

# SI231b: Matrix Computations

## Lecture 6: Solving Linear Equations (Direct Methods)

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- ▶ Example of LU Factorization
- ▶ LU Factorization with Pivoting
- ▶ Implementation on Computers
- ▶ Computational Complexity of LU Factorization
- ▶ General Procedure of Direct Methods

# An Example of LU Factorization

Given

$$\mathbf{A} = \begin{bmatrix} 2 & 2 & 2 \\ 4 & 7 & 7 \\ 6 & 18 & 22 \end{bmatrix},$$

the LU factorization is given by

$$\mathbf{L} = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 4 & 1 \end{bmatrix}, \quad \mathbf{U} = \begin{bmatrix} 2 & 2 & 2 \\ 0 & 3 & 3 \\ 0 & 0 & 4 \end{bmatrix}$$

Please give  $\mathbf{L}$  and  $\mathbf{U}$  by yourself

# LU Factorization with Pivoting

## Step $k$ of LU factorization

$$\begin{bmatrix} a_{11}^{(0)} & \times & \times & \cdots & \times & \times \\ 0 & a_{22}^{(1)} & \times & \cdots & \times & \times \\ 0 & 0 & \times & \cdots & \times & \times \\ \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ 0 & 0 & 0 & a_{kk}^{(k-1)} & \times & \times \\ 0 & 0 & 0 & \times & \times & \times \\ 0 & 0 & 0 & \times & \times & \times \end{bmatrix} \longrightarrow \begin{bmatrix} a_{11}^{(0)} & \times & \times & \cdots & \times & \times \\ 0 & a_{22}^{(1)} & \times & \cdots & \times & \times \\ 0 & 0 & \times & \cdots & \times & \times \\ \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ 0 & 0 & 0 & a_{kk}^{(k-1)} & \times & \times \\ 0 & 0 & 0 & 0 & \times & \times \\ 0 & 0 & 0 & 0 & \times & \times \end{bmatrix}$$

- ▶  $\mathbf{A}^{(k)} = \mathbf{M}_k \mathbf{A}^{(k-1)}$
- ▶ Require:  $a_{kk}^{(k-1)} \neq 0$ 
  - under **which condition**?
  - if **unsatisfied**, what to do?

# LU Factorization with Pivoting

$$\begin{bmatrix} a_{11}^{(0)} & \times & \times & \cdots & \times & \times \\ 0 & a_{22}^{(1)} & \times & \cdots & \times & \times \\ 0 & 0 & \times & \cdots & \times & \times \\ \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ 0 & 0 & 0 & \textcircled{0} & \times & \times \\ 0 & 0 & 0 & \times & \times & \times \\ 0 & 0 & 0 & \textcircled{S} & \times & \times \end{bmatrix} \xrightarrow{\text{pivoting}} \begin{bmatrix} a_{11}^{(0)} & \times & \times & \cdots & \times & \times \\ 0 & a_{22}^{(1)} & \times & \cdots & \times & \times \\ 0 & 0 & \times & \cdots & \times & \times \\ \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ 0 & 0 & 0 & \textcircled{S} & \times & \times \\ 0 & 0 & 0 & \times & \times & \times \\ 0 & 0 & 0 & \textcircled{0} & \times & \times \end{bmatrix}$$

## ► partial pivoting

- finding  $p = \arg \max_{k \leq i \leq n} |a_{ik}^{(k-1)}|$
- let  $a_{kk}^{(k-1)} = a_{pk}^{(k-1)}$  (row exchange)

## ► complete pivoting

- finding  $[p_r, p_c] = \arg \max_{k \leq i, j \leq n} |a_{ij}^{(k-1)}|$
- let  $a_{kk}^{(k-1)} = a_{p_r p_c}^{(k-1)}$  (row and column exchange)

## Permutation Matrix

A square matrix with exactly one entry of 1 in each row and column and 0 elsewhere.

### Example

$$\mathbf{P} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}, \quad \mathbf{Px} = \begin{bmatrix} x_3 \\ x_2 \\ x_1 \end{bmatrix}, \quad \mathbf{x}^T \mathbf{P} = \begin{bmatrix} x_3 & x_2 & x_1 \end{bmatrix}$$

**PA**: exchange rows of **A**

**AP**: exchange columns of **A**

Properties:

- ▶ **P** is an orthogonal matrix, i.e.,  $\mathbf{P}^T \mathbf{P} = \mathbf{P} \mathbf{P}^T = \mathbf{I}$ .
- ▶  $\mathbf{P}^{-1} = \mathbf{P}^T$

# LU Factorization with Partial Pivoting

Step  $k$  of LU factorization

1. row exchange:  $\tilde{\mathbf{A}}^{(k-1)} = \mathbf{P}_k \mathbf{A}^{(k-1)}$
2. Gaussian elimination:  $\mathbf{A}^{(k)} = \mathbf{M}_k \tilde{\mathbf{A}}^{(k-1)}$

In general, the procedure follows

$$\mathbf{M}_{n-1} \mathbf{P}_{n-1} \mathbf{M}_{n-2} \mathbf{P}_{n-2} \cdots \mathbf{M}_1 \mathbf{P}_1 \mathbf{A} = \mathbf{U}.$$

Denote

$$\tilde{\mathbf{M}}_{n-1} = \mathbf{M}_{n-1},$$

$$\tilde{\mathbf{M}}_{n-2} = \mathbf{P}_{n-1} \mathbf{M}_{n-2} \mathbf{P}_{n-1}^T,$$

$$\vdots = \vdots$$

$$\tilde{\mathbf{M}}_k = \mathbf{P}_{n-1} \mathbf{P}_{n-2} \cdots \mathbf{P}_{k+1} \mathbf{M}_k \mathbf{P}_{k+1}^T \cdots \mathbf{P}_{n-2}^T \mathbf{P}_{n-1}^T$$

**Note:**  $\tilde{\mathbf{M}}_k$  has the same structure with  $\mathbf{M}_k$  (recall the structure of  $\mathbf{M}_k$ )

# LU Factorization with Partial Pivoting

Following the aforementioned procedure,

where  $\mathbf{PA} = \mathbf{LU}$ ,

- ▶  $\mathbf{P} = \mathbf{P}_{n-1}\mathbf{P}_{n-2}\cdots\mathbf{P}_1$  is again a permutation matrix (why?)
- ▶  $\mathbf{L} = \left(\tilde{\mathbf{M}}_{n-1}\tilde{\mathbf{M}}_{n-2}\cdots\tilde{\mathbf{M}}_1\right)^{-1}$  is a lower-triangular matrix with unit diagonals
- ▶ sometimes called **LUP** factorization
- ▶ always exists for any square  $\mathbf{A}$ , no matter  $\mathbf{A}$  is nonsingular or not<sup>1</sup>

## Another Interpretation

1. permute the rows of  $\mathbf{A}$  according to  $\mathbf{P}$
2. compute the LU factorization without pivoting to  $\mathbf{PA}$

**Note:** LU factorization with partial pivoting is not carried out in this way, since  $\mathbf{P}$  is unknown in advance.

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<sup>1</sup><https://arxiv.org/abs/math/0506382>



# A Simple Example

$$\mathbf{A} = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

Step 1, 1st row  $\longleftrightarrow$  3rd row of  $\mathbf{A}$ , then perform Gaussian elimination

$$\tilde{\mathbf{A}}^{(0)} = \mathbf{P}_1 \mathbf{A} = \begin{bmatrix} & & 1 & \\ & 1 & & \\ 1 & & & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

$$\mathbf{A}^{(1)} = \mathbf{M}_1 \tilde{\mathbf{A}}^{(0)} = \begin{bmatrix} 1 & & & \\ -\frac{1}{2} & 1 & & \\ -\frac{1}{4} & & 1 & \\ -\frac{3}{4} & & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \\ & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \end{bmatrix}$$

# A Simple Example

Step 2: 2nd row  $\longleftrightarrow$  4th row of  $\mathbf{A}^{(1)}$ , then repeat Gaussian elimination

$$\tilde{\mathbf{A}}^{(1)} = \mathbf{P}_2 \mathbf{A}^{(1)} = \begin{bmatrix} 1 & & & \\ & & & 1 \\ & & 1 & \\ & 1 & & \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} & \\ \frac{7}{4} & \frac{9}{4} & \frac{17}{4} & \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} & \\ -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} & \end{bmatrix}$$

$$\mathbf{A}^{(2)} = \mathbf{M}_2 \tilde{\mathbf{A}}^{(1)} = \begin{bmatrix} 1 & & & \\ & 1 & & \\ \frac{3}{7} & & 1 & \\ \frac{2}{7} & & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} & \\ -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} & \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{2}{7} & \frac{4}{7} \\ & & -\frac{6}{7} & -\frac{2}{7} \end{bmatrix}$$

Now, it's your turn to give  $\mathbf{P}_3$ ,  $\mathbf{M}_3$  and the final  $\mathbf{P}$ ,  $\mathbf{L}$ , and  $\mathbf{U}$

## A Simple Example

$$\underbrace{\begin{bmatrix} & & 1 & \\ & & & 1 \\ & 1 & & \\ 1 & & & \end{bmatrix}}_{\mathbf{P}} \underbrace{\begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix}}_{\mathbf{A}} = \underbrace{\begin{bmatrix} 1 & & & \\ \frac{3}{4} & 1 & & \\ \frac{1}{2} & -\frac{2}{7} & 1 & \\ \frac{1}{4} & -\frac{3}{7} & \frac{1}{3} & 1 \end{bmatrix}}_{\mathbf{L}} \underbrace{\begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & & \frac{2}{3} \end{bmatrix}}_{\mathbf{U}}$$

**In practice**, the permutation matrix  $\mathbf{P}$

- ▶ is not represented explicitly as a matrix or the product of permutation matrices
- ▶ an equivalent effect can be achieved via a permutation vector

**Note:**  $|\ell_{ij}| \leq 1$  for  $i \geq j$

# LU Factorization with Complete Pivoting

## LU with complete pivoting:

In matrix form, at each stage before Gaussian elimination

- ▶ permutation of rows with  $\mathbf{P}_k$  on the left
- ▶ permutation of columns with  $\mathbf{Q}_k$  on the right

$$\mathbf{M}_{n-1}\mathbf{P}_{n-1}\mathbf{M}_{n-2}\mathbf{P}_{n-2}\cdots\mathbf{M}_1\mathbf{P}_1\mathbf{A}\mathbf{Q}_1\mathbf{Q}_2\cdots\mathbf{Q}_{n-1} = \mathbf{U}.$$

By

- ▶ using the same definition of  $\mathbf{L}$ ,  $\mathbf{P}$  with LU factorization with partial pivoting,
- ▶ denoting  $\mathbf{Q} = \mathbf{Q}_1\mathbf{Q}_2\cdots\mathbf{Q}_{n-1}$ ,

the LU factorization with complete pivoting can be represented by

$$\mathbf{PAQ} = \mathbf{LU}$$

Too computationally expensive, why?

## LU Factorization without Pivoting:

```
U = A, L = I;  
for k = 1 : n-1  
    for j = k+1 : n  
         $\ell_{jk} = u_{jk} / u_{kk}$   
         $u_{j,k:n} = u_{j,k:n} - \ell_{jk} u_{k,k:n}$   
    end  
end
```

Operations count:

►  $\mathcal{O}\left(\frac{2}{3}n^3\right)$  flops

Please give your own explanation

## LU Factorization with Partial Pivoting:

```
U = A, L = I, P = I;  
for k = 1 : n-1  
    select  $i \geq k$  to maximize  $|u_{ik}|$   
     $u_{k,k:m} \leftrightarrow u_{i,k:m}$  (exchange of rows)  
     $\ell_{k,1:k-1} \leftrightarrow \ell_{i,1:k-1}$   
     $p_{k,:} \leftrightarrow p_{i,:}$   
    for j = k+1 : n  
         $\ell_{jk} = u_{jk}/u_{kk}$   
         $u_{j,k:n} = u_{j,k:n} - \ell_{jk}u_{k,k:n}$   
    end  
end
```

Operations count:

►  $\mathcal{O}\left(\frac{2}{3}n^3\right)$  flops, flops count of partial pivoting?

## General Procedure of Direct Methods

1. compute the LU factorization with partial pivoting,  $\mathbf{PA} = \mathbf{LU}$ ,  $\mathcal{O}(\frac{2}{3}n^3)$  flops
2. solve  $\mathbf{Lz} = \mathbf{Pb}$  using forward substitution,  $\mathcal{O}(n^2)$  flops
3. solve  $\mathbf{Ux} = \mathbf{z}$  using backward substitution,  $\mathcal{O}(n^2)$  flops

## Variant of LU Factorization: LDU Factorization

For LU factorization with partial pivoting  $\mathbf{PA} = \mathbf{LU}$

- ▶ denote  $\mathbf{D} = \text{diag}(u_{11}, u_{22}, \dots, u_{nn})$
- ▶  $\bar{\mathbf{U}} = \mathbf{D}^{-1}\mathbf{U}$ : upper-triangular matrix with unit diagonal entries, i.e.,  $\bar{u}_{ij} = u_{ij}/u_{ii}$  for  $i \leq j$

Then  $\mathbf{PA} = \mathbf{LD}\bar{\mathbf{U}}$  gives an LDU factorization of  $\mathbf{A}$

You are supposed to read

- ▶ Gene H. Golub and Charles F. Van Loan. *Matrix Computations*, Johns Hopkins University Press, 2013.

Chapter 3.1 – 3.4

- ▶ Lloyd N. Trefethen and David Bau III. *Numerical Linear Algebra*, SIAM, 1997.

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