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# **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**

- O The pair of elements  $a_{ij}$  and  $a_{ji}$  are called Conjugate Elements.
- O The elements  $a_{11}$ ,  $a_{22}$ ,  $a_{33}$ , ...  $a_{mm}$  are called Diagonal Elements. the line along which the diagonal elements lie is called "Principal or Leading diagonal." The quantity  $\Sigma a_{ii}$  = 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,

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

- 1. Both have the same order.
- **2.**  $a_{ii} = b_{ii}$  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**

If 
$$A = \begin{bmatrix} a & b & c \\ b & c & a \\ c & a & b \end{bmatrix}$$
, 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  $p \times q$  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**

- **1.** If A and B are two matrices such that
  - (i) AB = BA then A and B are said to commute
  - (ii) AB = -BA then A and B are said to anticommute
- **2. Matrix Multiplication is Associative:** If A, B & C are conformable for the product AB & BC, then (AB)C = A(BC).
- 3. Distributivity: A(B+C) = AB + AC (A+B)C = AC + BC, provided A, B and C are conformable for respective products.

### **Positive Integral Powers of a Square Matrix**

- **1.**  $A^m A^n = A^{m+n}$
- **2.**  $(A^m)^n = A^{mn} = (A^n)^m$
- **3.**  $I^m = I m, n \in N$

### **Orthogonal Matrix**

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

#### **Key Note**

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

# **Some Square Matrices**

- 1. Idempotent Matrix: A square matrix is idempotent provided  $A^2 = A$ . For idempotent matrix note the following:
  - (a)  $A^n = A \ \forall \ n \geq 2, n \in \mathbb{N}$ .
  - **(b)** determinant value of idempotent matrix is either 0 or 1.
  - **(c)** If idempotent matrix is invertible then its inverse will be identity matrix i.e. *I*.
- **2. 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.
- **3. Nilpotent Matrix:** A square matrix is said to be nilpotent matrix of order  $m, m \in N$ , if  $A^m = O$ ,  $A^{m-1} \neq O$ .
  - Note that a nilpotent matrix will not be invertible.
- **4. Involutary Matrix:** If  $A^2 = I$ , the matrix is said to be an involutary matrix.
  - Note that  $A = A^{-1}$  for an involutary matrix.



**5.** If A and B are square matrices of same order and AB = BA then

$$(A+B)^n = {^nC_0}A^n + {^nC_1}A^{n-1}B + {^nC_2}A^{n-2}B^2 + \dots + {^nC_n}B^n.$$

### 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  $1 \le i \le n \& 1 \le j \le m$  of order  $n \times m$ .

### **Properties of Transpose**

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

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

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

### **Symmetric & Skew Symmetric Matrix**

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

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

**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, then  $a_{ii} = -a_{ii} \implies a_{ii} = 0 \forall i$ .

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

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

- **3.** Properties of symmetric & skew symmetric matrix:
  - (a) Let A be any square matrix then,  $A + A^T$  is a symmetric matrix and  $A A^T$  is a skew symmetric matrix.
  - **(b)** The sum of two symmetric matrix is a symmetric matrix and the sum of two skew symmetric matrix is a skew symmetric matrix.
  - (c) If A & B are symmetric matrices then,
    - (i) AB + BA is a symmetric matrix.
    - (ii) AB BA is a skew symmetric matrix.
- **4.** 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

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

#### **Key Note**

If A be a square matrix of order n, then

**1.** 
$$A(\text{adj } A) = |A| I_n = (\text{adj } A) . A$$

**2.** 
$$| \operatorname{adj} A | = |A|^{n-1}$$

**3.** adj(adj 
$$A$$
) =  $|A|^{n-2}A$ 

**4.** 
$$|\operatorname{adj}(\operatorname{adj} A)| = |A|^{(n-1)^2}$$

**5.** 
$$adj(AB) = (adj B) (adj A)$$

**6.** adj 
$$(KA) = K^{n-1}$$
 (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

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

We have,  $A \cdot (\operatorname{adj} A) = |A| I_n$ 

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

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

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

**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**

- O If A be an invertible matrix, then  $A^T$  is also invertible and  $(A^T)^{-1} = (A^{-1})^T$ .
- $\circ$  If A is invertible,

(a) 
$$(A^{-1})^{-1} = A$$

(b) 
$$(A^k)^{-1} = (A^{-1})^k = A^{-k} : k \in \mathbb{N}$$

# **System of Equation and Criteria for Consistency**

Gauss - Jordan Method

Example:

$$a_1 x + b_1 y + c_1 z = d_1$$

$$a_2 x + b_2 y + c_2 z = d_2$$

$$a_3x + b_3y + c_3z = d_3$$

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

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

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

#### **Key Note**

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

#### **Matrix Method Fails**

If 
$$(adj A) \cdot B = O$$
 (null matrix)

Consistent
(infinite solutions)

If  $(adj A) \cdot B \neq O$ 

Inconsistent
(no solutions)