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Mathematics

# Matrices



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# Table of Content

- Definition. 1.
- Order of a matrix. 2.
- Equality of matrices. 3.
- Types of matrices. 4.
- 5. Addition and subtraction of matrices.
- Scalar multiplication of matrices. 6.
- Multiplication of matrices. 7.
- Positive integral powers of a matrix. 8.
- Matrix polynomial. 9.
- Transpose of a matrix. 10.
- determinants of a matrix. 11.
- 12. Special types of matrices.
- 13. Adjoint of a square matrix.
- Inverse of a matrix. 14.
- 15. Elementary transformation or Elementary operations of a matrix.
- 16. Elementary matrix.
- Rank of matrix. 17.
- 18. Echelon form of a matrix.













- 19. System of simultaneous linear equations.
- 20. Solution of a non-homogeneous system of linear equations.
- 21. Cayley-Hamilton theorem.
- 22. Geometrical transformations.
- 23. Matrices of rotations of axes.













# 1. Definition.

A rectangular arrangement of numbers (which may be real or complex numbers) in rows and columns, is called a matrix. This arrangement is enclosed by small () or big [] brackets. The numbers are called the elements of the matrix or entries in the matrix. A matrix is represented by capital letters *A*, *B*, *C* etc. and its elements by small letters *a*, *b*, *c*, *x*, *y* etc. The following are some examples of matrices:

$$A = \begin{bmatrix} 1 & 4 \\ 2 & 3 \end{bmatrix}, B = \begin{bmatrix} 2+i & -3 & 2 \\ 1 & -3+i & -5 \end{bmatrix}, C = [1, 4, 9], D = \begin{bmatrix} a \\ g \\ h \end{bmatrix}, E = [l]$$

## 2. Order of a Matrix.

A matrix having *m* rows and *n* columns is called a matrix of order  $m \times n$  or simply  $m \times n$  matrix (read as 'an *m* by *n* matrix). A matrix *A* of order  $m \times n$  is usually written in the following manner

	$a_{11}$	$a_{12}$	$a_{13}$	<i>a</i> <sub>1j</sub>	$a_{1n}$	
<i>A</i> =	<i>a</i> <sub>21</sub>	$a_{22}$	<i>a</i> <sub>23</sub>	$a_{2j}$	a <sub>2n</sub>	
				•••••		or $A = [a, 1]$ where $i = 1, 2, \dots, m$
	$a_{i1}$	$a_{i2}$	$a_{i3}$	a <sub>ij</sub>	a <sub>in</sub>	j = 1, 2,,n
				•••••		
	$a_{m1}$	$a_{m2}$	$a_{m3}$	a <sub>mj</sub>	a <sub>mn</sub>	

Here  $a_{ij}$  denotes the element of i<sup>th</sup> row and j<sup>th</sup> column. *Example:* order of matrix  $\begin{bmatrix} 3 & -1 & 5 \\ 6 & 2 & -7 \end{bmatrix}$  is 2×3

Note: A matrix of order  $m \times n$  contains mn elements. Every row of such a matrix contains n elements and every column contains m elements.





## 3. Equality of Matrices.

Two matrix *A* and *B* are said to be equal matrix if they are of same order and their corresponding elements are equal *Example*. If  $A = \begin{bmatrix} 1 & 6 & 3 \\ 5 & 2 & 1 \end{bmatrix}$  and  $B = \begin{bmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{bmatrix}$  are equal matrices.

Then  $a_1 = 1, a_2 = 6, a_3 = 3, b_1 = 5, b_2 = 2, b_3 = 1$ 

## 4. Types of Matrices.

(1) **Row matrix**: A matrix is said to be a row matrix or row vector if it has only one row and any number of columns. *Example*: [5 0 3] is a row matrix of order 1× 3 and [2] is a row matrix of order 1×1.

(2) **Column matrix:** A matrix is said to be a column matrix or column vector if it has only one column and any number of rows. *Example*:  $\begin{bmatrix} 2\\3\\-6 \end{bmatrix}$  is a column matrix of order 3×1 and [2] is a column matrix of

order 1×1. Observe that [2] is both a row matrix as well as a column matrix.

(3) **Singleton matrix**: If in a matrix there is only one element then it is called singleton matrix. Thus,  $A = [aij]_{m \times n}$  is a singleton matrix if m = n = 1 *Example*: [2], [3], [a], [-3] are singleton matrices.

(4) **Null or zero matrix:** If in a matrix all the elements are zero then it is called a zero matrix and it is generally denoted by *O*. Thus  $A = [a_{ij}]_{m \times n}$  is a zero matrix if  $a_{ij} = 0$  for all *i* and *j*.

*Example*:  $[0], \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0$ 

(5) **Square matrix**: If number of rows and number of columns in a matrix are equal, then it is called a square matrix. Thus  $A = [a_{ij}]_{m \times n}$  is a square matrix if m = n. *Example*:  $\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$  is a square matrix of order

3×3













(i) If  $m \neq n$  then matrix is called a rectangular matrix.

(ii) The elements of a square matrix A for which i = j, *i.e.*  $a_{11}, a_{22}, a_{33}, \dots, a_{nn}$  are called diagonal elements and the line joining these elements is called the principal diagonal or leading diagonal of matrix *A*.

(iii) **Trace of a matrix:** The sum of diagonal elements of a square matrix. *A* is called the trace of matrix A, which is denoted by tr A.  $tr A = \sum_{i=1}^{n} a_{ii} = a_{11} + a_{22} + \dots + a_{nn}$ 

**Properties of trace of a matrix:** Let  $A = [a_{ii}]_{n \times n}$  and  $B = [b_{ij}]_{n \times n}$  and  $\lambda$  be a scalar

(i)  $tr(\lambda A) = \lambda tr(A)$ (ii) tr(A - B) = tr(A) - tr(B)(iii) tr(AB) = tr(BA)(iv) tr(A) = tr(A') or  $tr(A^T)$ (v)  $tr(I_n) = n$ (vi) tr(0) = 0(vii)  $tr(AB) \neq tr A . tr B$ 

(6) **Diagonal matrix:** If all elements except the principal diagonal in a square matrix are zero, it is called a diagonal matrix. Thus a square matrix  $A = [a_{ij}]$  is a diagonal matrix if  $a_{ij} = 0$ , when  $i \neq j$ .

*Example*:  $\begin{bmatrix} 2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 4 \end{bmatrix}$  is a diagonal matrix of order 3×3, which can be denoted by diag [2, 3, and 4]

Note: No element of principal diagonal in a diagonal matrix is zero.

**D** Number of zeros in a diagonal matrix is given by  $n^2 - n$  where *n* is the order of the matrix.

**A** diagonal matrix of order  $n \times n$  having  $d_1, d_2, \dots, d_n$  as diagonal elements is denoted by diag  $[d_1, d_2, \dots, d_n]$ .





(7) **Identity matrix**: A square matrix in which elements in the main diagonal are all '1' and rest are all zero is called an identity matrix or unit matrix. Thus, the square matrix  $A = [a_{ii}]$  is an identity matrix, if

$$a_{ij} = \begin{cases} 1, \text{ if } i = j \\ 0, \text{ if } i \neq j \end{cases}$$

We denote the identity matrix of order n by  $I_n$ .

*Example*: [1],  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ ,  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$  are identity matrices of order 1, 2 and 3 respectively.

(8) **Scalar matrix** : A square matrix whose all non-diagonal elements are zero and diagonal elements are equal is called a scalar matrix. Thus, if  $A = [a_{ij}]$  is a square matrix and  $a_{ij} = \begin{cases} \alpha, \text{if } i = j \\ 0, \text{ if } i \neq j \end{cases}$ , then A is a scalar matrix.

*Example*. [2],  $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ ,  $\begin{bmatrix} 5 & 0 & 0 \\ 0 & 5 & 0 \\ 0 & 0 & 5 \end{bmatrix}$  are scalar matrices of order 1, 2 and 3 respectively.

Note: Unit matrix and null square matrices are also scalar matrices.

(9) **Triangular Matrix**: A square matrix  $[a_{ij}]$  is said to be triangular matrix if each element above or below the principal diagonal is zero. It is of two types

(i) **Upper Triangular matrix:** A square matrix  $[a_{ij}]$  is called the upper triangular matrix, if  $a_{ij} = 0$  when i > j.

Example:  $\begin{bmatrix} 3 & 1 & 2 \\ 0 & 4 & 3 \\ 0 & 0 & 6 \end{bmatrix}$  is an upper triangular matrix of order 3×3.

(ii) **Lower Triangular matrix:** A square matrix  $[a_{ij}]$  is called the lower triangular matrix, if  $a_{ij} = 0$  when *i* < *j*.

Example:  $\begin{bmatrix} 1 & 0 & 0 \\ 2 & 3 & 0 \\ 4 & 5 & 2 \end{bmatrix}$  is a lower triangular matrix of order 3×3.













Note: Minimum number of zeros in a triangular matrix is given by  $\frac{n(n-1)}{2}$  where *n* is order of matrix.

Diagonal matrix is both upper and lower triangular.

□ A triangular matrix  $a = [a_{ij}]_{n \times n}$  is called strictly triangular if  $a_{ij} = 0$  for  $1 \le i \le n$ 

## 5. Addition and Subtraction of Matrices.

If  $A = [a_{ij}]_{m \times n}$  and  $B = [b_{ij}]_{m \times n}$  are two matrices of the same order then their sum A + B is a matrix whose each element is the sum of corresponding elements. *i.e.*  $A + B = [a_{ii} + b_{ij}]_{m \times n}$ 

Example: If  $A = \begin{bmatrix} 5 & 2 \\ 1 & 3 \\ 4 & 1 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 5 \\ 2 & 2 \\ 3 & 3 \end{bmatrix}$ , then  $A + B = \begin{bmatrix} 5+1 & 2+5 \\ 1+2 & 3+2 \\ 4+3 & 1+3 \end{bmatrix} = \begin{bmatrix} 6 & 7 \\ 3 & 5 \\ 7 & 4 \end{bmatrix}$ 

Similarly, their subtraction A - B is defined as  $A - B = [a_{ij} - b_{ij}]_{m \times n}$ 

*i.e.* in above example  $A - B = \begin{bmatrix} 5 - 1 & 2 - 5 \\ 1 - 2 & 3 - 2 \\ 4 - 3 & 1 - 3 \end{bmatrix} = \begin{bmatrix} 4 & -3 \\ -1 & 1 \\ 1 & -2 \end{bmatrix}$ 

Note: Matrix addition and subtraction can be possible only when matrices are of the same order.

#### Properties of matrix addition: If A, B and C are matrices of same order, then

- (i) A + B = B + A (Commutative law)
- (ii) (A + B) + C = A + (B + C) (Associative law)
- (iii) A + O = O + A = A, Where O is zero matrix which is additive identity of the matrix.
- (iv) A + (-A) = 0 = (-A) + A, where (-A) is obtained by changing the sign of every element of A, which is additive inverse of the matrix.

(v) 
$$\begin{array}{c} A+B=A+C\\ B+A=C+A \end{array}$$
  $\Rightarrow$   $B=C$  (Cancellation law)













## 6. Scalar Multiplication of Matrices.

Let  $A = [a_{ij}]_{m \times n}$  be a matrix and k be a number, then the matrix which is obtained by multiplying every element of A by k is called scalar multiplication of A by k and it is denoted by kA.

Thus, if  $A = [a_{ij}]_{m \times n}$ , then  $kA = Ak = [ka_{ij}]_{m \times n}$ . *Example*. If  $A = \begin{bmatrix} 2 & 4 \\ 3 & 1 \\ 4 & 6 \end{bmatrix}$ , then  $5A = \begin{bmatrix} 10 & 20 \\ 15 & 5 \\ 20 & 30 \end{bmatrix}$ 

#### Properties of scalar multiplication:

If A, B are matrices of the same order and  $\lambda$ ,  $\mu$  are any two scalars then

(i)  $\lambda(A + B) = \lambda A + \lambda B$ (ii)  $(\lambda + \mu)A = \lambda A + \mu A$ (iii)  $\lambda(\mu A) = (\lambda \mu A) = \mu(\lambda A)$ (iv)  $(-\lambda A) = -(\lambda A) = \lambda(-A)$ 

Note: All the laws of ordinary algebra hold for the addition or subtraction of matrices and their multiplication by scalars.

### 7. Multiplication of Matrices.

Two matrices A and B are conformable for the product AB if the number of columns in A (pre-multiplier) is same as the number of rows in *B* (post multiplier). Thus, if  $A = [a_{ij}]_{m \times n}$  and  $B = [b_{ij}]_{n \times p}$  are two matrices of order  $m \times n$  and  $n \times p$  respectively, then their product *AB* is of order  $m \times p$  and is defined as

$$(AB)_{ij} = \sum_{r=1}^{n} a_{ir} b_{rj}$$
  
=  $[a_{i1}a_{i2}...a_{in}] \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix}$  =  $(t^{\text{h}} \text{ row of A})(t^{\text{h}} \text{ column of } B) \dots (i)$ , where  $i=1, 2, ..., m$  and  $j=1, 2, ..., p$ 

Now we define the product of a row matrix and a column matrix.

Let  $A = [a_1 a_2 \dots a_n]$  be a row matrix and  $B = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$  be a column matrix.

Then  $AB = [a_1b_1 + a_2b_2 + ... + a_nb_n]$  ...(ii). Thus, from (i),

 $(AB)_{ij}$  = Sum of the product of elements of  $t^{h}$  row of A with the corresponding elements of  $j^{th}$  column of B.





#### **Properties of matrix multiplication**

If A, B and C are three matrices such that their product is defined, then

(i)  $AB \neq BA$ (Generally not commutative)(ii) (AB)C = A(BC)(Associative Law)(iii) IA = A = AIWhere I is identity matrix for matrix multiplication(iv) A(B+C) = AB + AC(Distributive law)(v) If  $AB = AC \Rightarrow B = C$ (Cancellation law is not applicable)(vi) If AB = 0It does not mean that A = 0 or B = 0, again product of two non zero matrix

(vi) If AB = 0 It does not mean that A = 0 or B = 0, again product of two non zero matrix may be a zero matrix.

Note: If *A* and *B* are two matrices such that *AB* exists, then *BA* may or may not exist. The multiplication of two triangular matrices is a triangular matrix. The multiplication of two diagonal matrices is also a diagonal matrix and diag  $(a_1, a_2, ..., a_n) \times \text{diag} (b_1, b_2, ..., b_n) = \text{diag} (a_1b_1, a_2b_2, ..., a_nb_n)$ The multiplication of two scalar matrices is also a scalar matrix. If *A* and *B* are two matrices of the same order, then

(i)  $(A + B)^2 = A^2 + B^2 + AB + BA$ (ii)  $(A - B^2) = A^2 + B^2 - AB - BA$ (iii)  $(A - B)(A + B) = A^2 - B^2 + AB - BA$ (iv)  $(A + B)(A - B) = A^2 - B^2 - AB + BA$ (v) A(-B) = (-A)B = -(AB)

## 8. Positive Integral Powers of a Matrix.

The positive integral powers of a matrix *A* are defined only when *A* is a square matrix. Also then  $A^2 = A.A$ ,  $A^3 = A.A.A = A^2A$ . Also for any positive integers *m*,*n*. (i)  $A^m A^n = A^{m+n}$ (ii)  $(A^m)^n = A^{nm} = (A^n)^m$ (iii)  $I^n = I, I^m = I$ 

(iv)  $A^0 = I_n$  Where A is a square matrix of order n.





# 9. Matrix Polynomial.

Let  $f(x) = a_0 x^n + a_1 x^{n-1} + a_2 x^{n-2} + \dots + a_{n-1} x + a_n$  be a polynomial and let A be a square matrix of order n. Then  $f(A) = a_0 A^n + a_1 A^{n-1} + a_2 A^{n-2} + \dots + a_{n-1} A + a_n I_n$  is called a matrix polynomial.

*Example*: If  $f(x) = x^2 - 3x + 2$  is a polynomial and A is a square matrix, then  $A^2 - 3A + 2I$  is a matrix polynomial.

## 10. Transpose of a Matrix.

The matrix obtained from a given matrix A by changing its rows into columns or columns into rows is called transpose of Matrix A and is denoted by  $A^T$  or A'.

From the definition it is obvious that if order of A is  $m \times n$ , then order of  $A^T$  is  $n \times m$ 

Example: Transpose of matrix  $\begin{bmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{bmatrix}_{2\times 3}$  is  $\begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{bmatrix}_{3\times 2}$ 

Properties of transpose: Let A and B be two matrices then

- (i)  $(A^T)^T = A$
- (ii)  $(A + B)^T = A^T + B^T$ , A and B being of the same order
- (iii)  $(kA)^T = kA^T$ , k be any scalar (real or complex)
- (iv)  $(AB)^T = B^T A^T$ , A and B being conformable for the product AB
- (v)  $(A_1 A_2 A_3 \dots A_{n-1} A_n)^T = A_n^T A_{n-1}^T \dots A_3^T A_2^T A_1^T$ (vi)  $I^T = I$













# 11. Determinant of a Matrix.

If  $A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$  be a square matrix, then its determinant, denoted by |A| or Det (A) is defined as  $|A| = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$ 

#### **Properties of determinant of a matrix**

- (i) |A| Exists  $\Leftrightarrow A$  is square matrix
- (ii)  $|AB| \neq A ||B|$
- (iii)  $|A^T| = |A|$
- (iv)  $|kA| = k^n |A|$ , If A is a square matrix of order n
- (v) If A and B are square matrices of same order then |AB| = |BA|
- (vi) If A is a skew symmetric matrix of odd order then |A| = 0
- (vii) If  $A = \text{diag}(a_1, a_2, \dots, a_n)$  then  $|A| = a_1 a_2 \dots a_n$
- (viii)  $|A|^n \neq A^n |, n \in N.$

## 12. Special Types of Matrices.

#### (1) Symmetric and skew-symmetric matrix

(i) **Symmetric matrix**: A square matrix  $A = [a_{ij}]$  is called symmetric matrix if  $a_{ij} = a_{ji}$  for all *i*, *j* or  $A^T = A$ 

Example:  $\begin{bmatrix} a & h & g \\ h & b & f \\ g & f & c \end{bmatrix}$ 

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Note: Every unit matrix and square zero matrix are symmetric matrices.

□ Maximum number of different elements in a symmetric matrix is  $\frac{n(n+1)}{2}$ 

(ii) **Skew-symmetric matrix**: A square matrix  $A = [a_{ij}]$  is called skew- symmetric matrix if  $a_{ij} = -a_{ji}$  for all *i*, *j* 

or  $A^{T} = -A$ . Example:  $\begin{bmatrix} 0 & h & g \\ -h & 0 & f \\ -g & -f & 0 \end{bmatrix}$ 

Note: All principal diagonal elements of a skew- symmetric matrix are always zero because for any diagonal element.  $a_{ij} = -a_{ij} \Rightarrow a_{ij} = 0$ 

□ Trace of a skew symmetric matrix is always 0.

#### Properties of symmetric and skew-symmetric matrices:

(i) If A is a square matrix, then  $A + A^T$ ,  $AA^T$ ,  $A^TA$  are symmetric matrices, while  $A - A^T$  is skew- symmetric matrix.

(ii) If A is a symmetric matrix, then  $-A, KA, A^T, A^n, A^{-1}, B^TAB$  are also symmetric matrices, where  $n \in N$ ,  $K \in R$  and B is a square matrix of order that of A

- (iii) If A is a skew-symmetric matrix, then
- (a)  $A^{2n}$  is a symmetric matrix for  $n \in N$ ,
- (b)  $A^{2n+1}$  is a skew-symmetric matrix for  $n \in N$ ,
- (c) kA is also skew-symmetric matrix, where  $k \in R$ ,
- (d)  $B^T A B$  is also skew- symmetric matrix where B is a square matrix of order that of A.
- (iv) If A, B are two symmetric matrices, then
- (a)  $A \pm B$ , AB + BA are also symmetric matrices,
- (b) AB BA is a skew- symmetric matrix,
- (c) AB is a symmetric matrix, when AB = BA.
- (v) If *A*,*B* are two skew-symmetric matrices, then
- (a)  $A \pm B$ , AB BA are skew-symmetric matrices,
- (b) AB + BA is a symmetric matrix.













(vi) If A a skew-symmetric matrix and C is a column matrix, then  $C^T A C$  is a zero matrix.

(vii) Every square matrix *A* can uniquelly be expressed as sum of a symmetric and skew-symmetric matrix *i.e.* 

 $A = \left[\frac{1}{2}(A + A^{T})\right] + \left[\frac{1}{2}(A - A^{T})\right].$ 

(2) **Singular and Non-singular matrix:** Any square matrix *A* is said to be non-singular if  $|A| \neq 0$ , and a square matrix *A* is said to be singular if |A| = 0. Here |A| (or det(*A*) or simply det |A| means corresponding determinant of square matrix *A*.

*Example*:  $A = \begin{bmatrix} 2 & 3 \\ 4 & 5 \end{bmatrix}$  then  $|A| = \begin{vmatrix} 2 & 3 \\ 4 & 5 \end{vmatrix} = 10 - 12 = -2 \Rightarrow A$  is a non-singular matrix.

(3) Hermitian and skew-Hermitian matrix: A square matrix  $A = [a_{ij}]$  is said to be hermitian matrix if

$$a_{ij} = \overline{a}_{ji} \quad \forall i.j \ i.e. \ A = A^{\theta}. \ \text{Example.} \begin{bmatrix} a & b+ic \\ b-ic & d \end{bmatrix}, \begin{bmatrix} 3 & 3-4i & 5+2i \\ 3+4i & 5 & -2+i \\ 5-2i & -2-i & 2 \end{bmatrix} \text{ are Hermitian matrices.}$$

Note: If *A* is a Hermitian matrix then  $a_{ii} = \overline{a}_{ii} \Rightarrow a_{ii}$  is real  $\forall i$ , thus every diagonal element of a Hermitian matrix must be real.

□ A Hermitian matrix over the set of real numbers is actually a real symmetric matrix and a square matrix,  $A = |a_{ij}|$  is said to be a skew-Hermitian if  $a_{ij} = -\overline{a}_{ji}$ .  $\forall i, ji.e. A^{\theta} = -A$ .

Example: 
$$\begin{bmatrix} 0 & -2+i \\ 2-i & 0 \end{bmatrix}, \begin{bmatrix} 3i & -3+2i & -1-i \\ 3+2i & -2i & -2-4i \\ 1-i & 2-4i & 0 \end{bmatrix}$$
 are skew-Hermitian matrices.

□ If *A* is a skew-Hermitian matrix, then  $a_{ii} = -\overline{a}_{ii} \Rightarrow a_{ii} + \overline{a}_{ii} = 0$  i.e.  $a_{ii}$  must be purely imaginary or zero. □ A skew-Hermitian matrix over the set of real numbers is actually a real skew-symmetric matrix.

(4) **Orthogonal matrix:** A square matrix *A* is called orthogonal if  $AA^{T} = I = A^{T}A$  *i.e.* if  $A^{-1} = A^{T}$  *Example:*  $A = \begin{bmatrix} \cos \alpha & -\sin \alpha \\ -\sin \alpha & \cos \alpha \end{bmatrix}$  is orthogonal because  $A^{-1} = \begin{bmatrix} \cos \alpha & \sin \alpha \\ -\sin \alpha & \cos \alpha \end{bmatrix} = A^{T}$ In fact every unit matrix is orthogonal.

(5) **Idempotent matrix**: A square matrix *A* is called an idempotent matrix if  $A^2 = A$ . *Example*:  $\begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}$  is an idempotent matrix, because  $A^2 = \begin{bmatrix} 1/4 + 1/4 & 1/4 + 1/4 \\ 1/4 + 1/4 & 1/4 + 1/4 \end{bmatrix} = \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix} = A$ .





Also, 
$$A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$
 and  $B = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$  are idempotent matrices because  $A^2 = A$  and  $B^2 = B$ .

In fact every unit matrix is indempotent.

(6) **Involutory matrix:** A square matrix A is called an involutory matrix if  $A^2 = I$  or  $A^{-1} = A$ *Example*:  $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$  is an involutory matrix because  $A^2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$ 

In fact every unit matrix is involutory.

(7) **Nilpotent matrix:** A square matrix A is called a nilpotent matrix if there exists a  $p \in N$  such that  $A^{p} = 0$ 

*Example*:  $A = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$  is a nilpotent matrix because  $A^2 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = 0$  (Here P = 2)

(8) **Unitary matrix:** A square matrix is said to be unitary, if  $\overline{A} \cdot A = I$  since  $|\overline{A}'| = |A|$  and  $|\overline{A} \cdot A| = |\overline{A}'|| |A|$  therefore if  $\overline{A}' A = I$ , we have  $|\overline{A}'|| |A| = 1$ 

Thus the determinant of unitary matrix is of unit modulus. For a matrix to be unitary it must be nonsingular.

Hence  $\overline{A}' A = I \Longrightarrow A \overline{A}' = I$ 

(9) **Periodic matrix**: A matrix *A* will be called a periodic matrix if  $A^{k+1} = A$  where *k* is a positive integer. If, however *k* is the least positive integer for which  $A^{k+1} = A$ , then *k* is said to be the period of *A*.

(10) **Differentiation of a matrix:** If 
$$A = \begin{bmatrix} f(x) & g(x) \\ h(x) & l(x) \end{bmatrix}$$
 then  $\frac{dA}{dx} = \begin{bmatrix} f'(x) & g'(x) \\ h'(x) & l'(x) \end{bmatrix}$  is a differentiation of matrix

А.

Example: If  $A = \begin{bmatrix} x^2 & \sin x \\ 2x & 2 \end{bmatrix}$  then  $\frac{dA}{dx} = \begin{bmatrix} 2x & \cos x \\ 2 & 0 \end{bmatrix}$ 

(11) Submatrix : Let A be m×n matrix, then a matrix obtained by leaving some rows or columns or both,

of *A* is called a sub matrix of *A*. *Example*: If  $A' = \begin{bmatrix} 2 & 1 & 0 \\ 3 & 2 & 2 \\ 2 & 5 & 3 \end{bmatrix}$  and  $\begin{bmatrix} 2 & 2 \\ 5 & 3 \end{bmatrix}$  are sub matrices of matrix

	2	1	0	-1	
<i>A</i> =	3	2	2	4	
	2	5	3	1	

(12) **Conjugate of a matrix:** The matrix obtained from any given matrix A containing complex number as its elements, on replacing its elements by the corresponding conjugate complex numbers is called





conjugate of A and is denoted by  $\overline{A}$ . *Example*:  $A = \begin{bmatrix} 1+2i & 2-3i & 3+4i \\ 4-5i & 5+6i & 6-7i \\ 8 & 7+8i & 7 \end{bmatrix}$  then

 $\overline{A} = \begin{bmatrix} 1 - 2i & 2 + 3i & 3 - 4i \\ 4 + 5i & 5 - 6i & 6 + 7i \\ 8 & 7 - 8i & 7 \end{bmatrix}$ 

### **Properties of conjugates:**

- (i)  $\left(\overline{A}\right) = A$
- (ii)  $\overline{(A+B)} = \overline{A} + \overline{B}$
- (iii)  $\overline{(\alpha A)} = \overline{\alpha}\overline{A}, \alpha$  being any number
- (iv)  $(\overline{AB}) = \overline{A} \overline{B}$ , A and B being conformable for multiplication.

(13) **Transpose conjugate of a matrix:** The transpose of the conjugate of a matrix A is called transposed conjugate of A and is denoted by  $A^{\theta}$ . The conjugate of the transpose of A is the same as the transpose of the conjugate of A *i.e.*  $\overline{(A')} = (\overline{A})' = A^{\theta}$ .

If  $A = [a_{ij}]_{m \times n}$  then  $A^{\theta} = [b_{ji}]_{n \times m}$  where  $b_{ji} = \overline{a}_{ij}$  *i.e.* the  $(j, i)^{th}$  element of  $A^{\theta}$  = the conjugate of  $(i, j)^{th}$  element of A.

Example: If 
$$A = \begin{bmatrix} 1+2i & 2-3i & 3+4i \\ 4-5i & 5+6i & 6-7i \\ 8 & 7+8i & 7 \end{bmatrix}$$
, then  $A^{\theta} = \begin{bmatrix} 1-2i & 4+5i & 8 \\ 2+3i & 5-6i & 7-8i \\ 3-4i & 6+7i & 7 \end{bmatrix}$ 

#### Properties of transpose conjugate

- (i)  $(A^{\theta})^{\theta} = A$
- (ii)  $(A+B)^{\theta} = A^{\theta} + B^{\theta}$
- (iii)  $(kA)^{\theta} = \overline{K}A^{\theta}$ , *K* being any number
- (iv)  $(AB)^{\theta} = B^{\theta}A^{\theta}$





# 13. Adjoint of a Square Matrix.

Let  $A = [a_{ij}]$  be a square matrix of order *n* and let  $C_{ij}$  be cofactor of  $a_{ij}$  in A. Then the transpose of the matrix of cofactors of elements of *A* is called the adjoint of *A* and is denoted by *adj* A Thus,  $adj A = [C_{ij}]^T \Rightarrow (adj A)_{ij} = C_{ji} = \text{cofactor of } a_{ji} \text{ in } A.$ 

If 
$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$
, then  $adjA = \begin{bmatrix} C_{11} & C_{12} & C_{13} \\ C_{21} & C_{22} & C_{23} \\ C_{31} & C_{32} & C_{33} \end{bmatrix}^T = \begin{bmatrix} C_{11} & C_{21} & C_{31} \\ C_{12} & C_{22} & C_{32} \\ C_{13} & C_{23} & C_{33} \end{bmatrix}$ ;

Where  $C_{ii}$  denotes the cofactor of  $a_{ij}$  in A.

Example: 
$$A = \begin{bmatrix} p & q \\ r & s \end{bmatrix}, C_{11} = s, C_{12} = -r, C_{21} = -q, C_{22} = p$$
  
$$\therefore adj A = \begin{bmatrix} s & -r \\ -q & p \end{bmatrix}^T = \begin{bmatrix} s & -q \\ -r & p \end{bmatrix}$$

Note: The adjoint of a square matrix of order 2 can be easily obtained by interchanging the diagonal elements and changing signs of off diagonal elements.

