

## CHAPTER

## 3



## Matrices

## Special Type of Matrices

**1. Row Matrix (Row vector):**  $A = [a_{11}, a_{12}, \dots, a_{1n}]$  i.e., row matrix has exactly one row.

**2. Column Matrix (Column vector):**  $A = \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix}$  i.e., column

matrix has exactly one column.

**3. Zero or Null Matrix:** ( $A = O_{m \times n}$ ), an  $m \times n$  matrix whose all entries are zero.

**4. Horizontal Matrix:** A matrix of order  $m \times n$  is a horizontal matrix if  $n > m$ .

**5. Vertical Matrix:** A matrix of order  $m \times n$  is a vertical matrix if  $m > n$ .

**6. Square Matrix:** (Order  $n$ ) if number of rows = number of column, then matrix is a square matrix.

## Key Note

- The pair of elements  $a_{ij}$  and  $a_{ji}$  are called Conjugate Elements.
- The elements  $a_{11}, a_{22}, a_{33}, \dots, a_{mm}$  are called Diagonal Elements. the line along which the diagonal elements lie is called "Principal or Leading diagonal." The quantity  $\sum a_{ii} =$  trace of the matrix written as,  $t_r(A)$ .

**7. Unit/Identity Matrix:** A square matrix, in which every non-diagonal element is zero and every diagonal element is 1, is called unit matrix or an identity matrix,

$$\text{i.e. } a_{ij} = \begin{cases} 0, & \text{when } i \neq j \\ 1, & \text{when } i = j \end{cases}$$

**8. Upper Triangular Matrix:** A square matrix  $A = [a_{ij}]_{n \times n}$  is called a upper triangular matrix, if  $a_{ij} = 0, \forall i > j$ .

**9. Lower Triangular Matrix:** A square matrix  $A = [a_{ij}]_{n \times n}$  is called a lower triangular matrix, if  $a_{ij} = 0, \forall i < j$ .

**10. Submatrix:** A matrix which is obtained from a given matrix by deleting any number of rows or columns or both is called a submatrix of the given matrix.

**11. Equal Matrices:** Two matrices  $A$  and  $B$  are said to be equal, if both having same order and corresponding elements of the matrices are equal.

**12. Principal Diagonal of a Matrix:** In a square matrix, the diagonal from the first element of the first row to the last element of the last row is called the principal diagonal of a matrix.

e.g. If  $A = \begin{bmatrix} 1 & 2 & 3 \\ 7 & 6 & 5 \\ 1 & 1 & 2 \end{bmatrix}$ , the principal diagonal of  $A$  is 1, 6, 2.

**13. Singular Matrix:** A square matrix  $A$  is said to be singular matrix, if determinant of  $A$  denoted by  $\det(A)$  or  $|A|$  is zero, i.e.  $|A| = 0$ , otherwise it is a non-singular matrix.

## Equality of Matrices

Let  $A = [a_{ij}]$  &  $B = [b_{ij}]$  are equal if,

1. Both have the same order.
2.  $a_{ij} = b_{ij}$  for each pair of  $i$  &  $j$ .

## Algebra of Matrices

**Addition:**  $A + B = [a_{ij} + b_{ij}]$  where  $A$  &  $B$  are of the same order.

1. **Addition of matrices is commutative:**  $A + B = B + A$ .
2. **Matrix addition is associative:**  $(A + B) + C = A + (B + C)$ .

## Multiplication of a Matrix By a Scalar

$$\text{If } A = \begin{bmatrix} a & b & c \\ b & c & a \\ c & a & b \end{bmatrix}, \text{ then } kA = \begin{bmatrix} ka & kb & kc \\ kb & kc & ka \\ kc & ka & kb \end{bmatrix}$$

## Multiplication of Matrices (Row by Column)

Let  $A$  be a matrix of order  $m \times n$  and  $B$  be a matrix of order  $n \times p$  then the matrix multiplication  $AB$  is possible if and only if  $n = p$ .

Let  $A_{m \times n} = [a_{ij}]$  and  $B_{n \times p} = [b_{ij}]$ , then order of  $AB$  is  $m \times p$  and

$$(AB)_{ij} = \sum_{r=1}^n a_{ir} b_{rj}$$

## Characteristic Equation

Let  $A$  be a square matrix. Then the polynomial  $|A - xI|$  is called as characteristic polynomial of  $A$  & the equation  $|A - xI| = 0$  is called characteristic equation of  $A$ .

## Properties of Matrix Multiplication

- If  $A$  and  $B$  are two matrices such that
  - $AB = BA$  then  $A$  and  $B$  are said to commute
  - $AB = -BA$  then  $A$  and  $B$  are said to anticommute
- Matrix Multiplication is Associative:** If  $A$ ,  $B$  &  $C$  are conformable for the product  $AB$  &  $BC$ , then  $(AB)C = A(BC)$ .
- Distributivity:** 
$$\left. \begin{aligned} A(B + C) &= AB + AC \\ (A + B)C &= AC + BC \end{aligned} \right\}, \text{ provided } A, B \text{ and } C$$
 are conformable for respective products.

## Positive Integral Powers of a Square Matrix

- $A^m A^n = A^{m+n}$
- $(A^m)^n = A^{mn} = (A^n)^m$
- $I^m = I$ ,  $n \in \mathbb{N}$

## Orthogonal Matrix

A square matrix is said to be orthogonal matrix if  $AA^T = I$ .

### Key Note

- The determinant value of orthogonal matrix is either 1 or  $-1$ .  
Hence orthogonal matrix is always invertible.
- $AA^T = I = A^T A$ . Hence  $A^{-1} = A^T$ .

## Some Square Matrices

- Idempotent Matrix:** A square matrix is idempotent provided  $A^2 = A$ . For idempotent matrix note the following:
  - $A^n = A \forall n \geq 2, n \in \mathbb{N}$ .
  - determinant value of idempotent matrix is either 0 or 1.
  - If idempotent matrix is invertible then its inverse will be identity matrix i.e.  $I$ .
- Periodic Matrix:** A square matrix which satisfies the relation  $A^{K+1} = A$ , for some positive integer  $K$ , is a periodic matrix. The period of the matrix is the least value of  $K$  for which this holds true.  
Note that period of an idempotent matrix is 1.
- Nilpotent Matrix:** A square matrix is said to be nilpotent matrix of order  $m$ ,  $m \in \mathbb{N}$ , if  $A^m = O$ ,  $A^{m-1} \neq O$ .  
Note that a nilpotent matrix will not be invertible.
- Involuntary Matrix:** If  $A^2 = I$ , the matrix is said to be an involuntary matrix.  
Note that  $A = A^{-1}$  for an involuntary matrix.

- If  $A$  and  $B$  are square matrices of same order and  $AB = BA$  then  
$$(A + B)^n = {}^n C_0 A^n + {}^n C_1 A^{n-1} B + {}^n C_2 A^{n-2} B^2 + \dots + {}^n C_n B^n.$$

## Transpose of a Matrix (Changing Rows & Columns)

Let  $A$  be any matrix of order  $m \times n$ . Then  $A^T$  or  $A' = [a_{ij}]$  for  $1 \leq i \leq n$  &  $1 \leq j \leq m$  of order  $n \times m$ .

## Properties of Transpose

If  $A^T$  &  $B^T$  denote the transpose of  $A$  and  $B$

- $(A + B)^T = A^T + B^T$ ; note that  $A$  &  $B$  have the same order.
- $(AB)^T = B^T A^T$  (Reversal law)  $A$  &  $B$  are conformable for matrix product  $AB$
- $(A^T)^T = A$
- $(kA)^T = kA^T$ , where  $k$  is a scalar.

**General:**  $(A_1 \cdot A_2 \cdot \dots \cdot A_n)^T = A_n^T \cdot \dots \cdot A_2^T \cdot A_1^T$  (reversal law for transpose)

## Symmetric & Skew Symmetric Matrix

- Symmetric matrix:** For symmetric matrix  $A = A^T$ .

**Note:** Maximum number of distinct entries in any symmetric matrix of order  $n$  is  $\frac{n(n+1)}{2}$ .

- Skew symmetric matrix:** Square matrix  $A = [a_{ij}]$  is said to be skew symmetric if  $a_{ij} = -a_{ji} \forall i \& j$ . Hence if  $A$  is skew symmetric, hence  $a_{ii} = -a_{ii} \Rightarrow a_{ii} = 0 \forall i$ .

Thus the diagonal elements of a skew square matrix are all zero, but not the converse.

For a skew symmetric matrix  $A = -A^T$ .

- Properties of symmetric & skew symmetric matrix:
  - Let  $A$  be any square matrix then,  $A + A^T$  is a symmetric matrix and  $A - A^T$  is a skew symmetric matrix.
  - The sum of two symmetric matrix is a symmetric matrix and the sum of two skew symmetric matrix is a skew symmetric matrix.
  - If  $A$  &  $B$  are symmetric matrices then,
    - $AB + BA$  is a symmetric matrix.
    - $AB - BA$  is a skew symmetric matrix.
- Every square matrix can be uniquely expressed as a sum or difference of a symmetric and a skew symmetric matrix.

$$A = \underbrace{\frac{1}{2}(A + A^T)}_{\text{symmetric}} + \underbrace{\frac{1}{2}(A - A^T)}_{\text{skew symmetric}}$$

and 
$$A = \frac{1}{2}(A^T + A) - \frac{1}{2}(A^T - A)$$

## Adjoint of a Square Matrix

Let  $A = [a_{ij}] = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}$  be a square matrix and let the

matrix formed by the cofactors of  $[a_{ij}]$  in determinant  $|A|$  is

$$\begin{pmatrix} C_{11} & C_{12} & C_{13} \\ C_{21} & C_{22} & C_{23} \\ C_{31} & C_{32} & C_{33} \end{pmatrix}. \text{ Then } (\text{adj } A) = \begin{pmatrix} C_{11} & C_{21} & C_{31} \\ C_{12} & C_{22} & C_{32} \\ C_{13} & C_{23} & C_{33} \end{pmatrix}.$$

### Key Note

If  $A$  be a square matrix of order  $n$ , then

- $A(\text{adj } A) = |A| I_n = (\text{adj } A) \cdot A$
- $|\text{adj } A| = |A|^{n-1}$
- $\text{adj}(\text{adj } A) = |A|^{n-2} A$
- $|\text{adj}(\text{adj } A)| = |A|^{(n-1)^2}$
- $\text{adj}(AB) = (\text{adj } B)(\text{adj } A)$
- $\text{adj}(KA) = K^{n-1}(\text{adj } A)$ , where  $K$  is a scalar

## Inverse of a Matrix (Reciprocal Matrix)

A square matrix  $A$  (non singular) said to be invertible, if there exists a matrix  $B$  such that,  $AB = I = BA$ .

$B$  is called the inverse (reciprocal) of  $A$  and is denoted by  $A^{-1}$ . Thus

$$A^{-1} = B \Leftrightarrow AB = I = BA$$

We have,  $A \cdot (\text{adj } A) = |A| I_n$   
 $A^{-1} \cdot A(\text{adj } A) = A^{-1} I_n |A|$   
 $I_n (\text{adj } A) = A^{-1} |A| I_n$

$$\therefore A^{-1} = \frac{(\text{adj } A)}{|A|}$$

**Note:** The necessary and sufficient condition for a square matrix  $A$  to be invertible is that  $|A| \neq 0$ .

**Theorem:** If  $A$  and  $B$  are invertible matrices of the same order, then  $(AB)^{-1} = B^{-1}A^{-1}$ .

### Key Note

- If  $A$  be an invertible matrix, then  $A^T$  is also invertible and  $(A^T)^{-1} = (A^{-1})^T$ .
- If  $A$  is invertible,
  - $(A^{-1})^{-1} = A$
  - $(A^k)^{-1} = (A^{-1})^k = A^{-k}$ ;  $k \in N$

## System of Equation and Criteria for Consistency

### Gauss - Jordan Method

**Example:**

$$a_1x + b_1y + c_1z = d_1$$

$$a_2x + b_2y + c_2z = d_2$$

$$a_3x + b_3y + c_3z = d_3$$

$$\Rightarrow \begin{bmatrix} a_1x + b_1y + c_1z \\ a_2x + b_2y + c_2z \\ a_3x + b_3y + c_3z \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} \Rightarrow \begin{bmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix}$$

$$\Rightarrow AX = B \Rightarrow A^{-1}AX = A^{-1}B$$

$$\Rightarrow X = A^{-1}B = \frac{\text{Adj } A}{|A|} \cdot B$$

### Key Note

- If  $|A| \neq 0$ , system is consistent having unique solution.
- If  $|A| = 0$  and  $(\text{adj } A) \cdot B \neq O$  (Null matrix), system is inconsistent having unique non-trivial solution.
- If  $|A| \neq 0$  and  $(\text{adj } A) \cdot B = O$  (Null matrix), system is consistent having trivial solution.
- If  $|A| = 0$ , then

