#### **Linear Algebra**

## Session-1:

**Matrix:** A matrix is an array of numbers. A set of mn numbers (real or complex) arranged in a rectangular array having m rows (horizontal lines) and n columns (vertical lines), the numbers /entries being enclosed by brackets  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  or  $\begin{pmatrix} 0 \\ 1 \end{bmatrix}$  is called an  $m \times n$  matrix (read as "m by n matrix"). In what follows, a matrix is denoted by an upper-case alphabet in boldface (e.g., A), and its (i, j) th element (the element at the *i* th row and *j* th column) is denoted by the corresponding lower-case alphabet with subscripts ij

(e.g. $a_{ij}$ ). Specifically, an  $m \times n$  matrix A contains m rows and n columns and can be expressed as

$$\boldsymbol{A} = \begin{bmatrix} \boldsymbol{a}^{11} & \boldsymbol{a}^{12} \cdots & \boldsymbol{a}^{1n} \\ \vdots & \ddots & \vdots \\ \boldsymbol{a}^{m1} & \boldsymbol{a}^{m2} \cdots & \boldsymbol{a}^{mn} \end{bmatrix}$$

# **Basic Concepts:**

1) Matrix Addition: Let  $\mathbf{A} = \begin{bmatrix} a^{ij} \end{bmatrix} \mathbf{B} = \begin{bmatrix} b^{ij} \end{bmatrix}$  be two matrices of same order, then their sum, written

A+B is obtained by adding the corresponding entries. Matrices of different orders cannot be added.

Example: If  $\mathbf{A} = \begin{bmatrix} 1 & 3 & -5 \\ 5 & 0 & 9 \end{bmatrix}$  and  $\mathbf{B} = \begin{bmatrix} 3 & 5 & 8 \\ -8 & 2 & 7 \end{bmatrix}$  then  $\mathbf{A} + \mathbf{B} = \begin{bmatrix} 4 & 8 & 3 \\ -3 & 2 & 16 \end{bmatrix}$ 

2) Scalar Multiplication: The product of any  $m \times n$  matrix  $\mathbf{A} = \begin{bmatrix} a^{ij} \end{bmatrix}$  and any scalar  $\lambda$ , written

 $\lambda \mathbf{A}$ , is the  $m \times n$  matrix  $\lambda \mathbf{A} = \begin{bmatrix} \lambda a^{ij} \end{bmatrix}$  obtained by multiplying each entry in  $\mathbf{A}$  by  $\lambda$ .

Example: If  $\mathbf{A} = \begin{bmatrix} 1 & 3 & -5 \\ 5 & 0 & 9 \end{bmatrix}$  and  $\lambda = -5$  then  $\lambda \mathbf{A} = \begin{bmatrix} -5 & -15 & 25 \\ -25 & 0 & -45 \end{bmatrix}$ 

3) Properties of Matrix Addition:

a) Matrix Addition is commutative. That is A+B=B+A.

b) Matrix Addition is associative. That is (A+B) + C = A + (B+C).

c) For any matrix A, there exists a null matrix 0 of same order such that A+0=0+A.

d) For any matrix A, there exists a matrix -A of same order such that A + (-A) = (-A) + A = 0.

Equality of Matrices: Two matrices **A** and **B** are said to be **equal** iff they have the same order and their corresponding elements are equal. Thus, if  $A = \begin{bmatrix} a^{ij} \end{bmatrix}$  and  $B = \begin{bmatrix} b^{ij} \end{bmatrix}$  then A=B iff  $a^{ij} = b^{ij}$ ,

for all i and j.

**Matrix Multiplication:** Two matrices **A** and **B** are said to be conformable for the product (in this very order of A and B) if the number of columns in **A** (called the prefactor) is equal to the number of rows in B (called the post factor).

Here (i, j) th element of AB

=Sum of the products of the elements of ith row of A

with the corresponding jth column of  ${\bf B}$ 



#### **Matrix Multiplication**

**Solved Example**: If  $\mathbf{A} = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$  and  $\mathbf{B} = \begin{bmatrix} k & l \\ m & n \\ o & p \end{bmatrix}$ , then  $\mathbf{AB} = \begin{bmatrix} ak + bm + co & al + bn + cp \\ dk + em + fo & dl + en + fp \\ gk + hm + io & gl + hn + ip \end{bmatrix}$ 

# **Properties of Matrix Multiplication:**

a) Matrix Multiplication is not commutative: That is **AB**≠**BA**.

b) Matrix Multiplication is associative: That is (AB) C=A (BC).

c) Matrix Multiplication is distributive with respect to Matrix Addition : That is A (B+C) =AB+AC.

d) If **A** and **I** are square matrices of same order then **AI=IA=A**.

e) Power of a matrix: If A is a square matrix of order n, then

$$\mathbf{A} \times \mathbf{A} = \mathbf{A}^2$$
,  $\mathbf{A} \times \mathbf{A} \times \mathbf{A} = \mathbf{A}^3$ ,  $\mathbf{A} \times \mathbf{A} \times \mathbf{A} \times \mathbf{A} \times \mathbf{A} \wedge \Lambda \wedge \Lambda \wedge \Lambda$  (m times)  $= \mathbf{A}^m$ .

f) For a positive integer n,  $I^n = I$ .

g) **AB=0** does not necessarily imply **A=0** or **B=0** or **BA=0**.

Example: Let 
$$A = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}$$
 and  $B = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$  then  $AB = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ ,  $BA = \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix}$ 

h) AC=AD does not necessarily imply C=D (even when  $A \neq 0$ ).

Example: Let  $\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}$ ,  $\mathbf{C} = \begin{bmatrix} 2 & 1 \\ 2 & 2 \end{bmatrix}$  and  $\mathbf{D} = \begin{bmatrix} 3 & 0 \\ 1 & 3 \end{bmatrix}$  then  $\mathbf{A}\mathbf{C} = \begin{bmatrix} 4 & 3 \\ 8 & 6 \end{bmatrix} = \mathbf{A}\mathbf{D}$  but  $\mathbf{C} \neq \mathbf{D}$ .

**Transposition:** Given a matrix A, then the matrix obtained from A by changing its rows to columns and columns into rows is called the **transpose** of A and denoted by $A^T$ . Clearly if the order of A is  $m \times n$ .

then the order of  $A^T$  is  $n \times m$  and (i, j) th element of  $A^T = (j, i)$  th element of A.

**Solved Example:** If 
$$\mathbf{A} = \begin{bmatrix} 1 & 3 & -5 \\ 5 & 0 & 9 \end{bmatrix}$$
 then  $A^T = \begin{bmatrix} 1 & 5 \\ 3 & 0 \\ -5 & 9 \end{bmatrix}$ .

### **Properties of Transpose:**

a) 
$$(A^{T})^{T} = A$$
  
b)  $(A + B)^{T} = A^{T} + B^{T}$   
c) $(AB)^{T} = B^{T}A^{T}$ .

## **Special Matrices:**

a) Row and Column Matrix: A matrix having a single row is called a row matrix or row vector.

Example: A = [2 -9 11 4].

A matrix having a single column is called a **column matrix** or **column vector**.

Example:  $\mathbf{B} = \begin{bmatrix} 2 \\ -6 \\ 5 \\ 13 \end{bmatrix}$ .

b) Real Matrix: A matrix is said to be real if all its elements are real numbers.

Example:  $\mathbf{A} = \begin{bmatrix} .5 & \sqrt{3} & 0 \\ -6 & 11 & \sqrt[3]{7} \end{bmatrix}$ .

c) Square Matrix: A matrix in which the number of rows is equal to the number of columns is called a square matrix, otherwise, it is said to be a rectangular matrix.

Example:  $\mathbf{A} = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$ , the elements a, e, i are called its diagonal elements and the diagonal along which these elements lie is called the **principal diagonal** or **leading diagonal**.

**d)** Trace of a (square) Matrix: The sum of the diagonal elements of a square matrix **A** is called the trace of **A**. Thus, the trace of the n-rowed square matrix  $A = \begin{bmatrix} a^{ij} \end{bmatrix}$  is

$$a_{11} + a_{22} + a_{33}\Lambda \Lambda + a_{nn} = \sum_{i=1}^{n} a_{ii}$$

e) **Diagonal Matrix:** A square matrix in which all non-diagonal elements are zero is called a **diagonal matrix**.

Example:  $\mathbf{A} = \begin{bmatrix} -3 & 0 & 0 \\ 0 & 10 & 0 \\ 0 & 0 & 7 \end{bmatrix}$ .

**f)** Scalar Matrix: A diagonal matrix in which all the diagonal elements are equal is called a scalar matrix.

Example:  $\mathbf{A} = \begin{bmatrix} -3 & 0 & 0 \\ 0 & -3 & 0 \\ 0 & 0 & -3 \end{bmatrix}$ .

**g)** Unit Matrix/Identity Matrix: A scalar matrix, in which each diagonal element is unit, is called a **unit matrix/identity matrix**.

Example:  $\mathbf{A} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ .

**h)** Null Matrix/Zero Matrix: A matrix, in which each element is zero, is called a null/zero matrix.

Example: If  $\mathbf{A} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ .

# **Special Matrices (Continued):**

**i) Triangular Matrix:** A square matrix in which all the elements either below or above the principal diagonal are zero is called a triangular marix. Thus a triangular matrix is either upper triangular or lower triangular.

1) Upper Triangular Matrix: A square matrix in which all the elements below the principal

diagonal are zero is called an upper triangular matrix.

Example:  $\mathbf{A} = \begin{bmatrix} -5 & 11 & 2 \\ 0 & 10 & 8 \\ 0 & 0 & 27 \end{bmatrix}$ .

2) Lower Triangular Matrix: A square matrix in which all the elements above the principal

diagonal are zero is called a lower triangular matrix.

Example:  $\mathbf{A} = \begin{bmatrix} -8 & 0 & 0 \\ -3 & 20 & 0 \\ 9 & 2 & 7 \end{bmatrix}$ .

**j)** Symmetric Matrix: A square matrix  $\mathbf{A} = [a_{ij}]$  is said to be symmetric when  $a_{ij} = a_{ji}$ 

for all i and j.That is,  $A^T = A$ .

Example: 
$$\mathbf{A} = \begin{bmatrix} a & e & f \\ e & b & g \\ f & g & c \end{bmatrix}$$
.

# **k)** Skew-Symmetric Matrix (or Anti- Symmetric Matrix): A square matrix $\mathbf{A} = [a_{ij}]$ is said to be skew-symmetric

when 1)  $a_{ij} = -a_{ji}$  for all i and j

and 2) All diagonal elements  $a_{ii}=0$ .

That is, 
$$A^T = -A$$
.

Example: 
$$\mathbf{A} = \begin{bmatrix} 0 & e & f \\ -e & 0 & -g \\ -f & g & 0 \end{bmatrix}.$$

**I)** Inverse of a Square Matrix: The inverse of an  $n \times n$  square matrix  $\mathbf{A} = [a_{ij}]$  is denoted by  $A^{-1}$  and is an  $n \times n$  matrix such that  $AA^{-1} = A^{-1}A = I$ , where **I** is the  $n \times n$ unit matrix.

If A has an inverse, then A is called a **non-singular matrix** (That is, Det (A)  $\neq 0$ ).

If A has no inverse, then A is called a **singular matrix** (That is, Det (A) = 0).

Inner Product of Vectors: If a is a row vector and b is a column vector, both with n components then the matrix multiplication gives a  $1 \times 1$  matrix, which is a single entry. This product is called the inner product or dot product of **a** and **b** and is denoted by **a.b**.

Let 
$$\mathbf{a} = \begin{bmatrix} a_1 & a_2 & \dots & a_{n-1} \\ a_1 & a_2 & \dots & a_{n-1} \end{bmatrix}$$
 and  $\mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ \vdots \\ b_{n-1} \\ b_n \end{bmatrix}$   
$$a_1 b_1 + a_2 b_2 + \Lambda \Lambda \Lambda + a_n b_n = \sum_{i=1}^n a_i b_i$$
  
Then  $\mathbf{a} \mathbf{b} =$ 

Then a.b=

Example: 
$$\mathbf{a} = \begin{bmatrix} 2 & -3 & 11 \end{bmatrix}$$
,  $\mathbf{b} = \begin{bmatrix} 2 \\ -6 \\ 5 \end{bmatrix}$ , then  $\mathbf{a} \cdot \mathbf{b} = \begin{pmatrix} 2 \times 2 \end{pmatrix} + \begin{pmatrix} -3 \end{pmatrix} \times \begin{pmatrix} -6 \end{pmatrix} + \begin{pmatrix} 11 \times 5 \end{pmatrix} = 77$ .

# Matrix Product in Terms of Row and Column Vectors: Let A be a

matrix of order  $m \times n$  and **B** be a matrix of order  $n \times p$  represented as:

$$\mathbf{A} = \begin{bmatrix} a^{11} & a^{12} \Lambda & a^{1n} \\ a^{21} & a^{22} \Lambda & a^{2n} \\ \Lambda & \Lambda & \Lambda & \Lambda \\ a^{m1} & a^{m2} \Lambda & a^{mn} \end{bmatrix} = \begin{bmatrix} a^{1} \\ a^{2} \\ N \\ N \\ a^{m-1} \\ a^{m} \end{bmatrix}, \text{ where } a_{1}, a_{2}, \dots, a_{m} \text{ are row vectors of order}$$

 $1 \times n$ .

and 
$$\mathbf{B} = \begin{bmatrix} b^{11} & b^{12} \Lambda & b^{1p} \\ b^{21} & b^{22} \Lambda & b^{2p} \\ \Lambda & \Lambda & \Lambda \\ b^{n1} & b^{n2} \Lambda & b^{np} \end{bmatrix} = \begin{bmatrix} b^{1} & b^{2} & \Lambda & \Lambda & b^{p-1} & b^{p} \end{bmatrix}$$
, where  $b_{1}, b_{2}, \dots, \dots, b_{p}$ 

are column vectors of order  $n \times 1$ .

Then 
$$\mathbf{AB} = \begin{bmatrix} a^1 \cdot b^1 & a^1 \cdot b^2 \wedge & a^1 \cdot b^p \\ a^2 \cdot b^1 & a^2 \cdot b^2 \wedge & a^2 \cdot b^p \\ \wedge & \wedge & \wedge \\ a^m \cdot b^1 & a^m \cdot b^2 \wedge & a^m \cdot b^p \end{bmatrix}$$
 is in terms of row and column vectors.

**Solved Example:** Express the product of the matrices in terms of row and column vectors where

$$A = \begin{bmatrix} 4 & 3\\ 7 & 2\\ 9 & 0 \end{bmatrix} and B = \begin{bmatrix} 2 & 5\\ 1 & 6 \end{bmatrix}.$$

Solution: 
$$A = \begin{bmatrix} 4 & 3 \\ 7 & 2 \\ 9 & 0 \end{bmatrix} = \begin{bmatrix} a^{1} \\ a^{2} \\ a^{3} \end{bmatrix}$$
  
where  $a_{1} = \begin{bmatrix} 4 & 3 \end{bmatrix}, a_{2} = \begin{bmatrix} 7 & 2 \end{bmatrix}, and a_{3} = \begin{bmatrix} 9 & 0 \end{bmatrix}$   
and  $B = \begin{bmatrix} 2 & 5 \\ 1 & 6 \end{bmatrix} = \begin{bmatrix} b^{1} & b^{2} \end{bmatrix}$   
where  $b_{1} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$  and  $b_{2} = \begin{bmatrix} 5 \\ 6 \end{bmatrix}$   
Then  $AB = \begin{bmatrix} a^{1} \cdot b^{1} & a^{1} \cdot b^{2} \\ a^{2} \cdot b^{1} & a^{2} \cdot b^{2} \\ a^{3} \cdot b^{1} & a^{3} \cdot b^{2} \end{bmatrix}$   
 $= \begin{bmatrix} (4 \times 2 + 3 \times 1) & (4 \times 5 + 3 \times 6) \\ (7 \times 2 + 2 \times 1) & (7 \times 5 + 2 \times 6) \\ (9 \times 2 + 0 \times 1) & (9 \times 5 + 0 \times 6) \end{bmatrix} = \begin{bmatrix} 11 & 38 \\ 16 & 47 \\ 18 & 45 \end{bmatrix}$ .

#### **System of Linear Equations:**

The concept of a matrix was historically introduced to simplify the solution of linear systems although they today have much greater and broad-reaching applications. Let's see how a linear system can be represented using a matrix.

Consider a general system of m linear equations with n unknowns:  $a_{11}x_1 + a_{12}x_2 + A_{12}x_2 + A_{13}A_{13}A_{14} + a_{14}x_{14} = b_1$ 

This system is equivalent to a matrix equation of the form

Where A is an  $m \times n$  matrix, X is a column matrix with n entries, and B is a column matrix with m entries.

$$A = \begin{bmatrix} a^{11} & a^{12} \cdots & a^{1n} \\ a^{21} & a^{22} \cdots & a^{2n} \\ \cdots & \cdots & \cdots \\ a^{m1} & a^{m2} \cdots & a^{mn} \end{bmatrix}, X = \begin{bmatrix} x_1 \\ x_2 \\ N \\ x_n \end{bmatrix}, and B = \begin{bmatrix} b_1 \\ b_2 \\ N \\ b_m \end{bmatrix}$$

Clearly our manner of defining matrix multiplication is used in representing the linear system in this fashion because now the product of the matrix A (called the **coefficient matrix**) and the matrix x gives us precisely the matrix b (called the **right hand side matrix**).

$$\boldsymbol{K} = \begin{bmatrix} a^{11} & a^{12} \cdots & a^{1n} & \vdots & b^{1} \\ a^{21} & a^{22} \cdots & a^{2n} & \vdots & b^{2} \\ \cdots & \cdots & \cdots & \vdots & \cdots \\ a^{m1} & a^{m2} \cdots & a^{mn} & \vdots & b^{m} \end{bmatrix}$$
 is called the **Augmented Matrix**.

If the elements in **B** are not all zero, then (i) is called **Non-homogeneous System**, otherwise it is called **Homogeneous System**. A solution of (1) is defined by a set of values of the variables  $x_1, x_2, x_3, \ldots, x_n$  which satisfy simultaneously all the equations of (1). If the

system given by (1) has a solution, it is called a **consistent system**, otherwise the system is called **inconsistent**. In fact, a consistent system has either unique solution or infinitely many solutions. Obviously any homogeneous system of equations is always a consistent system, since  $x_i=0$  for all i is a solution which is known as trivial solution.

### **Gauss Elimination Method:**

**Solution by Gauss Elimination:** The Gauss Elimination is a standard method for solving linear equations. It is a systematic elimination. It is a reduction to triangular form, from which we shall obtain the values of the unknowns by substitution. This method will be very clear by taking the following problems.

Solved Example: Solve the following system by Gauss Elimination:

 $-x_1 + x_2 - x_3 = 0$  $10x_2 + 25x_3 = 90$  $20x_1 + 10x_2 = 80$ 

Solution's We2ctarY3wTife the system and its augmented matrix side by side:

$-x_1 + x_2 - x_3 = 0$					
$10x_2 + 25x_3 = 90$	[	1	-1	1 <sup>N</sup> 0	-
$20x_1 + 10x_2 = 80$	K =	-1	1	$-1^{N_{0}}$	
		0	10	25 <sup>N</sup> 90	
		20	10	0 <sub>N</sub> 80	_

#### **Step -1: Elimination of** $x_1$

The first equation is called pivot equation and the coefficient 1 of its  $x_1$  –term the pivot. Using pivot equation, we can eliminate  $x_1$  from other equation. For this, we have to do:

Adding 1 times pivot equation to the  $2^{nd}$  and -20 times pivot equation to the  $4^{th}$  equation, the result  $\bar{rs}$ .

0 = 0

$10x_2 + 25x_3 = 90$	[	1	-1	1 <sup>N</sup> 0	]
$30x_2 - 20x_3 = 80$	<i>v</i> _	0	0	0 <sup>N</sup> 0	
2 5	Λ -	0	10	2 <sup>N</sup> 90	
		0	30	-20 <sup>N</sup> 80	

# **Step -2: Elimination of** $x_2$

The 1<sup>st</sup> equation which has just served as the pivot equation remains unchanged. The 2<sup>nd</sup> equation is the next pivot equation, but for this problem 2<sup>nd</sup> equation contains no  $x_2$  –term.

Now it is required to change the order of the equation to get a non-zero pivot. We can put  $2^{nd}$  equation at the end and the  $3^{rd}$  and  $4^{th}$  equations one place up. This is called **partial pivoting**. We get:

$$\begin{aligned} x_1 - x_2 + x_3 &= 0 \\ 10x_2 + 25x_3 &= 90 \\ 30x_2 - 20x_3 &= 80 \\ 0 &= 0 \end{aligned} \qquad \qquad K = \begin{bmatrix} 1 & -1 & 1 & ^{N}0 \\ 0 & 10 & 25 & ^{N}90 \\ 0 & 30 & -20 & ^{N}80 \\ 0 & 0 & 0 & ^{N}0 \end{bmatrix}. \end{aligned}$$

Adding-Brine pivot equation to the 3rd equation, the result is:

[	1	-1	1 <sup>N</sup> 0	]
	0	10	25 <sup>N</sup> 90	
K =	0	0	$-95 \ ^{N}-190$	ŀ
	0	0	0 <sup>N</sup> 0	
	K =	$K = \begin{bmatrix} & 1 \\ & 0 \\ & 0 \\ & 0 \end{bmatrix}$	$K = \begin{bmatrix} 1 & -1 \\ 0 & 10 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$	$K = \begin{bmatrix} 1 & -1 & 1 & {}^{N} 0 \\ 0 & 10 & 25 & {}^{N} 90 \\ 0 & 0 & -95 & {}^{N} - 190 \\ 0 & 0 & 0 & {}^{N} 0 \end{bmatrix}$

**Back Substitution:** Determinations of  $\underline{x}_3$ ,  $\underline{x}_{90}$  and  $\underline{x}_3$  (in this order)

$$10x_2 + 25x_3 = 90 \Longrightarrow x_2 = 4$$
$$x_1 - x_2 + x_3 = 0 \Longrightarrow x_1 = 2$$

The solution of the system is:

$$x_1 = 2, x_2 = 4, and x_3 = 2.$$