

# System of Linear Equations

①

Given,

$$x_1 + x_2 + 3x_4 = 4 \quad : E1$$

$$2x_1 + x_2 - x_3 + x_4 = 1 \quad : E2$$

$$3x_1 - x_2 - x_3 + 2x_4 = -3 \quad : E3$$

$$-x_1 + 2x_2 + 3x_3 - x_4 = 4 \quad : E4$$



$$(E2 - 2E1) \rightarrow (E2)$$

↓ produces

$$(2x_1 + x_2 - x_3 + x_4) - 2(x_1 + x_2 + 3x_4) = 1 - 2 \times 4 = -7$$

↓ System becomes

$$x_1 + x_2 + 3x_4 = 4 \quad : E1$$

$$-x_2 - x_3 - 5x_4 = -7 \quad : E2$$

$$-4x_2 - x_3 - 5x_4 = -15 \quad : E3$$

$$3x_2 + 3x_3 + 2x_4 = 8 \quad : E4$$

↓ by performing  
(E3 - 4E2) → E3  
(E4 + 3E2) → E4

$$x_1 + x_2 + 3x_4 = 4$$

$$-x_2 - x_3 - 5x_4 = -7$$

$$3x_3 + 13x_4 = 13$$

$$-13x_4 = -13$$

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☐ The system of equations is now in triangular form and can be ~~used~~ solved using the backward-substitution process.

Here,  $E_4$  gives  $x_4 = 1$ .

We use  $x_4 = 1$  to obtain  $x_3$  from  $E_3$

$$x_3 = \frac{1}{3} (13 - 13x_4) = \frac{1}{3} (13 - 13) = 0$$

Now,  $E_2$  gives —

$$\begin{aligned} x_2 &= -(-7 + 5x_4 + x_3) \\ &= -(-7 + 5 + 0) = 2 \end{aligned}$$

and  $E_1$  gives —

$$x_1 = 4 - 3x_4 - x_2 = 4 - 3 - 2 = -1$$

So, Sol<sup>n</sup> becomes

$$\begin{array}{l|l} x_1 = -1 & x_3 = 0 \\ x_2 = 2 & x_4 = 1 \end{array}$$

☐ Matrix can be used to represent linear system:

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 \dots a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 \dots a_{2n}x_n = b_2$$

⋮

$$a_{n1}x_1 + a_{n2}x_2 + a_{n3}x_3 \dots a_{nn}x_n = b_n$$

Finally, we find that

$$\begin{aligned}
\omega_1 &= 1 + \frac{1}{6} (k_1 + 2k_2 + 2k_3 + k_4) \\
&= 1 + \frac{1}{6} (2 + 2(2.3333) + 2(2.4444) \\
&\quad + 2.7222) \\
&= 3.379630
\end{aligned}$$

### System of Equations [ IVP in vector form ]

Let us consider a case

$$\left. \begin{aligned}
\frac{dx_1}{dt} &= a_1 x_1 - b_1 x_1^2 + c_1 x_1 x_2 \\
\frac{dx_2}{dt} &= a_2 x_2 - b_2 x_2^2 + c_2 x_1 x_2
\end{aligned} \right\} \begin{array}{l} \text{Predator-} \\ \text{Prey} \\ \text{Model.} \end{array}$$

Here,  $x_1$  and  $x_2$  are the interacting population and we need to solve this using IVP method.

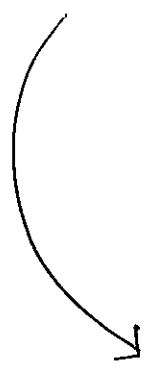
- ✓ Analysis of such systems require more advanced techniques than the scalar case, as they often tend to exhibit a wider variety of system behavior.
- ✓ However, fortunately, numerical analysis requires little more than a notational change from the scalar case.

So, we can represent using

$$\bar{A} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & & & \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \text{ and } \bar{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

We also can combine them as follows:

$$[\bar{A}, \bar{b}] \equiv \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} & | & b_1 \\ \vdots & & & & | & b_2 \\ a_{n1} & a_{n2} & \dots & a_{nn} & | & b_n \end{bmatrix}$$

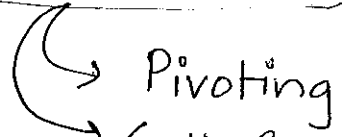


Augmented matrix



We perform elementary row operations to obtain a form as below

$$\begin{bmatrix} a_{11} & a_{12} & \dots & \dots & a_{1n} & | & a_{1, n+1} \\ 0 & \dots & a_{22} & \dots & \dots & a_{2n} & | & a_{2, n+1} \\ \vdots & & & & & & & \vdots \\ 0 & \dots & \dots & \dots & 0 & a_{nn} & | & a_{n, n+1} \end{bmatrix}$$



Called Gaussian elimination

Now, we can solve the linear system to find the unknowns.

The method is called Gaussian elimination with backward substitution

Example

$$\begin{aligned}
 x_1 + x_2 + 3x_4 &= 4 \\
 2x_1 + x_2 - x_3 + x_4 &= 1 \\
 3x_1 - x_2 - x_3 + 2x_4 &= -3 \\
 -x_1 + 2x_2 + 3x_3 - x_4 &= 4
 \end{aligned}$$

Augmented Matrix —

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



$$\left[ \begin{array}{cccc|c}
 1 & 1 & 0 & 3 & 4 \\
 0 & -1 & -1 & -5 & -7 \\
 0 & -4 & -1 & -7 & -15 \\
 0 & 3 & 3 & 2 & 8
 \end{array} \right] \longrightarrow \left[ \begin{array}{cccc|c}
 1 & 1 & 0 & 3 & 4 \\
 0 & -1 & -1 & -5 & -7 \\
 0 & 0 & 3 & 13 & 13 \\
 0 & 0 & 0 & -13 & -13
 \end{array} \right]$$



Now, we use back-substitution and the whole process is known as —

∪ Gaussian Elimination with backward substitution ∪



# TRANSFORMING REAL-LIFE PROBLEMS TO $AX=B$

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Let's consider a Boundary Value Problem of the heat diffusion equation.

$$\frac{\partial u}{\partial t} = e^{\gamma} \frac{\partial^2 u}{\partial x^2} ; a < x < b$$

$$u(x, 0) = u_0(x), \quad u(a, t) = A, \quad u(b, t) = B - e^{-t}$$

steady-state problem is

$$e^{\gamma} \frac{\partial^2 u}{\partial x^2} = 0, \quad a < x < b$$

↗ length of the bar

$$u(a) = A \quad \text{or} \quad u(a) = u_a$$

$$u(b) = B \quad \text{or} \quad u(b) = u_b$$

we generalize by considering some non-zero  $f(x)$

So, we consider the below generalize system

$$\frac{\partial^2 u}{\partial x^2} = f(x) \quad a < x < b$$

$$u \text{ at } "a" = u_a$$

$$u \text{ at } "b" = u_b$$

We can discretize the spatial domain ( $x$ ) into a total of "n" sub-intervals.

$$h = \frac{b-a}{n}$$

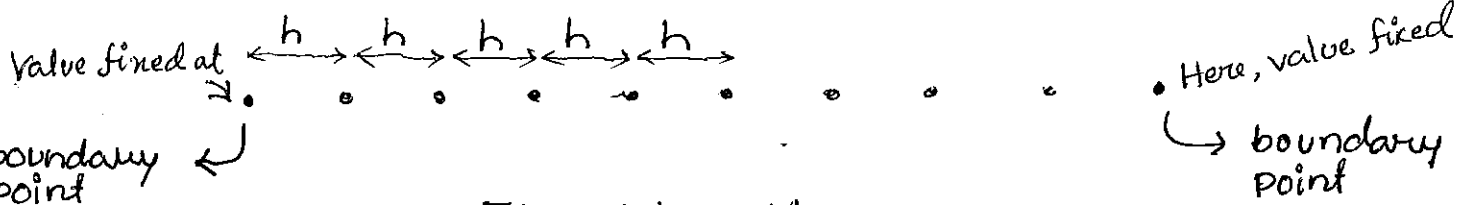


Fig. Discretized spatial Domain. (Mesh)

As the values <sup>are</sup> fixed at the two boundaries, values don't change. So, we treat the intermediate points as below:

$$\frac{\partial^2 u}{\partial x^2} = f(x)$$

$$\Rightarrow \frac{u_{i-1} - 2u_i + u_{i+1}}{h^2} = f(x_i) \quad \text{Using central Difference Method}$$

$$\Rightarrow \frac{u_{i-1} - 2u_i + u_{i+1}}{h^2} = f(x_i)$$

Now, at mesh node  $i = 1$

$$\frac{u_0 - 2u_1 + u_2}{h^2} = f(x_1)$$

$$\Rightarrow \frac{u_0 - 2u_1 + u_2}{h^2} = f(x_1) \quad \left| \begin{array}{l} u \text{ at "0" mesh} \\ \text{is equal to} \\ u_0 \end{array} \right.$$

@ mesh node  $i = 2$

$$\frac{u_1 - 2u_2 + u_3}{h^2} = f(x_2)$$

@ mesh node  $i = 3$

$$\frac{u_2 - 2u_3 + u_4}{h^2} = f(x_3)$$

@ mesh node  $i = n-2$

$$\frac{u_{n-3} - 2u_{n-2} + u_{n-1}}{h^2} = f(x_{n-2})$$

@ mesh node  $i = n-1$

$$\frac{(u_{n-2} - 2u_{n-1} + u_n)}{h^2} = f(x_{n-1})$$





$$\boxed{\square} \quad AX = b$$

↳ Linear system representation

- ✱ Many real world problems are directly linear system
- ✱ There are other instances where the linear system occur as a part of the numerical analysis of other problems.

↳ this could be from the numerical analysis of many non-linear equations.



So, it is very important that a system of equations represented as

$$AX = B$$

is solved efficiently.

$\boxed{\square}$  Information on the matrix "A"

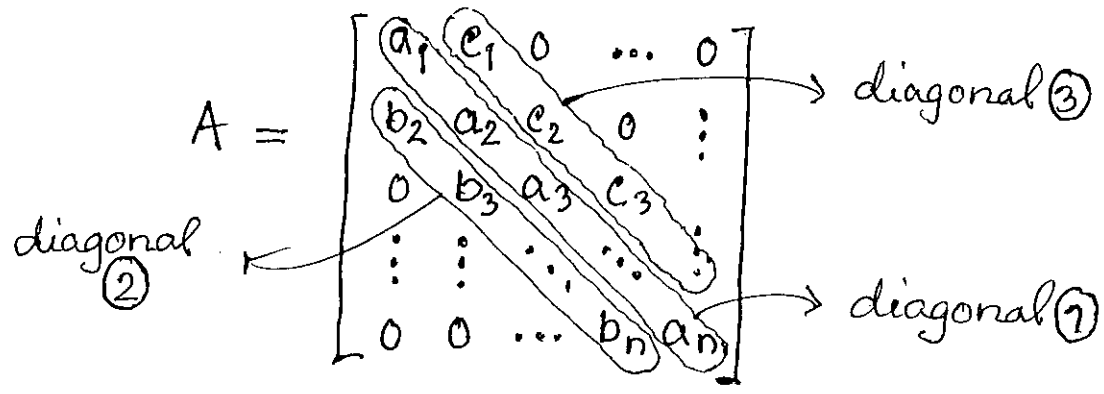
The matrix "A" could be sparse or dense

**Sparse:** In many linear system, the matrix A contains a large number of zero's as its elements.

These matrices are known as sparse matrices

☐ For example :

In many systems, the matrix A becomes tri-diagonal. as below :

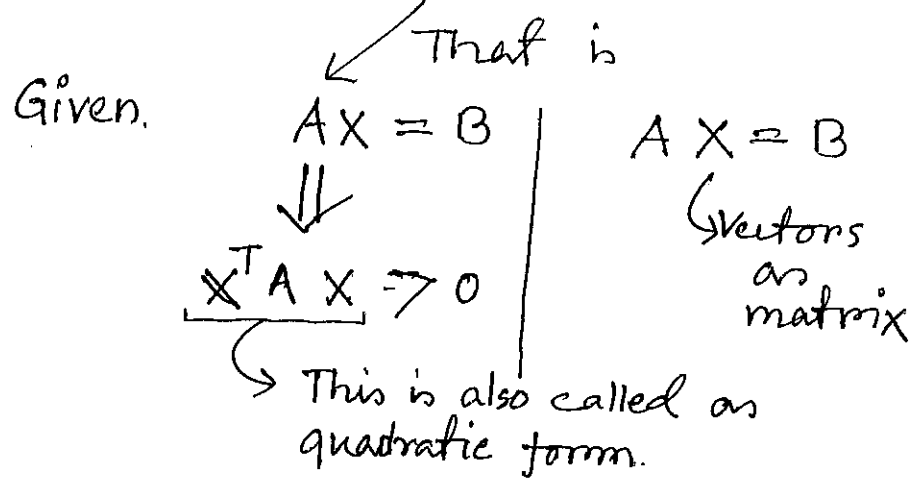


☐ Special properties :

Often the matrix "A" has the special properties. For instance -

✦ It could be tridiagonal, banded  
 ↳ example given above

✦ It could be positive definite



All these properties are generally used to develop more efficient and computationally less expensive methods to solve  $AX = B$ .

# ☐ Solution Approach for $AX = b$

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## Direct Methods:

Given that arithmetic operations are done accurately, using some methods, the exact solutions "x" of  $AX = b$  can be calculated. These methods are known as Direct Methods.

↓ example

One of the most popular direct methods is the Gaussian Elimination.

↓ it reduces "A" to  
Triangular form

↳ Then we do back substitution

☐

Another example of direct method relies on the factorization of the matrix A as the product of two triangular matrices.

Here, factorization transforms  $AX = b$  into two other easily invertible systems.

$$\begin{array}{l|l} AX = b & \text{Factorization} \\ \Rightarrow B \underline{C} X = b & A = BC \\ \Rightarrow By = b & \\ \text{and } CX = y & \end{array}$$

We invert the above systems and obtain X.

## Iterative Methods :

Generally, these methods are applicable for all types of linear systems. However, they are commonly used for large sparse matrix.

↳ often produced by discretizing partial differential equations.

## LU Decomposition

Gaussian elimination with back-substitution shows that systems of equations involving triangular coefficient matrices ( $A$ ) are easy to deal with.

Gaussian elimination:  
transforms  $A$  to  
a triangular matrix

From  $A$  to a upper triangular matrix.

Then, back-substitution

As the triangular matrices are easy to deal with,  
LU decomposition transforms the matrix  $A$  into two different triangular matrices.

$A = LU$ .  
 $\rightarrow$  Upper triangular matrix  
 $\rightarrow$  Lower triangular matrix

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

$$L = \begin{bmatrix} 1 & 0 & 0 \\ L_{21} & 1 & 0 \\ L_{31} & L_{32} & 1 \end{bmatrix}$$

Lower triangular

$$\text{and } U = \begin{bmatrix} U_{11} & & U_{13} \\ 0 & U_{22} & U_{23} \\ 0 & 0 & U_{33} \end{bmatrix}$$

Upper triangular

So, Multiplying  $L$  and  $U$  obtain,

$$LU = \begin{bmatrix} U_{11} & U_{12} & U_{13} \\ L_{21}U_{11} & L_{21}U_{12} + U_{22} & L_{21}U_{13} + U_{23} \\ L_{31}U_{11} & L_{31}U_{12} + L_{32}U_{22} & L_{31}U_{13} + L_{32}U_{23} + U_{33} \end{bmatrix}$$

$\parallel$  equals to

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

Element-wise comparison between  $LU$  &  $A$  provides:

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

$$\begin{array}{l} L_{21}U_{11} = 3 \quad | \quad L_{21} + U_{22} \quad | \quad L_{21} + U_{23} = 14 \quad | \quad L_{31}U_{11} = 2 \\ \Rightarrow L_{21} = 3 \quad | \Rightarrow U_{22} = 2 \quad | \Rightarrow U_{23} = 2, \quad | \Rightarrow L_{31} = 2 \end{array}$$

$$\begin{array}{l} L_{31}U_{12} + L_{32}U_{22} = 6 \quad | \quad L_{31}U_{13} + L_{32}U_{23} + U_{33} = 13 \\ \Rightarrow 2 \times 2 + L_{32} \times 2 = 6 \quad | \Rightarrow (2 \times 4) + (1 \times 2) + U_{33} = 13 \\ \Rightarrow L_{32} = 1 \quad | \Rightarrow U_{33} = 3 \end{array}$$

So, we obtain,

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

$\therefore$  Multiply  $LU$  to verify that it equals to  $A$

### Alternative Approach:

Given a matrix "A", we can transform "A" into an Upper triangular matrix by using Gauss elimination.

↳ Here, we perform elementary row operation.

For instance, let's assume  $A = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 3 & 5 \\ 4 & 6 & 8 \end{bmatrix}$

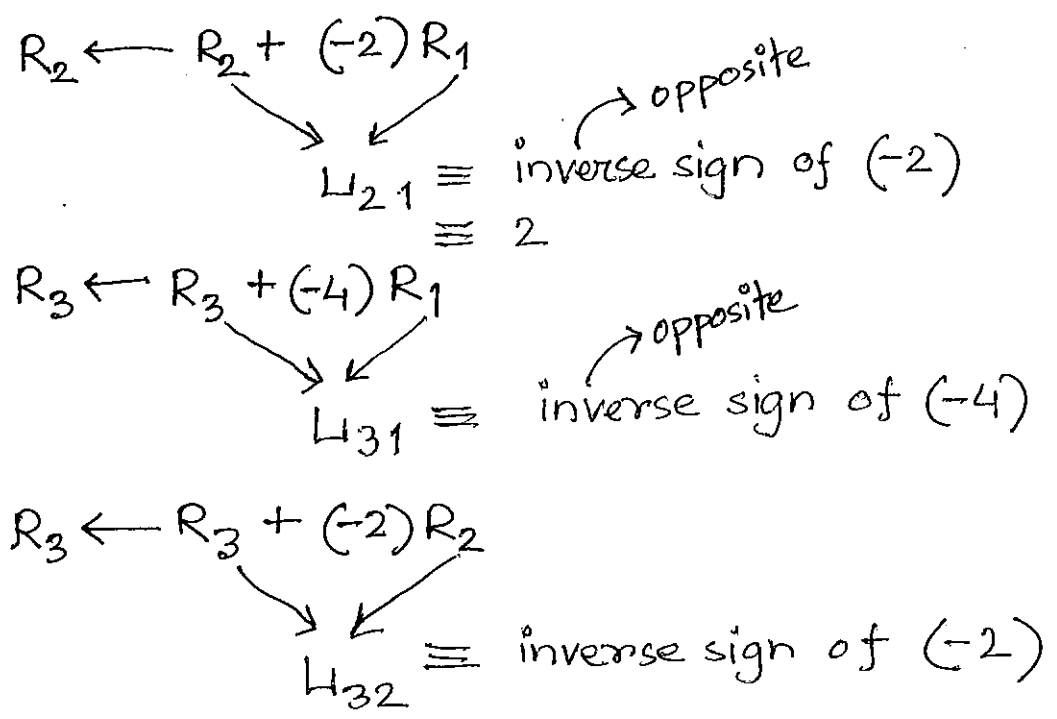
$$A = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 3 & 5 \\ 4 & 6 & 8 \end{bmatrix} \begin{matrix} R_1 \\ R_2 \\ R_3 \end{matrix} = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 3 \\ 0 & 2 & 4 \end{bmatrix} \begin{matrix} R_2 \leftarrow R_2 - 2R_1 \equiv R_2 \leftarrow R_2 + (-2)R_1 \\ R_3 \leftarrow R_3 - 4R_1 \equiv R_3 \leftarrow R_3 + (-4)R_1 \end{matrix}$$

we make it zero

$$\equiv \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 3 \\ 0 & 0 & -2 \end{bmatrix} \begin{matrix} R_3 \leftarrow R_3 - 2R_2 \equiv R_3 \leftarrow R_3 + (-2)R_2 \end{matrix}$$

↳ This is upper-triangular matrix.

Consider all the row-operations.



So, Lower unit triangular matrix L

$$L = \begin{bmatrix} 1 & 0 & 0 \\ L_{21} & 1 & 0 \\ L_{31} & L_{32} & 1 \end{bmatrix}$$

$$\Rightarrow L = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 4 & 2 & 1 \end{bmatrix}$$

So,

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 3 & 5 \\ 4 & 6 & 8 \end{bmatrix} = LU = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 4 & 2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 3 \\ 0 & 0 & -2 \end{bmatrix}$$

A
L
U



Using LU decomposition to solve systems of equations.

Given a linear system

$$AX = B$$

we write B. could write b also; notation only.

⇒

$$LU\underbrace{X}_Y = B$$

as  $A = LU$

⇒

$$LY = B$$

So, we solve

$$LY = B$$

Triangular system (1) ↓ obtain Y

$$UX = Y$$

Triangular system (2) ↓ obtain X

For instance: Given,

$$\begin{matrix}
 \begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} & \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} & = & \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix} \\
 A & X & & B
 \end{matrix}$$

Step 1: Calculate L & U according to the LU decomposition method. Here,

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

Step 2:

Solve  $LY = B$  to obtain  $Y = \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \end{bmatrix}$

so, we obtain,

$$LY = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 2 & 1 & 1 \end{bmatrix} \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix} = B$$

Now, we can use forward substitution to obtain

$$Y_1 = 3, \quad 3Y_1 + Y_2 = 13, \quad \text{and}$$

$$\text{Hence, } Y_2 = 4.$$

$$2Y_1 + Y_2 + Y_3 = 4, \quad \text{and using } Y_1 = 3, \quad \text{and } Y_2 = 4, \quad \text{we obtain}$$

$$Y_3 = -6$$

Step 3: Now, we know  $\begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \\ -6 \end{bmatrix}$ , so, we use it in  $UX = Y$

$$\text{So, } \begin{bmatrix} 1 & 2 & 4 \\ 0 & 2 & 2 \\ 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \\ -6 \end{bmatrix}, \quad \text{We use back-substitution.$$

$$\text{We get } 3x_3 = -6 \Rightarrow x_3 = -2$$

$$2x_2 + 2x_3 = 4, \Rightarrow x_2 = 4, \quad \text{and}$$

$$x_1 + 2x_2 + 4x_3 = 3, \quad \text{so, } x_1 = 3$$

So, we found the solution of  $AX=B$  using LU decomposition, and the solution is as follows:

$$\begin{bmatrix} 1 & 2 & 4 \\ 3 & 8 & 14 \\ 2 & 6 & 13 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ 13 \\ 4 \end{bmatrix}; \text{ We found } x = \begin{bmatrix} 3 \\ 4 \\ -2 \end{bmatrix}$$

□ Practice Problem:

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