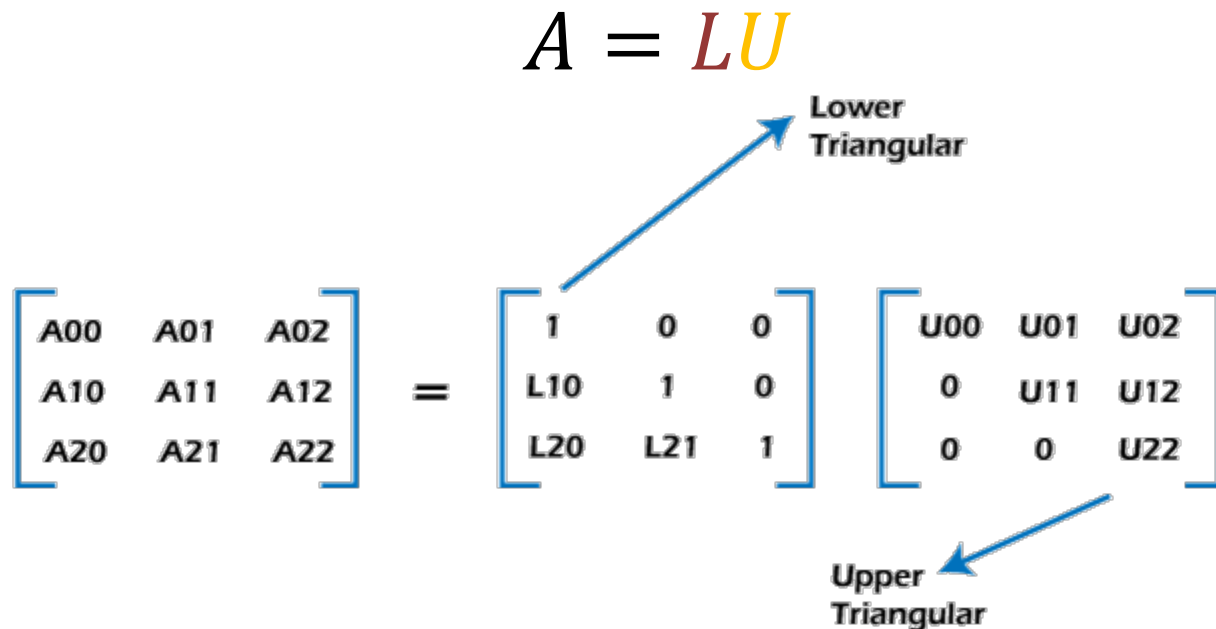


Today's Agenda

- Review on Gaussian elimination
- Testing the pseudo-code
- Pivoting (motivation and basic idea)
- Error analysis

LU factorization

Gaussian elimination transforms a matrix into the product of a unit lower triangular matrix and an upper triangular matrix, i.e.,

$$A = LU$$


$$\begin{bmatrix} A_{00} & A_{01} & A_{02} \\ A_{10} & A_{11} & A_{12} \\ A_{20} & A_{21} & A_{22} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ L_{10} & 1 & 0 \\ L_{20} & L_{21} & 1 \end{bmatrix} \begin{bmatrix} U_{00} & U_{01} & U_{02} \\ 0 & U_{11} & U_{12} \\ 0 & 0 & U_{22} \end{bmatrix}$$

Lower
Triangular

Upper
Triangular

Forward elimination

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & & \vdots \\ a_{i1} & a_{i2} & a_{i3} & \cdots & a_{in} \\ \vdots & \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_i \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ \vdots \\ b_i \\ \vdots \\ b_n \end{bmatrix}$$

$$k + 1 \leq i \leq n$$

$$\begin{cases} a_{ij} \leftarrow a_{ij} - \left(\frac{a_{ik}}{a_{kk}} \right) a_{kj} & (k \leq j \leq n) \\ b_i \leftarrow b_i - \left(\frac{a_{ik}}{a_{kk}} \right) b_k \end{cases}$$

Back substitution

$$\left\{ \begin{array}{rcl} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \cdots & + a_{1n}x_n = & b_1 \\ & a_{22}x_2 + a_{23}x_3 + \cdots & + a_{2n}x_n = b_2 \\ & a_{33}x_3 + \cdots & + a_{3n}x_n = b_3 \\ & \ddots & \vdots \\ & a_{n-1,n-1}x_{n-1} + a_{n-1,n}x_n = & b_{n-1} \\ & & a_{nn}x_n = b_n \end{array} \right.$$

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

Pseudo-code

```
%% forward elimination
```

```
for k = 1:n-1
    for i = k+1:n
        Mtp = A(i,k)/A(k,k);
        for j = k:n
            A(i,j) = A(i,j) - Mtp * A(k,j);
        end
        b(i) = b(i) - Mtp * b(k);
    end
end
```

```
%% backward substitution
```

```
b(n) = b(n)/A(n,n);
for i = n-1 : -1 : 1
    tmp = b(i);
    for j = i+1:n
        tmp = tmp - A(i,j)*b(j);
    end
    b(i) = tmp/A(i,i);
end
```

Testing the pseudocode

- Polynomial interpolation via **Vandermonde**
- Suppose the underlying polynomial

$$p(t) = 1 + t + t^2 + \dots + t^{n-1} = \sum_{j=1}^n t^{j-1}$$

- Evaluate

$$p(1+i) = \sum_{j=1}^n (1+i)^{j-1} = (1+i)^n - 1/(1+i) - 1 = [(1+i)^n - 1]/i$$

- The roundoff error is propagated and magnified throughout the back substitution phase.

- Recall $a_{ij} \leftarrow a_{ij} - \left(\frac{a_{ik}}{a_{kk}}\right) a_{kj}$
- We must expect all quantities to be infected with **roundoff errors**.
- The roundoff error in a_{kj} is multiplied by $\left(\frac{a_{ik}}{a_{kk}}\right)$.
- The small **pivot elements** would lead to large multipliers and to worse roundoff errors.

Naïve Gaussian can fail

- Gaussian elimination would fail if $a_{11} = 0$.

$$\left\{ \begin{array}{l} 0x_1 + x_2 = 1 \\ x_1 + x_2 = 2 \end{array} \right. \quad \left\{ \begin{array}{l} \varepsilon x_1 + x_2 = 1 \\ x_1 + x_2 = 2 \end{array} \right.$$

- How about this for a small number $\varepsilon \neq 0$?

$$\left\{ \begin{array}{l} \varepsilon x_1 + x_2 = 1 \\ (1 - \varepsilon^{-1})x_2 = 2 - \varepsilon^{-1} \end{array} \right.$$

On 8-digit decimal computer

- Consider $\epsilon = 10^{-9} \Rightarrow \epsilon^{-1} = 10^9$.
- To compute $2 - \epsilon^{-1}$, the computer must interpret the numbers as

$$\epsilon^{-1} = 10^9 = 0.10000\,000 \times 10^{10} = 0.10000\,00000\,00000\,0 \times 10^{10}$$

$$2 = 0.20000\,000 \times 10^1 = 0.00000\,00002\,00000\,0 \times 10^{10}$$

- Thus $2 - \epsilon^{-1}$ is rounded to ϵ^{-1} .

Remedy

$$\left\{ \begin{array}{l} 0x_1 + x_2 = 1 \\ x_1 + x_2 = 2 \end{array} \right. \quad \left\{ \begin{array}{l} \varepsilon x_1 + x_2 = 1 \\ x_1 + x_2 = 2 \end{array} \right.$$

Switch the two rows

$$\left\{ \begin{array}{l} x_1 + x_2 = 2 \\ 0x_1 + x_2 = 1 \end{array} \right. \quad \left\{ \begin{array}{l} x_1 + x_2 = 2 \\ \varepsilon x_1 + x_2 = 1 \end{array} \right.$$

- Given

$$\begin{cases} x_1 + x_2 = 2 \\ \varepsilon x_1 + x_2 = 1 \end{cases}$$

- After elimination

$$\begin{cases} x_1 + x_2 = 2 \\ (1 - \varepsilon)x_2 = 1 - 2\varepsilon \end{cases}$$

- Solution

$$x_2 = 1 - 2\varepsilon / 1 - \varepsilon \approx 1$$

$$x_1 = 2 - x_2 \approx 1$$

- Consider

$$\begin{cases} x_1 + \varepsilon^{-1}x_2 = \varepsilon^{-1} \\ x_1 + x_2 = 2 \end{cases}$$

- After elimination

$$\begin{cases} x_1 + \varepsilon^{-1}x_2 = \varepsilon^{-1} \\ (1 - \varepsilon^{-1})x_2 = 2 - \varepsilon^{-1} \end{cases}$$

- Solution

$$x_2 = (2 - \varepsilon^{-1}) / (1 - \varepsilon^{-1}) \approx 1$$

$$x_1 = \varepsilon^{-1} - \varepsilon^{-1}x_2 \approx 0$$

Error analysis

For a linear system $Ax = b$ having the **true solution** x and a **computed solution** \tilde{x} , we define

- Error vector: $e = \tilde{x} - x$
- Residual vector: $r = A\tilde{x} - b$

For two solutions, how do we evaluate which one is better?

- Look at the residual vector: smaller the better!

Condition number