 & 0 \\ l_{21} & 1 & 0 \\ l_{31} & l_{32} & 1 \end{bmatrix} \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix}$$

$$=\begin{bmatrix} u_{11} & u_{12} & u_{13} \\ l_{21}u_{11} & l_{21}u_{12} + u_{22} & l_{21}u_{13} + l_{22}u_{23} \\ l_{31}u_{11} & l_{31}u_{12} + l_{32}u_{22} & l_{31}u_{13} + l_{32}u_{23} + u_{33} \end{bmatrix}$$

• 求解

$$\begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & -2 \\ -3 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \\ -6 \end{bmatrix}$$

• Doolittle分解的伪代码

integer 
$$i, k, n$$
; real array  $(a_{ij})_{1:n \times 1:n}, (\ell_{ij})_{1:n \times 1:n}, (u_{ij})_{1:n \times 1:n}$  for  $k = 1$  to  $n$  do  $\ell_{kk} \leftarrow 1$  把1赋值给 $L$ 的对角元 for  $j = k$  to  $n$  do  $u_{kj} \leftarrow a_{kj} - \sum_{s=1}^{k-1} \ell_{ks} u_{sj}$  更新 $U$ 的第 $k$ 行 end do for  $i = k+1$  to  $n$  do  $\ell_{ik} \leftarrow \left(a_{ik} - \sum_{s=1}^{k-1} \ell_{is} u_{sk}\right) / u_{kk}$  更新 $L$ 的第 $k$ 列 end do end do

• Doolittle分解计算量?

```
integer i, k, n; real array (a_{ij})_{1:n \times 1:n}, (\ell_{ij})_{1:n \times 1:n}, (u_{ij})_{1:n \times 1:n}
for k = 1 to n do
      \ell_{kk} \leftarrow 1
      for j = k to n do
           u_{kj} \leftarrow a_{kj} - \sum_{s=1}^{\infty} \ell_{ks} u_{sj} k-1次乘法,k-1次减法
      end do
      for i = k + 1 to n do
           \ell_{ik} \leftarrow \left(a_{ik} - \sum_{s=1}^{k-1} \ell_{is} u_{sk}\right) / u_{kk} k - 1 次乘法, k - 1 次减法, 1次除法
      end do
end do
```

总计算量= 
$$\sum_{k=1}^{n} [2(k-1)(n-k+1) + (2k-1)(n-k)] = \frac{2}{3}n^3 - \frac{1}{2}n^2 - \frac{1}{6}n$$

- · LU分解的存在性?
- LU分解定理: 若 $n \times n$ 矩阵A的n个前主子式非奇异,则A有LU分解。

A的第k个前主子式定义为:

$$A_k = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1k} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2k} \\ a_{31} & a_{32} & a_{33} & \dots & a_{3k} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{k1} & a_{k2} & a_{k3} & \dots & a_{kk} \end{bmatrix}$$

• LU分解相比Gauss消元有什么优势?

- 直接法有误差吗?
- 如何衡量误差?
- 向量 $x \in \mathbb{R}^n$ 的范数 $||x|| \in \mathbb{R}$ 满足
- 1. 正定性:  $||x|| \ge 0$ ,仅当x = 0时等号成立
- 2. 齐次性:  $\|\alpha x\| = |\alpha| \cdot \|x\|$
- 3. 三角不等式:  $||x + y|| \le ||x|| + ||y||$
- 例:

$$||x||_1 = |x_1| + \dots + |x_n|$$

$$||x||_2 = \sqrt{x_1^2 + \dots + x_n^2}$$

$$||x||_{\infty} = \max_{1 \le i \le n} |x_i|$$

- 矩阵A的范数||A|| ∈ ℝ满足
- 1. 正定性:  $||A|| \ge 0$ ,仅当A = 0时等号成立
- 2. 齐次性:  $\|\alpha A\| = |\alpha| \cdot \|A\|$
- 3. 三角不等式:  $||A + B|| \le ||A|| + ||B||$
- 算子(从属矩阵)范数:  $||A|| = max \frac{||Ax||}{||x||} (\forall x \in \mathbb{R}^n, x \neq 0)$

性质1.||I|| = 1

性质2. $||A|| \cdot ||x|| \ge ||Ax||, \forall x \in \mathbb{R}^n$ 

性质 $3. \|A\| \cdot \|B\| \ge \|AB\|$ 

• 算子范数的例子

1. 
$$||A||_1 = \max_{1 \le j \le n} (|a_{1j}| + \dots + |a_{nj}|)$$
,最大绝对列和

2. 
$$||A||_2 = \sqrt{\rho(A^T A)}$$
,  $A^T A$ 的谱半径(特征值的最大模)的平方根

$$3. \|A\|_{\infty} = \max_{1 \le i \le n} (|a_{i1}| + \dots + |a_{in}|),$$
 最大绝对行和

- 求解Ax = b的误差定义:
- 1. 近似解 $x_a$ 的前向误差 $||x_a x||_{\infty}$
- 2. 后向误差 $\|b Ax_a\|_{\infty} = \|r\|_{\infty}$
- 3. 误差放大因子=  $\frac{\|x_a x\|_{\infty}/\|x\|_{\infty}}{\|r\|_{\infty}/\|b\|_{\infty}}$
- 4. 也可以采用其它范数来定义误差

• 定义: n阶可逆方阵A的条件数cond(A)为求解Ax = b时,对于所有 $b \neq 0$ 可能出现的最大误差放大因子,即:

$$cond(A) = \max_{b \in \mathbb{R}^n, b \neq 0} \frac{\|x_a - x\|_{\infty} / \|x\|_{\infty}}{\|r\|_{\infty} / \|b\|_{\infty}}$$

• 定理:可逆方阵A的条件数 $cond(A) = ||A||_{\infty} ||A^{-1}||_{\infty}$ ,即:

$$\frac{\|x_a - x\|_{\infty}}{\|x\|_{\infty}} \le \|A\|_{\infty} \|A^{-1}\|_{\infty} \frac{\|r\|_{\infty}}{\|b\|_{\infty}}$$

• 定理: 求解方程Ax = b时, 前向误差和后向误差满足:

$$\frac{1}{cond(A)} \frac{\|r\|_{\infty}}{\|b\|_{\infty}} \le \frac{\|x_a - x\|_{\infty}}{\|x\|_{\infty}} \le cond(A) \frac{\|r\|_{\infty}}{\|b\|_{\infty}}$$

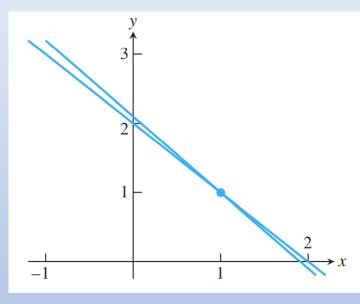
• 大条件数的矩阵是病态的。

• 例: 
$$\begin{cases} x_1 + x_2 = 2 \\ 1.0001x_1 + x_2 = 2.0001 \end{cases}$$

近似解 $x_a = [-1,3.0001]^T$ 的前向误差与后向误差? 系数矩阵条件数?

$$A = \begin{bmatrix} 1 & 1 \\ 1.0001 & 1 \end{bmatrix}, A^{-1} = \begin{bmatrix} -10000 & 10000 \\ 10001 & -10000 \end{bmatrix}$$

 $cond(A) = ||A||_{\infty} ||A^{-1}||_{\infty} \approx 40000$ 



问题的几何解释

```
>> A = [1 1;1.0001 1]; b=[2;2.0001];

>> xa = A\b

xa = 11位

1.000000000000222

0.99999999999778 Matlab求解
```

• Hilbert矩阵 $H\left(h_{ij} = \frac{1}{i+j-1}\right)$ 是有名的病态矩阵

$$H_{5} = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{3} & \frac{1}{4} & \frac{1}{5} \\ \frac{1}{2} & \frac{1}{3} & \frac{1}{4} & \frac{1}{5} & \frac{1}{6} \\ \frac{1}{3} & \frac{1}{4} & \frac{1}{5} & \frac{1}{6} & \frac{1}{7} \\ \frac{1}{4} & \frac{1}{5} & \frac{1}{6} & \frac{1}{7} & \frac{1}{8} \\ \frac{1}{5} & \frac{1}{6} & \frac{1}{7} & \frac{1}{8} & \frac{1}{9} \end{bmatrix}$$

#### Matlab实验

```
>> n=10; H=hilb(n);
>> cond(H,inf)
                 条件数
ans =
    3.535371683074594e+013
>> b=H*ones(n,1);
>> xa=H\b
xa =
   0.9999999875463
   1.00000010746631
   0.99999771299818
   1.00002077769598
   0.99990094548472
   1.00027218303745
  0.99955359665722
   1.00043125589482
   0.99977366058043
   1.00004976229297
```

• 例: 
$$\begin{cases} 10^{-20}x_1 + x_2 = 1\\ x_1 + 2x_2 = 4 \end{cases}$$
, 精确解为 $x_1 = \frac{2}{1 - 2 \times 10^{-20}}$ ,  $x_2 = \frac{1 - 4 \times 10^{-20}}{1 - 2 \times 10^{-20}}$ 

顺序高斯消元:  $\hat{x}_1 = 0, \hat{x}_2 = 1$ 

系数矩阵条件数?

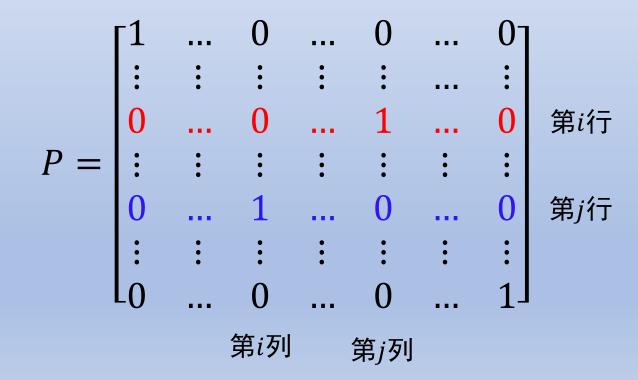
$$A = \begin{bmatrix} 10^{-20} & 1 \\ 1 & 2 \end{bmatrix}, A^{-1} = \frac{1}{2 \times 10^{-20} - 1} \begin{bmatrix} 2 & -1 \\ -1 & 10^{-20} \end{bmatrix}$$

调换方程顺序消元:  $\hat{x}_1 = 2, \hat{x}_2 = 1$ 

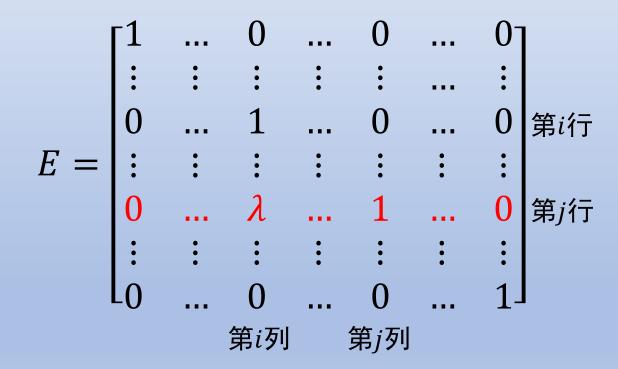
'淹没'引起的误差放大可以通过列主消元避免

• 列主消元: 在消除第i列元素前找到绝对值最大的元素 $a_{ii}$ ( $i \le j \le n$ )

- 高斯消元基本的行变换操作与左乘初等矩阵对应
- 1. 交换第i行和第j行⇔左乘初等矩阵



2. 非零数λ乘第i行加到第j行⇔左乘初等矩阵



- 顺序高斯消元 $\Leftrightarrow E_m ... E_1 A = U \Leftrightarrow A = E_1^{-1} ... E_m^{-1} U, m \leq \frac{n(n-1)}{2}$
- $E_1^{-1} \dots E_m^{-1} = L$ ?

$$\begin{bmatrix} 1 & \dots & 0 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \dots & \vdots \\ 0 & \dots & 1 & \dots & 0 & \dots & 0 \\ \vdots & \vdots \\ 0 & \dots & \lambda & \dots & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & \dots & 0 & \dots & 0 & \dots & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & \dots & 0 & \dots & 0 & \dots & 0 \\ \vdots & \vdots \\ 0 & \dots & 1 & \dots & 0 & \dots & 0 \\ \vdots & \vdots \\ 0 & \dots & -\lambda & \dots & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & \dots & 0 & \dots & 0 & \dots & 1 \end{bmatrix}$$

$$E_{1}^{-1} \dots E_{m}^{-1} = E_{1}^{-1} \dots E_{m-1}^{-1} \cdot \begin{bmatrix} 1 & \dots & 0 & 0 & 0 \\ 0 & \ddots & \vdots & 0 & 0 \\ \vdots & \ddots & 1 & \vdots & \vdots \\ 0 & \ddots & 0 & 1 & 0 \\ 0 & \dots & 0 & c_{n,n-1} & 1 \end{bmatrix} = E_{1}^{-1} \dots E_{m-2}^{-1} \cdot \begin{bmatrix} 1 & \dots & 0 & 0 & 0 \\ 0 & \ddots & \vdots & 0 & 0 \\ \vdots & \ddots & 1 & \vdots & \vdots \\ 0 & \ddots & 0 & 1 & 0 \\ 0 & \dots & c_{n,n-2} & c_{n,n-1} & 1 \end{bmatrix}$$

$$=E_1^{-1}\dots E_{m-3}^{-1}\cdot \begin{bmatrix} 1 & \dots & 0 & 0 & 0 \\ 0 & \ddots & \vdots & 0 & 0 \\ \vdots & \ddots & 1 & \vdots & \vdots \\ 0 & \ddots & c_{n-1,n-2} & 1 & 0 \\ 0 & \dots & c_{n,n-2} & c_{n,n-1} & 1 \end{bmatrix} =E_1^{-1}\dots E_{n-2}^{-1}\cdot \begin{bmatrix} 1 & \dots & 0 & 0 & 0 \\ 0 & \ddots & \vdots & 0 & 0 \\ \vdots & \ddots & 1 & \vdots & \vdots \\ 0 & \ddots & c_{n-1,n-2} & 1 & 0 \\ c_{n,1} & \dots & c_{n,n-2} & c_{n,n-1} & 1 \end{bmatrix}$$

$$=E_1^{-1}\dots E_{n-3}^{-1}\cdot \begin{bmatrix} 1 & \dots & 0 & 0 & 0 \\ 0 & \ddots & \vdots & 0 & 0 \\ \vdots & \ddots & 1 & \vdots & \vdots \\ c_{n-1,1} & \ddots & c_{n-1,n-2} & 1 & 0 \\ c_{n,1} & \dots & c_{n,n-2} & c_{n,n-1} & 1 \end{bmatrix} = \begin{bmatrix} 1 & \dots & 0 & 0 & 0 \\ c_{21} & \ddots & \vdots & 0 & 0 \\ \vdots & \ddots & 1 & \vdots & \vdots \\ c_{n-1,1} & \ddots & c_{n-1,n-2} & 1 & 0 \\ c_{n,1} & \dots & c_{n,n-2} & c_{n,n-1} & 1 \end{bmatrix}$$

• 例: 推导A的LU分解

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & -2 \\ -3 & 1 & 1 \end{bmatrix}$$

• 伪代码

```
procedure LU_Factorization ((a_{ij}))
integer i, j, k, n; real xmult
real array (a_{ij})_{1:n\times 1:n}
for k = 1 to n - 1 do
    for i = k + 1 to n do
         xmult \leftarrow a_{ik}/a_{kk}
a_{ik} \leftarrow xmult
           for j = k + 1 to n do
               a_{ij} \leftarrow a_{ij} - (xmult)a_{kj}
          end for
    end for
end for
```

• 列主高斯消元
$$\Leftrightarrow E_{n(n-1)/2}P_{n-1}\dots E_{2n-3}\dots E_nP_2E_{n-1}\dots E_1P_1A=U$$
 
$$A=P_1^{-1}E_1^{-1}\dots E_{n-1}^{-1}P_2^{-1}E_n^{-1}\dots E_{2n-3}^{-1}\dots P_{n-1}^{-1}E_{n(n-1)/2}^{-1}U$$

$$\bullet \ P_i^{-1} = P_i$$

$$P_1^{-1}E_1^{-1} \dots E_{n-1}^{-1}P_2^{-1}E_n^{-1} \dots E_{2n-3}^{-1} \dots P_{n-1}^{-1}E_{n(n-1)/2}^{-1}$$

$$= P_1^{-1} E_1^{-1} \dots E_{n-1}^{-1} P_2^{-1} E_n^{-1} \dots E_{2n-3}^{-1} \dots P_{n-1}^{-1} \begin{bmatrix} 1 & \dots & 0 & 0 & 0 \\ 0 & \ddots & \vdots & 0 & 0 \\ \vdots & \ddots & 1 & \vdots & \vdots \\ 0 & \ddots & 0 & 1 & 0 \\ 0 & \dots & 0 & c_{n,n-1} & 1 \end{bmatrix}$$

$$= P_1^{-1} E_1^{-1} \dots E_{n-1}^{-1} P_2^{-1} E_n^{-1} \dots E_{2n-3}^{-1} \dots P_{n-2}^{-1} \begin{bmatrix} 1 & \dots & 0 & 0 & 0 \\ 0 & \ddots & \vdots & 0 & 0 \\ \vdots & \ddots & 1 & \vdots & \vdots \\ 0 & \ddots & c_{n,n-2} & c_{n,n-1} & 1 \\ 0 & \dots & c_{n-1,n-2} & 1 & 0 \end{bmatrix}$$

$$P_{n-1} \dots P_1 A = LU \Longrightarrow PA = LU, P = P_{n-1} \dots P_1$$

• PA = LU回代求解Ax = b

$$PAx = Pb \rightarrow LUx = Pb \rightarrow \begin{cases} Ly = Pb \\ Ux = y \end{cases}$$

• 例: 用PA = LU分解求解

$$\begin{bmatrix} 2 & 1 & 5 \\ 4 & 4 & -4 \\ 1 & 3 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 5 \\ 0 \\ 6 \end{bmatrix}$$

• 定义: 严格行对角占优矩阵是指

$$|a_{ii}| > \sum_{j=1, j \neq i}^{n} |a_{ij}|, \qquad (1 \le i \le n)$$

• 定理:不选主元Gauss消元法保持矩阵的严格行对角占优性质。

• 推论: 行对角占优矩阵非奇异且有LU分解。

• 例:

$$A = \begin{bmatrix} 4 & 1 & 1 & 1 \\ 2 & 4 & 0 & 1 \\ 1 & 2 & 5 & 1 \\ 3 & 1 & 1 & 6 \end{bmatrix}$$

# 带状方程组的求解

• 三对角方程组

$$\begin{bmatrix} d_1 & c_1 \\ a_1 & d_2 & c_2 \\ & a_2 & d_3 & c_3 \\ & & \ddots & \ddots & \ddots \\ & & & a_{i-1} & d_i & c_i \\ & & & \ddots & \ddots & \ddots \\ & & & & a_{n-2} & d_{n-1} & c_{n-1} \\ & & & & a_{n-1} & d_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_i \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ \vdots \\ b_i \\ \vdots \\ b_{n-1} \\ b_n \end{bmatrix}$$

# 带状方程组的求解

• 高斯消元法(追赶法)

$$\begin{bmatrix} d_1 & c_1 \\ 0 & d_2 & c_2 \\ 0 & d_3 & c_3 \\ & \ddots & \ddots & \ddots \\ & & 0 & d_i & c_i \\ & & \ddots & \ddots & \ddots \\ & & 0 & d_{n-1} & c_{n-1} \\ 0 & & d_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_i \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ \vdots \\ b_i \\ \vdots \\ b_{n-1} \\ b_n \end{bmatrix}$$

# 带状方程组的求解

• 追赶法的伪代码 **procedure**  $Tri(n, (a_i), (d_i), (c_i), (b_i), (x_i))$ integer i, n; real xmult **real array**  $(a_i)_{1:n}$ ,  $(d_i)_{1:n}$ ,  $(c_i)_{1:n}$ ,  $(b_i)_{1:n}$ ,  $(x_i)_{1:n}$ for i = 2 to n do  $xmult \leftarrow a_{i-1}/d_{i-1}$   $d_i \leftarrow d_i - (xmult)c_{i-1}$   $b_i \leftarrow b_i - (xmult)b_{i-1}$ end for 回代  $\begin{cases} x_n \leftarrow b_n/d_n \\ \text{for } i = n-1 \text{ to } 1 \text{ step } -1 \text{ do} \\ x_i \leftarrow (b_i - c_i x_{i+1})/d_i \end{cases}$  end for 计算量*0*(*n*) end procedure Tri

# 思考与练习

- 给定方程组 $\begin{bmatrix} 1 & 1 \\ 1+\delta & 1 \end{bmatrix}$  $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2+\delta \end{bmatrix}$ ,  $\delta > 0$ . 1、计算系数矩阵的条件数。2、计算近似解 $x_a = [-1,3+\delta]$ 的误差放大因子。
- 用Gauss消元法、列主消元法、LU分解法、PA = LU分解法求解下列方程组

$$\begin{bmatrix} 3 & 1 & 2 \\ 6 & 3 & 4 \\ 3 & 2 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 11 \\ 24 \\ 22 \end{bmatrix}$$

• 编写通用的Gauss消元法和LU分解法程序:1、求解以上方程组对程序进行验证;2、用程序求解 Hx = b, H设置为12阶和20阶Hilbert矩阵,  $b = H \cdot [1, ..., 1]^T$ 。