Suppose, we have system of linear equation

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

 $a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots + a_{2n}x_n = b_2$
 \vdots
 $a_{n1}x_1 + a_{n2}x_2 + a_{n3}x_3 + \dots + a_{nn}x_n = b_n$

Let
$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & & & \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix} \qquad X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \qquad b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$$b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$Ax=b$$

A Matrix is an array of numbers:

$$A = \begin{bmatrix} 4 & 10 & 5 \\ 5 & 12 & 20 \end{bmatrix}$$

(This one has 2 Rows and 3 Columns)

The Main Diagonal starts at the top left and goes down to the right:

Another example:

A **Transpose** is where we swap entries across the main diagonal (rows become columns) like this:

$$\begin{bmatrix} 6 & 4 & 24 \\ 1 & -9 & 8 \end{bmatrix} = \begin{bmatrix} 6 & 1 \\ 4 & -9 \\ 24 & 8 \end{bmatrix}$$

The main diagonal stays the same.

Square

A **square** matrix has the same number of rows as columns.

$$\begin{bmatrix} 2 & 0 \\ 1 & 8 \end{bmatrix}$$

A square matrix (2 rows, 2 columns)

Also a square matrix (3 rows, 3 columns)

Identity Matrix

An **Identity Matrix** has **1**s on the main diagonal and **0**s everywhere else:

$$\mathbf{I} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

A 3×3 Identity Matrix

- · It is square (same number of rows as columns)
- It can be large or small (2×2, 100×100, ... whatever)
- · Its symbol is the capital letter I

It is the matrix equivalent of the number "1", when we multiply with it the original is unchanged:

$$A \times I = A$$

$$I \times A = A$$

Diagonal Matrix

A diagonal matrix has zero anywhere not on the main diagonal:

Scalar Matrix

A scalar matrix has all main diagonal entries the same, with zero everywhere else:

A scalar matrix

Triangular Matrix

Lower triangular is when all entries above the main diagonal are zero:

A lower triangular matrix

Upper triangular is when all entries below the main diagonal are zero:

An upper triangular matrix

Zero Matrix (Null Matrix)

Zeros just everywhere:

Symmetric

In a Symmetric matrix matching entries either side of the main diagonal are equal, like this:

Symmetric matrix

It must be square, and is equal to its own transpose

$$A = A^{T}$$

Skew-Symmetric Matrix

Square Matrix A is said to be skew-symmetric if $a_{ij} = -a_{ji}$ for all i and j. In other words, we can say that matrix A is said to be skew-symmetric if transpose of matrix A is equal to negative of Matrix A i.e ($A^T = -A$). Note that all the main diagonal elements in skew-symmetric matrix are zero.

Lets take an example of **matrix**
$$A = \begin{bmatrix} 0 & -5 & 4 \\ 5 & 0 & -1 \\ -4 & 1 & 0 \end{bmatrix}$$
.

It is **skew-symmetric matrix** because $a_{ij}=-a_{ji}$ for all i and j. Example, a_{12} = -5 and a_{21} =5 which means $a_{12}=-a_{21}$. Similarly, this condition holds true for all other values of i and j.

We can also verify that **Transpose** of **Matrix A** is equal to negative of **matrix A** i.e $A^T = -A$.

$$A^T = \left[egin{array}{cccc} 0 & 5 & -4 \ -5 & 0 & 1 \ 4 & -1 & 0 \end{array}
ight] ext{ and } A = \left[egin{array}{cccc} 0 & -5 & 4 \ 5 & 0 & -1 \ -4 & 1 & 0 \end{array}
ight].$$

We can clearly see that $A^T = -A$ which makes **A skew-symmetric** matrix.

Hermitian

A Hermitian matrix is symmetric except for the <u>imaginary parts</u> that swap sign across the main diagonal:

Hermitian matrix

See how +i changes to -i and vice versa?

Changing the sign of the second part is called the conjugate, and so the correct definition is:

A Hermitian matrix is equal to its own conjugate transpose:

$$A = \overline{A^T}$$

This also means the main diagonal entries must be purely real (to be their own conjugate).

It is named after French mathematician Charles Hermite.

Orthogonal Matrix: A square matrix with real numbers or values is termed as an orthogonal matrix if its transpose is equal to the inverse matrix of it. In other words, the product of a square orthogonal matrix and its transpose will always give an identity matrix.

Suppose A is the square matrix with real values, of order $n \times n$. Also, let A^T is the transpose matrix of A. Then according to the definition: If, $A^T = A^{-1}$ condition is satisfied, then $\mathbf{A} \times \mathbf{A}^T = \mathbf{I}$

Example:
$$A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

$$A^{-1} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix}, A^{T} = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix}$$

Suppose, we have system of linear equation

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots &+ a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots &+ a_{2n}x_n = b_2 \\ &\vdots \\ a_{n1}x_1 + a_{n2}x_2 + a_{n3}x_3 + \dots &+ a_{nn}x_n = b_n \end{aligned}$$

Let
$$A = \begin{bmatrix} a_{11} & a_{12} & . & . & a_{1n} \\ a_{21} & a_{22} & . & . & a_{2n} \\ . & & & & \\ . & & & & \\ a_{n1} & a_{n2} & . & . & a_{nn} \end{bmatrix} \qquad X = \begin{bmatrix} x_1 \\ x_2 \\ . \\ . \\ x_n \end{bmatrix} \qquad b = \begin{bmatrix} b_1 \\ b_2 \\ . \\ . \\ b_n \end{bmatrix}$$

$$X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$$b = \begin{vmatrix} b_1 \\ b_2 \\ . \\ . \\ b_n \end{vmatrix}$$

$$Ax=b$$

For the given system of linear equations, we have

[A | b] : Augmented matrix

Case 1: if
$$|A| \neq 0$$
 $\implies A^{-1}$ exist.
 $\implies x = A^{-1}$ b

Case 2: if b=0 : Homogeneous system & if $|A| \neq 0$, $\Longrightarrow x = A^{-1}$ b=0, we will get a trivial solution i.e. 0

Consistency and inconsistency of System of Linear equation

Consistency: If solution exists for Ax=b

Inconsistency: If solution does not exists for Ax=b

For Augmented matrix, [A | b], n= no. of unknowns

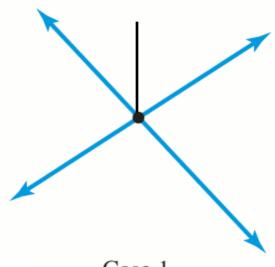
- 1. If $rank(A) = rank([A \mid b]) = n$
- ⇒ System is consistent and has unique solution.
- 2. If rank (A) = rank([A | b]) = n1 < n
- ⇒ system is consistent and has infinite number of solutions.
- 3. If rank $(A) \neq rank([A \mid b])$
 - ⇒ System is inconsistent and has no solution.

A **system of equations** is a group of two or more equations. To solve a system of equations means to find values for the variables that satisfy all of the equations in the system.

Systems of equations can involve any number of equations and variables; however, we will limit ourselves to situations containing two variables in this section.

Graphing a system of two linear equations in two unknowns gives one of three possible situations:

This point represents the solution to the system.

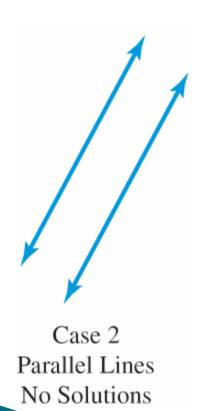


Case 1
Intersecting Lines
One Solution

Case 1: Lines intersecting in a single point. The ordered pair that represents this point is the *unique* solution for the system.

$$3x+y=17$$

 $4x-y=18$



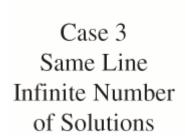
Case 2: Lines that are distinct parallel lines and therefore don't intersect at all. Because the lines have no common points, this means that the system has *no solutions*.

$$3x+2y=5$$

$$6x + 4y = 8$$

Case 3: Two lines that are the same line. The lines have an infinite number of points in common, so the system will have an infinite number of solutions.

$$2x+4y=8$$
$$x+2y=4$$



In numerical methods, there are two types of methods:

- •Direct Method
- •Iterative Method

Example 1: For System of linear equation Ax=b, If A is Diagonal Matrix,

$$\begin{bmatrix} a_{11} & 0 & 0 & 0 \\ 0 & a_{22} & 0 & 0 \\ 0 & 0 & . & 0 \\ 0 & 0 & 0 & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ . \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ . \\ b_n \end{bmatrix}$$

$$x_n = \frac{b_n}{a_{nn}}$$

Example 2: For System of linear equation Ax=b, If A is Lower Triangular Matrix,

$$\begin{bmatrix} l_{11} & 0 & 0 & 0 \\ l_{21} & l_{22} & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots \\ l_{n1} & l_{n2} & \vdots & l_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$x_1 = \frac{b_1}{l_{11}}, x_2 = \frac{1}{l_{22}} (b_2 - l_{12}x_1)$$

$$x_n = \frac{1}{l_{22}} (b_n - l_{1n}x_1 - l_{2n}x_2.....)$$

This is called forward substitution method.

Example 3: For System of linear equation Ax=b, If A is Upper Triangular Matrix,

$$\begin{bmatrix} U_{11} & U_{12} & \dots & U_{1n} \\ 0 & U_{22} & \dots & U_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \vdots & U_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$x_n = \frac{b_n}{U_{nn}}$$

This is called Backward substitution method.

Gauss Elimination Method

A set of *n* equations and *n* unknowns

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

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

$$\vdots \qquad \vdots \qquad \vdots$$

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

(n-1) steps of forward elimination

Step 1

For Equation 2, divide Equation 1 by a_{11} and multiply by a_{21} .

$$\left[\frac{a_{21}}{a_{11}}\right](a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots + a_{1n}x_n = b_1)$$

$$a_{21}x_1 + \frac{a_{21}}{a_{11}}a_{12}x_2 + \dots + \frac{a_{21}}{a_{11}}a_{1n}x_n = \frac{a_{21}}{a_{11}}b_1$$

Subtract the result from Equation 2.

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

$$a_{21}x_1 + \frac{a_{21}}{a_{11}}a_{12}x_2 + \dots + \frac{a_{21}}{a_{11}}a_{1n}x_n = \frac{a_{21}}{a_{11}}b_1$$

$$\left(a_{22} - \frac{a_{21}}{a_{11}}a_{12}\right)x_2 + \dots + \left(a_{2n} - \frac{a_{21}}{a_{11}}a_{1n}\right)x_n = b_2 - \frac{a_{21}}{a_{11}}b_1$$

or
$$a_{22}x_2 + ... + a_{2n}x_n = b_2$$

Repeat this procedure for the remaining equations to reduce the set of equations as

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

$$a'_{22}x_2 + a'_{23}x_3 + \dots + a'_{2n}x_n = b'_2$$

$$a'_{32}x_2 + a'_{33}x_3 + \dots + a'_{3n}x_n = b'_3$$

$$\vdots \qquad \vdots \qquad \vdots$$

$$a'_{n2}x_2 + a'_{n3}x_3 + \dots + a'_{nn}x_n = b'_n$$

End of Step 1

Step 2

Now eliminate x2 from the last (n-2) equations.

$$a_{11}x_{1} + a_{12}x_{2} + a_{13}x_{3} + \dots + a_{1n}x_{n} = b_{1}$$

$$a'_{22}x_{2} + a'_{23}x_{3} + \dots + a'_{2n}x_{n} = b'_{2}$$

$$a''_{33}x_{3} + \dots + a''_{3n}x_{n} = b''_{3}$$

$$\vdots \qquad \vdots$$

$$a''_{n3}x_{3} + \dots + a''_{nn}x_{n} = b''_{n}$$

End of Step 2

Repeat this process till we finally obtain upper triangular form:

$$a_{11}x_{1} + a_{12}x_{2} + a_{13}x_{3} + \dots + a_{1n}x_{n} = b_{1}$$

$$a'_{22}x_{2} + a'_{23}x_{3} + \dots + a'_{2n}x_{n} = b'_{2}$$

$$a''_{33}x_{3} + \dots + a''_{3n}x_{n} = b''_{3}$$

$$\vdots$$

$$a_{nn}^{(n-1)}x_{n} = b_{n}^{(n-1)}$$

End of Step (n-1)

Matrix Form at End of Forward Elimination

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ 0 & a'_{22} & a'_{23} & \cdots & a'_{2n} \\ 0 & 0 & a''_{33} & \cdots & a''_{3n} \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & 0 & 0 & a_{nn}^{(n-1)} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b'_2 \\ b''_3 \\ \vdots \\ b'^{(n-1)} \\ b'^{(n-1)} \end{bmatrix}$$

Back Substitution Starting Eqns

$$a_{11}x_{1} + a_{12}x_{2} + a_{13}x_{3} + \dots + a_{1n}x_{n} = b_{1}$$

$$a_{22}x_{2} + a_{23}x_{3} + \dots + a_{2n}x_{n} = b_{2}$$

$$a_{33}x_{3} + \dots + a_{n}x_{n} = b_{3}$$

$$\vdots$$

$$a_{nn}^{(n-1)}x_{n} = b_{n}^{(n-1)}$$

Where, $a_{nn}^{(n-1)}$ indicates that the element a_{nn} has changed (n-1) times

Back Substitution

The required solution is obtained from the last equation because it has only one unknown

$$x_n = \frac{b_n^{(n-1)}}{a_{nn}^{(n-1)}}$$

Then, X_{n-1} , X_{n-2} $X_{1 \text{ can be calculated.}}$

Example-1 Use Gauss Elimination method to solve

$$2x_1+x_2+x_3=10$$
(1)

$$3x_1+2x_2+3x_3=18$$
(2)

$$x_1 + 4x_2 + 9x_3 = 16$$
(3)

Solution: At first eliminate x_1 from eq. (2) and (3)

Multiply eq.(1) by (3/2) and subtract from eq(2)

$$3x_1 + 2x_2 + 3x_3 = 18$$

-
$$[3x_1+(3/2)x_2+(3/2)x_3=10(3/2)]$$

$$(2-3/2)x_2 + (3-3/2)x_3 = 18-15$$

 $(1/2)x_2 + (3/2)x_3 = 3$

$$x_2 + 3x_3 = 6$$
(4)

Multiply eq.(1) by (1/2) and subtract from eq(3)

$$x_1 + 4x_2 + 9x_3 = 16$$

-
$$[x_1+(1/2)x_2+(1/2)x_3=10(1/2)]$$

$$(4-1/2)x_2 + (9-1/2)x_3 = 16-5$$

$$(7/2)x_2+(17/2)x_3=11$$

$$7x_2 + 17x_3 = 22$$
(5)

Solution (Cont.....)

```
Now eliminate x_2 from eq. (5)
Multiply eq.(4) by (7) and subtract from eq(5)
          7x_2 + 17x_3 = 22
    - [7x_2 + 21x_3 = 42]
          (17-21)x_3=22-42
          (-4)x_3 = -20
         x_3 = 5
Now from eq.(4)
          x_2 + 3x_3 = 6
         x_2 + 3 \times 5 = 6
x_2 = 6 - 15 = -9
Now from eq.(1)
          2x_1+x_2+x_3=10
         2x_1 - 9 + 5 = 10
```

Pivot

If, in the system of linear equation pivot (a_{11}) is zero or very small compared to other coefficients of the equation, then we can find the largest available coefficient in the column below the pivot equation & then interchange the two rows.

In this way we can find the pivot equation with a non zero pivot.

Note: if we search both column and rows for largest element, the procedure is called "Complete pivoting"

Example-1 Use Gauss Elimination method with partial pivoting to solve

$$x_1+2x_2+4x_3=9$$
(1)
 $2x_1+5x_2-3x_3=9$ (2)
 $5x_1+12x_2-2x_3=25$ (3)

$$\begin{bmatrix} 1 & 2 & 4 & | & 9 \\ 2 & 5 & -3 & | & 9 \\ 5 & 12 & -2 & | & 25 \end{bmatrix}$$

$$\begin{bmatrix} 5 & 12 & -2 & | & 25 \\ 2 & 5 & -3 & | & 9 \\ 1 & 2 & 4 & | & 9 \end{bmatrix} R_1 \leftrightarrow R_3$$

$$\begin{bmatrix} 1 & 12/5 & -2/5 & | & 5 \\ 2 & 5 & -3 & | & 9 \\ 1 & 2 & 4 & | & 9 \end{bmatrix} R_1 \to R_1/5$$

$$\begin{bmatrix} 1 & 12/5 & -2/5 & | & 5 \\ 2 & 4 & | & 9 \end{bmatrix} R_1 \to R_1/5$$

$$\begin{bmatrix} 1 & 12/5 & -2/5 & | & 5 \\ 0 & 1/5 & -11/5 & | & -1 \\ 0 & -2/5 & 22/5 & | & 4 \end{bmatrix} R_2 \to R_2 - 2R_1, R_3 \to R_3 - R_1$$

$$\begin{bmatrix} 1 & 12/5 & -2/5 & | & 5 \\ 0 & -2/5 & 22/5 & | & 4 \\ 0 & 1/5 & -11/5 & | & -1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 12/5 & -2/5 & | & 5 \\ 0 & 1 & -11 & | & -10 \\ 0 & 1/5 & -11/5 & | & -1 \end{bmatrix} R_2 \rightarrow R_2 / (-2/5)$$

$$\begin{bmatrix} 1 & 12/5 & -2/5 & | & 5 \\ 0 & 1 & -11 & | & -10 \\ 0 & 0 & 0 & | & 1 \end{bmatrix} R_3 \to R_3 - R_2(1/5)$$

Since
$$rank(A) \neq rank(A \mid b)$$

 $2 \neq 3$
So no solution exists

Practice Problems

1. Solve the following linear system of equation using Gauss elimination method:

$$y+z=2$$

$$2x+3z=5$$

$$x+y+z=3$$

2. Use Gauss Elimination method with partial pivoting to solve

$$2x+3y+z=9$$

$$x+2y+3z=6$$

$$3x+y+2z=8$$

Suggested books

1. Numerical Methods by S.R.K Lyenger & R.K. Jain.

2. Numerical Analysis by Richard L. Burden.

3. Introductory methods of Numerical analysis by **S.S. Sastry**.

Thank you