

Lecture 11

Gaussian Elimination, The LU Factorization

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Introduction to Numerical Methods

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The LU Factorization

- Transform $A \in \mathbb{C}^{m \times m}$ into upper triangular U by subtracting multiples of rows
- Each L_i introduces zeros below diagonal of column i :

$$\underbrace{L_{m-1} \cdots L_2 L_1}_{L^{-1}} A = U \quad \implies \quad A = LU \quad \text{where } L = L_1^{-1} L_2^{-1} \cdots L_{m-1}^{-1}$$

$$\begin{array}{c}
 \left[\begin{array}{cccc}
 \times & \times & \times & \times \\
 \times & \times & \times & \times \\
 \times & \times & \times & \times \\
 \times & \times & \times & \times
 \end{array} \right] \\
 A
 \end{array}
 \xrightarrow{L_1}
 \begin{array}{c}
 \left[\begin{array}{cccc}
 \times & \times & \times & \times \\
 \mathbf{0} & \mathbf{\times} & \mathbf{\times} & \mathbf{\times} \\
 \mathbf{0} & \mathbf{\times} & \mathbf{\times} & \mathbf{\times} \\
 \mathbf{0} & \mathbf{\times} & \mathbf{\times} & \mathbf{\times}
 \end{array} \right] \\
 L_1 A
 \end{array}
 \xrightarrow{L_2}
 \begin{array}{c}
 \left[\begin{array}{cccc}
 \times & \times & \times & \times \\
 & \times & \times & \times \\
 & \mathbf{0} & \mathbf{\times} & \mathbf{\times} \\
 & \mathbf{0} & \mathbf{\times} & \mathbf{\times}
 \end{array} \right] \\
 L_2 L_1 A
 \end{array}
 \xrightarrow{L_3}
 \begin{array}{c}
 \left[\begin{array}{cccc}
 \times & \times & \times & \times \\
 & \times & \times & \times \\
 & & \times & \times \\
 & & \mathbf{0} & \mathbf{\times}
 \end{array} \right] \\
 L_3 L_2 L_1 A
 \end{array}$$

- “Triangular triangularization”

Forming L

- The L matrix contains all the multipliers in one matrix (with plus signs)

$$L = L_1^{-1} L_2^{-1} \cdots L_{m-1}^{-1} = \begin{bmatrix} 1 & & & & \\ \ell_{21} & 1 & & & \\ \ell_{31} & \ell_{32} & 1 & & \\ \vdots & \vdots & \ddots & \ddots & \\ \ell_{m1} & \ell_{m2} & \cdots & \ell_{m,m-1} & 1 \end{bmatrix}$$

- Define $\ell_k = (0, \dots, 0, \ell_{k+1,k}, \dots, \ell_{m,n})$. Then $L_k = I - \ell_k e_k^*$.
 - First, $L_k^{-1} = I + \ell_k e_k^*$, since $e_k^* \ell_k = 0$ and

$$(I - \ell_k e_k^*)(I + \ell_k e_k^*) = I - \ell_k e_k^* \ell_k e_k^* = I$$
 - Also, $L_k^{-1} L_{k+1}^{-1} = I + \ell_k e_k^* + \ell_{k+1} e_{k+1}^*$, since $e_k^* \ell_{k+1} = 0$ and

$$(I + \ell_k e_k^*)(I + \ell_{k+1} e_{k+1}^*) = I + \ell_k e_k^* + \ell_{k+1} e_{k+1}^*$$

Gaussian Elimination without Pivoting

- Factorize $A \in \mathbb{C}^{m \times m}$ into $A = LU$:

Algorithm: Gaussian Elimination (no pivoting)

$$U = A, L = I$$

for $k = 1$ **to** $m - 1$

for $j = k + 1$ **to** m

$$\ell_{jk} = u_{jk} / u_{kk}$$

$$u_{j,k:m} = u_{j,k:m} - \ell_{jk} u_{k,k:m}$$

- The inner loop can be written using matrix operations instead of for-loop
- Operation count $\sim \sum_{k=1}^m 2(m-k)(m-k) \sim 2 \sum_{k=1}^m k^2 \sim 2m^3/3$

Pivoting

- At step k , we used matrix element k, k as pivot and introduced zeros in entry k of remaining rows

$$\begin{bmatrix} \times & \times & \times & \times & \times \\ & \mathcal{X}_{kk} & \mathbf{X} & \mathbf{X} & \mathbf{X} \\ & & \times & \times & \times \\ & & \times & \times & \times \\ & & \times & \times & \times \end{bmatrix} \rightarrow \begin{bmatrix} \times & \times & \times & \times & \times \\ & \mathcal{X}_{kk} & \times & \times & \times \\ & & \mathbf{0} & \mathbf{X} & \mathbf{X} & \mathbf{X} \\ & & \mathbf{0} & \mathbf{X} & \mathbf{X} & \mathbf{X} \\ & & \mathbf{0} & \mathbf{X} & \mathbf{X} & \mathbf{X} \end{bmatrix}$$

- But any other element $i \geq k$ in column k can be used as pivot:

$$\begin{bmatrix} \times & \times & \times & \times & \times \\ & \times & \times & \times & \times \\ & \times & \times & \times & \times \\ & & \mathcal{X}_{ik} & \mathbf{X} & \mathbf{X} & \mathbf{X} \\ & & & \times & \times & \times \end{bmatrix} \rightarrow \begin{bmatrix} \times & \times & \times & \times & \times \\ & \mathbf{0} & \mathbf{X} & \mathbf{X} & \mathbf{X} \\ & \mathbf{0} & \mathbf{X} & \mathbf{X} & \mathbf{X} \\ & & \mathcal{X}_{ik} & \times & \times & \times \\ & & & \mathbf{0} & \mathbf{X} & \mathbf{X} & \mathbf{X} \end{bmatrix}$$

Pivoting

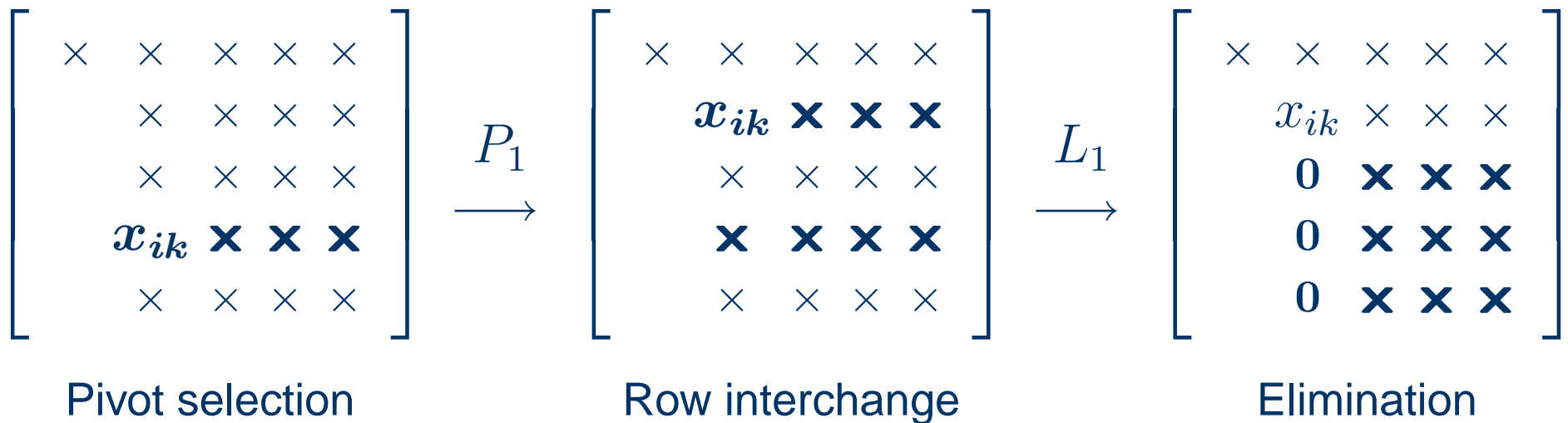
- Also, any other column $j \geq k$ can be used:

$$\begin{bmatrix} \times & \times & \times & \times & \times \\ \times & \times & \times & \times \\ \times & \times & \times & \times \\ \mathbf{\times} & x_{ij} & \mathbf{\times} & \mathbf{\times} \\ \times & \times & \times & \times \end{bmatrix} \rightarrow \begin{bmatrix} \times & \times & \times & \times & \times \\ & \mathbf{\times} & \mathbf{0} & \mathbf{\times} & \mathbf{\times} \\ & \mathbf{\times} & \mathbf{0} & \mathbf{\times} & \mathbf{\times} \\ & \times & x_{ij} & \times & \times \\ & \mathbf{\times} & \mathbf{0} & \mathbf{\times} & \mathbf{\times} \end{bmatrix}$$

- Choosing different pivots means we can avoid zero or very small pivots
- Instead of using pivots at different entries, change rows or columns and use the standard triangular algorithm (*pivoting*)
- A computer code might account for the pivoting indirectly instead of actually moving the data

Partial Pivoting

- Searching among all valid pivots is expensive (*complete pivoting*)
- Consider pivots in column k only and interchange rows (*partial pivoting*)



- In terms of matrices:

$$L_{m-1}P_{m-1} \cdots L_2P_2L_1P_1A = U$$

The $PA = LU$ Factorization

- To combine all L_k and all P_k into matrices, rewrite as

$$L_{m-1}P_{m-1} \cdots L_2P_2L_1P_1A = U$$
$$(L'_{m-1} \cdots L'_2L'_1)(P_{m-1} \cdots P_2P_1)A = U$$

where

$$L'_k = P_{m-1} \cdots P_{k+1}L_kP_{k+1}^{-1} \cdots P_{m-1}^{-1}$$

- This gives the LU factorization of A

$$PA = LU$$

Gaussian Elimination with Partial Pivoting

- Factorize $A \in \mathbb{C}^{m \times m}$ into $PA = LU$:

Algorithm: Gaussian Elimination (partial pivoting)

$$U = A, L = I, P = I$$

for $k = 1$ **to** $m - 1$

 Select $i \geq k$ to maximize $|u_{ik}|$

$u_{k,k:m} \leftrightarrow u_{i,k:m}$ (interchange two rows)

$l_{k,1:k-1} \leftrightarrow l_{i,1:k-1}$

$p_{k,:} \leftrightarrow p_{i,:}$

for $j = k + 1$ **to** m

$$l_{jk} = u_{jk}/u_{kk}$$

$$u_{j,k:m} = u_{j,k:m} - l_{jk}u_{k,k:m}$$

Complete Pivoting

- If pivots are selected from a different column, permutation matrices Q_k for the columns are required:

$$L_{m-1}P_{m-1} \cdots L_2P_2L_1P_1AQ_1Q_2 \cdots Q_{m-1} = U$$

$$(L'_{m-1} \cdots L'_2L'_1)(P_{m-1} \cdots P_2P_1)A(Q_1Q_2 \cdots Q_{m-1}) = U$$

- Set

$$L = (L'_{m-1} \cdots L'_2L'_1)^{-1}$$

$$P = P_{m-1} \cdots P_2P_1$$

$$Q = Q_1Q_2 \cdots Q_{m-1}$$

to obtain

$$PAQ = LU$$