

# I3344

## Numerical Simulation & Modelling

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# Solving a Linear Equation

# Solution of Linear System of Equations

# Linear System of Equations

- A linear system of  $n$  algebraic equations in  $n$  unknowns  $x_1, x_2, \dots, x_n$  is in the form:

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

- $a_{ij}$  ( $i, j = 1, 2, \dots, n$ ) and  $b_k$  ( $k = 1, 2, \dots, n$ ) are known constants
- $a_{ij}$ s are the coefficients
- If every  $b_k$  is zero, the system is **homogeneous**, otherwise it is **non-homogeneous**

# Linear System of Equations (cont'd)

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots a_{2n}x_n = b_2 \\ \dots \\ a_{n1}x_1 + a_{n2}x_2 + \dots a_{nn}x_n = b_n \end{cases} \Leftrightarrow Ax = b$$

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix}_{n \times n}, x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \dots \\ x_n \end{bmatrix}_{n \times 1}, b = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ \dots \\ b_n \end{bmatrix}_{n \times 1}$$

**Coefficient Matrix**

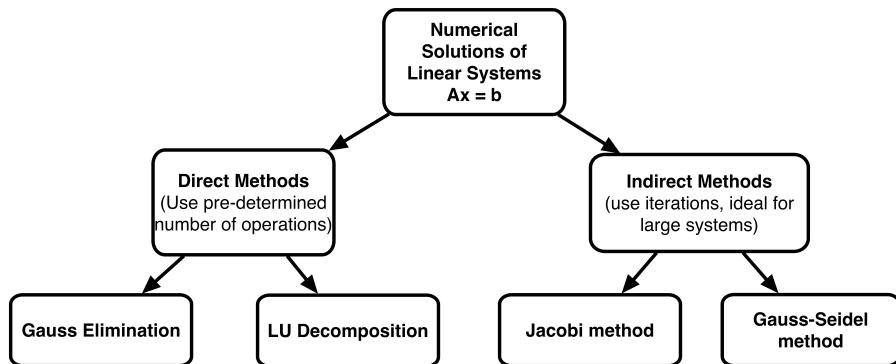
# Linear System of Equations (cont'd)

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots a_{2n}x_n = b_2 \\ \dots \\ a_{n1}x_1 + a_{n2}x_2 + \dots a_{nn}x_n = b_n \end{cases} \Leftrightarrow Ax = b$$

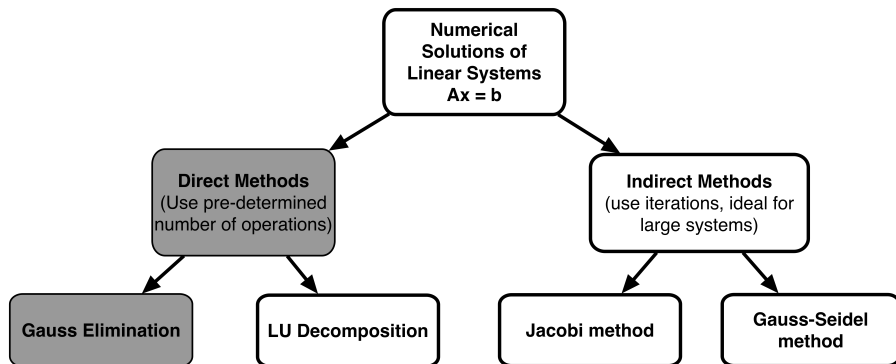
$$A = [A|b] = \left[ \begin{array}{ccccc|c} a_{11} & a_{12} & a_{13} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & \vdots & & \vdots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} & b_n \end{array} \right]_{n \times (n+1)}$$

**Augmented Matrix**

# Numerical Solutions of Linear Systems



# Numerical Solutions of Linear Systems





# Gauss Elimination Method

# Introduction

- **Gauss elimination** is a procedure that **transforms a linear system of equations** ( $Ax = b$ ) into **upper-triangular form**
  - ▶ The solution of the transformed system is found by **Back Substitution**.
- The **modifications must be applied to the augmented matrix  $[A \mid b]$**  and **not matrix A alone**
- The transformation into **upper-triangular form** is achieved by **using elementary row operations**

$$\left[ \begin{array}{ccc|c} a_{11} & a_{12} & a_{13} & b_1 \\ a_{21} & a_{22} & a_{23} & b_2 \\ a_{31} & a_{32} & a_{33} & b_n \end{array} \right]$$

↓ (*Forward Elimination*)

$$\left[ \begin{array}{ccc|c} a_{11} & a_{12} & a_{13} & b_1 \\ & a'_{22} & a'_{23} & b'_2 \\ & & a''_{33} & b''_3 \end{array} \right]$$

↓ (*Back Substitution*)

$$\begin{aligned} x_3 &= b''_3 / a''_{33} \\ x_2 &= (b'_2 - a'_{23}x_3) / a'_{22} \\ x_1 &= (b_1 - a_{12}x_2 - a_{13}x_3) / a_{11} \end{aligned}$$

# Introduction (cont'd)

- There are three types of elementary row operations:
  1. **Multiply a row** of the augmented matrix **by a non-zero constant**
  2. **Interchange any two rows** of the augmented matrix
  3. **Multiply the  $i^{th}$  row** of the augmented matrix **by a constant  $\alpha \neq 0$**  and **add the result to the  $k^{th}$  row**, then **replace the  $k^{th}$  row with the outcome**
    - ▶ The  $i^{th}$  row is called the **pivot row**
- The nature of a linear system is preserved under **elementary row operations**
  - ▶ The new system and the original one are called **row-equivalent**

# Gauss Elimination - How it works?

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

- The *first objective* is to **eliminate**  $x_1$  in all equations below the **first**, thus **the first row is the pivot row**
- The entry that plays the most important role here is  $a_{11}$  (known as the **pivot**)
- If  $a_{11} = 0$ , the first row must be interchanged with another row to **ensure that  $x_1$  has a non-zero coefficient**; This is called **Partial Pivoting**.
- **Partial pivoting** can be done also when the **pivot is very small**

# Gauss Elimination - How it works?

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

- The *next step* is to **focus on the coefficient of  $x_2$  in the second row** of this new system.
  - ▶ If it is nonzero, and not very small, we use it as the **pivot** and **eliminate  $x_2$  in all the lower-level equations**.
  - ▶ The second row is the **pivot row** and remains unchanged.
- This continues until an **upper-triangular system is formed**.
- Finally, **Back Substitution** is used to find the solution.

# Gauss Elimination - Example

$$\left[ \begin{array}{ccc|c} 1 & 3 & 1 & 9 \\ 1 & 1 & -1 & 1 \\ 3 & 11 & 5 & 35 \end{array} \right] \quad (1)$$

 $\Rightarrow$ 

$$\left[ \begin{array}{ccc|c} 1 & 3 & 1 & 9 \\ 0 & -2 & -2 & -8 \\ 0 & 2 & 2 & 8 \end{array} \right] \quad (2)$$

$$\left[ \begin{array}{ccc|c} 1 & 3 & 1 & 9 \\ 0 & -2 & -2 & -8 \\ 0 & 0 & 0 & 0 \end{array} \right] \quad (3)$$

 $\Rightarrow$ 

$$\left[ \begin{array}{ccc|c} 1 & 0 & -2 & -3 \\ 0 & 1 & 1 & 4 \\ 0 & 0 & 0 & 0 \end{array} \right] \quad (4)$$

## Exercise

- Using Gauss elimination, find the solution  $x_1, x_2, x_3, x_4$  of the system whose augmented matrix is

$$\left[ \begin{array}{cccc|c} -1 & 2 & 3 & 1 & 3 \\ 2 & -4 & 1 & 2 & -1 \\ -3 & 8 & 4 & -1 & 6 \\ 1 & 4 & 7 & -2 & -4 \end{array} \right]$$

## Exercise - Solution

$$\begin{array}{c}
 \textcircled{1} \quad \textcircled{-3} \quad \textcircled{2} \\
 \downarrow \quad \downarrow \quad \downarrow \\
 + \quad + \quad +
 \end{array}
 \left[ \begin{array}{cccc|c}
 -1 & 2 & 3 & 1 & 3 \\
 2 & -4 & 1 & 2 & -1 \\
 -3 & 8 & 4 & -1 & 6 \\
 1 & 4 & 7 & -2 & -4
 \end{array} \right] \xrightarrow{\text{Pivot row}} \left[ \begin{array}{cccc|c}
 -1 & 2 & 3 & 1 & 3 \\
 0 & 0 & 7 & 4 & 5 \\
 0 & 2 & -5 & -4 & -3 \\
 0 & 6 & 10 & -1 & -1
 \end{array} \right]$$

$$\begin{array}{c}
 \textcircled{-3} \\
 \downarrow \\
 +
 \end{array}
 \left[ \begin{array}{cccc|c}
 -1 & 2 & 3 & 1 & 3 \\
 0 & 2 & -5 & -4 & -3 \\
 0 & 0 & 7 & 4 & 5 \\
 0 & 6 & 10 & -1 & -1
 \end{array} \right] \xrightarrow{\text{Pivot row}} \left[ \begin{array}{cccc|c}
 -1 & 2 & 3 & 1 & 3 \\
 0 & 2 & -5 & -4 & -3 \\
 0 & 0 & 7 & 4 & 5 \\
 0 & 0 & 25 & 11 & 8
 \end{array} \right] \xrightarrow{\text{Pivot row}} \left[ \begin{array}{cccc|c}
 -1 & 2 & 3 & 1 & 3 \\
 0 & 2 & -5 & -4 & -3 \\
 0 & 0 & 7 & 4 & 5 \\
 0 & 0 & 0 & -\frac{23}{7} & -\frac{69}{7}
 \end{array} \right]$$

$$-\frac{23}{7}x_4 = -\frac{69}{7} \Rightarrow x_4 = 3 \Rightarrow$$

$$x_3 = \frac{1}{7}(5 - 4x_4) = -1$$

$$x_2 = \frac{1}{2}(5x_3 + 4x_4 - 3) = 2$$

$$x_1 = 2x_2 + 3x_3 + x_4 - 3 = 1$$



# Exercises

- Use Gaussian elimination to solve the following systems of linear equations

$$\begin{cases} 2x_1 + x_3 = -8 \\ x_1 - 2x_2 - 3x_3 = 0 \\ -x_1 + x_2 + 2x_3 = 3 \end{cases}$$

$$\begin{cases} x_1 - 2x_2 - 6x_3 = 12 \\ 2x_1 + 4x_2 + 12x_3 = -17 \\ x_1 - 4x_2 - 12x_3 = 22 \end{cases}$$

# Choosing the Pivot Row

## Partial Pivoting with Row Scaling

- When using **Partial Pivoting**, in the first step of the elimination process, it is common to **choose as the pivot row the row in which  $x_1$  has the largest (in absolute value) coefficient**.
  - ▶ The subsequent steps are treated in a similar manner.
- This is mainly to **handle round-off error** while dealing with large matrices.

# Choosing the Pivot Row

## Partial Pivoting with Row Scaling (cont'd)

- Assume  $A$  is  $n \times n$
- In each row  $i$  of  $A$ , find the **entry with the largest absolute value**; call it  $M_i$
- In each row  $i$ , find the **ratio of the absolute value of the coefficient of  $x_1$  to the absolute value of  $M_i$** :  $r_i = |a_{i1}| / |M_i|$
- Among  $r_i$  ( $i = 1, 2, \dots, n$ ) **pick the largest**.
  - ▶ Whichever row is responsible for this maximum value is picked as the **pivot row**.
  - ▶ Eliminate  $x_1$  to obtain a new system.
- In the new system, consider the  $(n - 1) \times (n - 1)$  sub-matrix of the coefficient matrix occupying the lower right corner.
  - ▶ In this matrix, use the same logic as above to choose the pivot row to eliminate  $x_2$ , and so on.

# Choosing the Pivot Row

## Partial Pivoting with Row Scaling (cont'd)

### Example

$$[A | \mathbf{b}] = \left[ \begin{array}{ccc|c} -4 & -3 & 5 & 0 \\ 6 & 7 & -3 & 2 \\ 2 & -1 & 1 & 6 \end{array} \right]$$

$$r_1 = \frac{|-4|}{|5|} = \frac{4}{5}, \quad r_2 = \frac{|6|}{|7|} = \frac{6}{7}, \quad r_3 = \frac{|2|}{|2|} = 1$$

$$\left[ \begin{array}{ccc|c} 2 & -1 & 1 & 6 \\ 6 & 7 & -3 & 2 \\ -4 & -3 & 5 & 0 \end{array} \right] \xrightarrow{\text{Eliminate } x_1} \left[ \begin{array}{ccc|c} 2 & -1 & 1 & 6 \\ 0 & 10 & -6 & -16 \\ 0 & -5 & 7 & 12 \end{array} \right]$$

B

# Choosing the Pivot Row

## Partial Pivoting with Row Scaling (cont'd)

Example (cont'd)

$$\left[ \begin{array}{ccc|c} 2 & -1 & 1 & 6 \\ 6 & 7 & -3 & 2 \\ -4 & -3 & 5 & 0 \end{array} \right] \xrightarrow{\text{Eliminate } x_1} \left[ \begin{array}{ccc|c} 2 & -1 & 1 & 6 \\ 0 & 10 & -6 & -16 \\ 0 & -5 & 7 & 12 \end{array} \right]$$

B

$$\frac{|10|}{|10|} = 1, \quad \frac{|-5|}{|7|} = \frac{5}{7}$$

$$\left[ \begin{array}{ccc|c} 2 & -1 & 1 & 6 \\ 0 & 10 & -6 & -16 \\ 0 & 0 & 4 & 4 \end{array} \right] \Rightarrow x_3 = 1, x_2 = -1, x_1 = 2$$

# Exercises

Solve by Gauss elimination with partial pivoting. Show all steps of the computation

$$\begin{cases} 2x_1 - 6x_2 - x_3 = -38 \\ -3x_1 - x_2 + 7x_3 = -34 \\ -8x_1 + x_2 - 2x_3 = -20 \end{cases}$$

# Exercises

- Solve by Gauss elimination with partial pivoting. Show all steps of the computation

$$\begin{cases} x_1 + x_2 - x_3 = -3 \\ 6x_1 + 2x_2 + 2x_3 = 2 \\ -3x_1 + 4x_2 + x_3 = 1 \end{cases}$$

# Gauss-Jordan Method

- The **Gauss-Jordan method** is a variation of Gauss elimination.
- The major difference is that when an unknown is eliminated in the Gauss-Jordan method, **it is eliminated from all other equations** rather than just the subsequent ones.



# Gauss-Jordan - How it works?

- All rows are normalized by **dividing them by their pivot elements**  
 $\Rightarrow$
- The **elimination step results in an identity matrix** rather than a triangular matrix  $\Rightarrow$
- it is **not necessary to employ back substitution** to obtain the solution.

## Gauss-Jordan - How it works? (cont'd)

$$\begin{cases} 3x_1 - 0.1x_2 - 0.2x_3 = 7.85 \\ 0.1x_1 + 7x_2 - 0.3x_3 = -19.3 \\ 0.3x_1 - 0.2x_2 + 10x_3 = 71.4 \end{cases} \Rightarrow \begin{bmatrix} 3 & -0.1 & -0.2 & 7.85 \\ 0.1 & 7 & -0.3 & -19.3 \\ 0.3 & -0.2 & 10 & 71.4 \end{bmatrix}$$

**Normalize the first row by dividing it by the pivot element: 3**

$$\Rightarrow \begin{bmatrix} 1 & -0.03333 & -0.06667 & 2.61667 \\ 0.1 & 7 & -0.3 & -19.3 \\ 0.3 & -0.2 & 10 & 71.4 \end{bmatrix}$$

**Eliminate  $x_1$  term from the second and third rows**

$$\Rightarrow \begin{bmatrix} 1 & -0.03333 & -0.06667 & 2.61667 \\ 0 & 7.03333 & -0.29333 & -19.5617 \\ 0 & -0.19000 & 10.02000 & 70.6150 \end{bmatrix}$$

# Gauss-Jordan - How it works? (cont'd)

**Normalize second row by dividing it by the pivot element: 7.00333**

$$\Rightarrow \begin{bmatrix} 1 & -0.03333 & -0.06667 & 2.61667 \\ 0 & 1 & -0.0418848 & -2.79320 \\ 0 & -0.19000 & 10.02000 & 70.6150 \end{bmatrix}$$

**Reduction of the  $x_2$  terms from the first and third equations**

$$\Rightarrow \begin{bmatrix} 1 & 0 & -0.0680629 & 2.52356 \\ 0 & 1 & -0.0418848 & -2.79320 \\ 0 & 0 & 10.01200 & 70.0843 \end{bmatrix}$$

**Normalize the third row by dividing it by 10.0120**

$$\Rightarrow \begin{bmatrix} 1 & 0 & -0.0680629 & 2.52356 \\ 0 & 1 & -0.0418848 & -2.79320 \\ 0 & 0 & 1 & 7.000 \end{bmatrix}$$

## Gauss-Jordan - How it works? (cont'd)

**The  $x_3$  terms can be reduced from the first and second equations**

$$\Rightarrow \begin{bmatrix} 1 & 0 & 0 & 3.0000 \\ 0 & 1 & 0 & -2.5000 \\ 0 & 0 & 1 & 7.0000 \end{bmatrix}$$

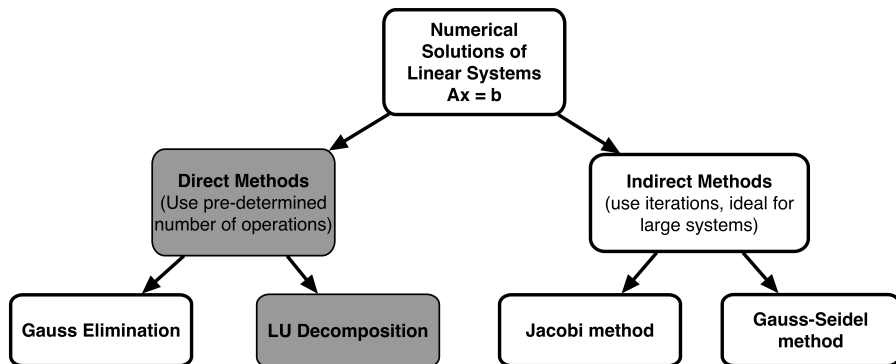
# Exercises

- Use Gauss-Jordan elimination to solve each of the following linear systems of Equations

$$\begin{cases} 2x_1 + x_2 - x_3 = 1 \\ 5x_1 + 2x_2 + 2x_3 = -4 \\ 3x_1 + x_2 + x_3 = 5 \end{cases}$$

$$\begin{cases} x_1 + x_2 - x_3 = -3 \\ 6x_1 + 2x_2 + 2x_3 = 2 \\ -3x_1 + 4x_2 + x_3 = 1 \end{cases}$$

# Numerical Solutions of Linear Systems



# LU Decomposition

# Introduction

- Main advantage of LU decomposition
  - ▶ Involves **only operations on the matrix of coefficients  $[A]$** .
  - ▶ Well suited for those situations where **many right-hand-side vectors  $B$  must be evaluated** for a single value of  $[A]$ .
    - Gauss Elimination becomes **inefficient** in such cases
- One motive for introducing LU decomposition is that it provides an **efficient means to compute the matrix inverse**.



# Overview of LU Decomposition

## ■ Simple Presentation (without Pivoting)

$$[A]\{x\} - \{B\} = 0 \xrightarrow{\text{e.g., Gauss Elimination}} \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} d_1 \\ d_2 \\ d_3 \end{Bmatrix}$$

$[U] \qquad \qquad \{X\} \qquad \qquad \{D\}$

$$\Rightarrow \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} - \begin{Bmatrix} d_1 \\ d_2 \\ d_3 \end{Bmatrix} = 0 \Rightarrow [U]\{X\} - \{D\} = 0$$

**Assume that there is a lower diagonal matrix with 1's on the diagonal**

$$[L] = \begin{bmatrix} 1 & 0 & 0 \\ l_{21} & 1 & 0 \\ l_{31} & l_{32} & 1 \end{bmatrix} \Bigg/ [L]\{[U]\{X\} - \{D\}\} = [A]\{X\} - \{B\} \Rightarrow \begin{cases} [L][U] = [A] \\ [L]\{D\} = \{B\} \end{cases}$$

# LU Decomposition - Example

- How to factorize/ Decompose A?

$$A = \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} = LU$$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ L_{21} & 1 & 0 \\ L_{31} & L_{32} & 1 \end{bmatrix} \text{ and } U = \begin{bmatrix} U_{11} & U_{12} & U_{13} \\ 0 & U_{22} & U_{23} \\ 0 & 0 & U_{33} \end{bmatrix}$$

$$\begin{bmatrix} U_{11} & U_{12} & U_{13} \\ L_{21}U_{11} & L_{21}U_{12} + U_{22} & L_{21}U_{13} + U_{23} \\ L_{31}U_{11} & L_{31}U_{12} + L_{32}U_{22} & L_{31}U_{13} + L_{32}U_{23} + U_{33} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix}$$

# LU Decomposition - Example (cont'd)

$$\begin{bmatrix} U_{11} & U_{12} & U_{13} \\ L_{21}U_{11} & L_{21}U_{12} + U_{22} & L_{21}U_{13} + U_{23} \\ L_{31}U_{11} & L_{31}U_{12} + L_{32}U_{22} & L_{31}U_{13} + L_{32}U_{23} + U_{33} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix}$$

$$\boxed{U_{11} = 1}, \boxed{U_{12} = 2}, \boxed{U_{13} = 4}$$

$$L_{21}U_{11} = 3 \Rightarrow L_{21} \times 1 = 3 \Rightarrow \boxed{L_{21} = 3}$$

$$L_{21}U_{12} + U_{22} = 8 \Rightarrow 3 \times 2 + U_{22} = 8 \Rightarrow \boxed{U_{22} = 2}$$

$$L_{21}U_{13} + U_{23} = 14 \Rightarrow 3 \times 4 + U_{23} = 14 \Rightarrow \boxed{U_{23} = 2}$$

$$L_{31}U_{11} = 2 \Rightarrow L_{31} \times 1 = 2 \Rightarrow \boxed{L_{31} = 2}$$

$$L_{31}U_{12} + L_{32}U_{22} = 6 \Rightarrow 2 \times 2 + L_{32} \times 2 = 6 \Rightarrow \boxed{L_{32} = 1}$$

$$L_{31}U_{13} + L_{32}U_{23} + U_{33} = 13 \Rightarrow (2 \times 4) + (1 \times 2) + U_{33} = 13 \Rightarrow \boxed{U_{33} = 3}$$

## LU Decomposition - Example (cont'd)

$$A = \begin{bmatrix} 1 & 2 & 34 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 2 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 4 \\ 0 & 2 & 2 \\ 0 & 0 & 3 \end{bmatrix}$$

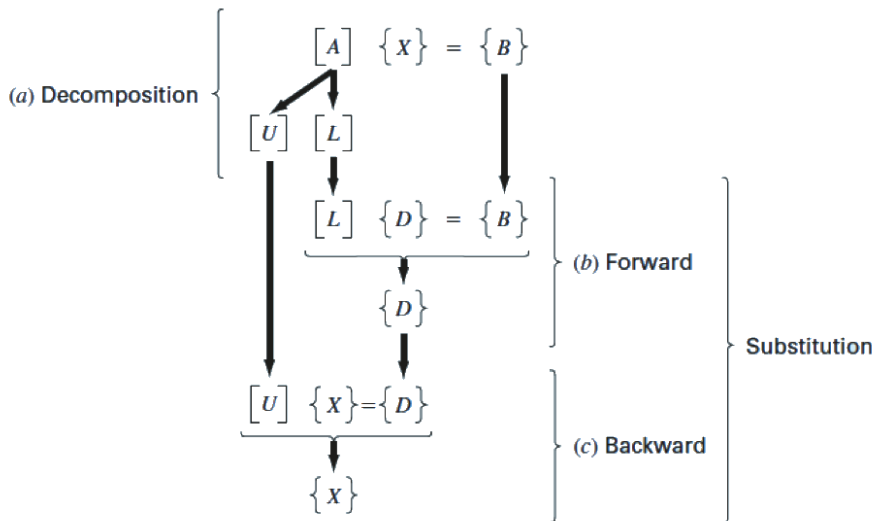
# Exercises

- Find an LU decomposition of  $\begin{bmatrix} 3 & 1 \\ -6 & -4 \end{bmatrix}$
- Find an LU decomposition of  $\begin{bmatrix} 3 & 1 & 6 \\ -6 & 0 & -16 \\ 0 & 8 & -17 \end{bmatrix}$

# LU Decomposition to solve Linear Systems

- Given  $A$ , find  $L$  and  $U$  so that  $A = LU \Rightarrow LUX = B$ .
- Let  $Y = UX \Rightarrow LY = B$ 
  - ▶ Solve this triangular system for  $Y$ .
- Solve the triangular system  $UX = Y$  for  $X$ .
- The benefit of this approach is that we only ever need to solve triangular systems.
- The cost is that we have to solve two of them.

# LU Decomposition to solve Linear Systems (cont'd)



# LU Decomposition to solve Linear Systems - Example

Find the solution of  $x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$  of  $\begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix}$



# LU Decomposition to solve Linear Systems - Example

Find the solution of  $x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$  of  $\begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix}$

**First step:** calculate the LU decomposition of A

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

# LU Decomposition to solve Linear Systems - Example

Find the solution of  $x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$  of  $\begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix}$

**First step:** calculate the LU decomposition of A

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

**Second step:** solve  $LY = B$

$$LY = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 2 & 1 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix} \Rightarrow Y = \begin{bmatrix} 3 \\ 4 \\ -6 \end{bmatrix}$$

# LU Decomposition to solve Linear Systems - Example

Find the solution of  $x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$  of  $\begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix}$

**First step:** calculate the LU decomposition of A

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

**Second step:** solve  $LY = B$

$$LY = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 2 & 1 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix} \Rightarrow Y = \begin{bmatrix} 3 \\ 4 \\ -6 \end{bmatrix}$$

**Third step:** solve  $UX = Y$

$$UX = \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \\ -6 \end{bmatrix} = Y \Rightarrow X = \begin{bmatrix} 3 \\ 4 \\ -2 \end{bmatrix}$$

## Exercise

- Use the LU decomposition (previously calculated) to solve

$$\begin{bmatrix} 3 & 1 & 6 \\ -6 & 0 & 16 \\ 0 & 8 & -17 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 4 \\ 17 \end{bmatrix}$$

# Conditions for LU decomposition

- Sometimes it is impossible to write a matrix in the form "lower triangular"  $\times$  "upper triangular"
- An invertible matrix  $A$  has an LU decomposition provided that all its leading sub-matrices have non-zero determinants.
- The  $k^{th}$  leading sub-matrix of  $A$  is denoted  $A_k$  and is the  $k \times k$  matrix found by looking only at the top  $k$  rows and leftmost  $k$  columns.

# Conditions for LU decomposition - Example

$$A = \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix}$$

$$A_1 = 1, A_2 = \begin{bmatrix} 1 & 2 \\ 3 & 8 \end{bmatrix}, A_3 = \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix}$$

$$|A_1| = 1$$

$$|A_2| = (1 \times 8) - (2 \times 3) = 2,$$

$$|A_3| = \begin{vmatrix} 8 & 14 \\ 6 & 13 \end{vmatrix} - 2 \begin{vmatrix} 3 & 14 \\ 2 & 13 \end{vmatrix} + 4 \begin{vmatrix} 3 & 8 \\ 2 & 6 \end{vmatrix} = 20 - (2 \times 11) + (4 \times 2) = 6$$

A has an LU decomposition because none of these determinants is zero

# Exercises

- Check if the following matrix has an LU decomposition

$$\begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 1 & 3 & 4 \end{bmatrix}$$

- Which, if any, of these matrices have an LU decomposition?

(i)  $A = \begin{bmatrix} 3 & 2 \\ 0 & 1 \end{bmatrix}$ , (ii)  $\begin{bmatrix} 0 & 1 \\ 3 & 2 \end{bmatrix}$ , (iii)  $A = \begin{bmatrix} 1 & -3 & 7 \\ -2 & 6 & 1 \\ 0 & 3 & -2 \end{bmatrix}$

## Conditions for LU decomposition (cont'd)

- Can we get around this problem?
- It is always possible to re-order the rows of an invertible matrix so that all of the sub-matrices have non-zero determinants
- Example: Reorder the rows of  $A$  so that the reordered matrix has an LU decomposition.

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



## Exercise

- Reorder the rows of  $A$  so that the reordered matrix has an LU decomposition.

$$A = \begin{bmatrix} 1 & -3 & 7 \\ -2 & 6 & 1 \\ 0 & 3 & -2 \end{bmatrix}$$

# Exercises

- Solve the following Linear Systems using LU decomposition

$$(a) \begin{bmatrix} 2 & 1 \\ -4 & -6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$(b) \begin{bmatrix} 2 & 1 & -4 \\ 2 & 1 & -2 \\ 6 & 3 & -11 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 15 \\ 7 \\ 41 \end{bmatrix}$$

$$(c) \begin{bmatrix} 1 & 3 & 2 \\ 2 & 8 & 5 \\ 1 & 11 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \\ 0 \end{bmatrix}$$

## Exercises

Consider  $A = \begin{bmatrix} 1 & 6 & 2 \\ 2 & 12 & 5 \\ -1 & -3 & -1 \end{bmatrix}$

- (a) Show that  $A$  does not have an LU decomposition
- (b) Re-order the rows of  $A$  and find an LU decomposition of the new matrix
- (c) Hence solve

$$\begin{cases} x_1 + 6x_2 + 2x_3 = 9 \\ 2x_1 + 12x_2 + 5x_3 = -4 \\ -x_1 - 3x_2 - x_3 = 17 \end{cases}$$