

Basics:

Linear Algebra

①

Trace of a matrix:

The sum of the diagonal elements of a square matrix A is called trace of A .

Properties

- $\text{tr}(A+B) = \text{tr}(A) + \text{tr}(B)$
- $\text{tr}(AB) = \text{tr}(A) \text{tr}(B)$
- $\text{tr}(AB) = \text{tr}(BA)$
- $\text{tr}(kA) = k \text{tr}(A)$ where k is scalar.

Diagonal matrix:

If all the non-diagonal elements in a square matrix are zero then the matrix is called diagonal matrix.

Note: If A & B are diagonal matrices.

$A \pm B$, A^2, B^2 , $A^n \pm B^n$, A^T , $A^T \pm B^T$, $\text{adj}(A)$, $\text{adj}(B)$, A^{-1} , B^{-1} are also diagonal matrices.

Upper triangular matrix:

$$\begin{bmatrix} 2 & -3 & 4 \\ 0 & 4 & 8 \\ 0 & 0 & 9 \end{bmatrix}$$

Lower triangular matrix:

$$\begin{bmatrix} 2 & 0 & 0 \\ 4 & 8 & 0 \\ 4 & 8 & 9 \end{bmatrix}$$

Triangular matrix: which may be either upper or lower triangular.

Diagonal matrix: Both upper & lower triangular.

The diagonal elements for skew symm matrix is equal to zero.

Properties of transpose.

- $(A^T)^T = A$
- $(A \pm B)^T = A^T \pm B^T$
- $(kA)^T = kA^T$; k is scalar.
- $(AB)^T = B^T A^T$
- $(A^2)^T = (A^T)^2$

Symmetric matrix: $A^T = A$

Skew symmetric matrix: $A^T = -A \Rightarrow$

Note: Every square matrix A can be uniquely expressed as a sum of sym & skew sym matrices

$$A = \frac{A + A^T}{2} + \frac{A - A^T}{2}$$

Properties of symmetric matrix.

If A & B are symmetric matrix then

- $A \pm B$, A^2 , B^2 , $AB + BA$, A^k , B^k , KA - symmetric
- AB , BA need not be symmetric
- $A + A^T$
 $A \cdot A^T$, $A^T A$ } - symmetric

Properties of skew symmetric matrices.

If A & B are skew sym matrix then

$A \pm B$, KA , A^3 , A^5 , A^7 - skew

A^2 , A^4 , A^6 } - symmetric

$A^2 \pm B^2$

AB , BA are not skew symmetric.

$AB - BA$
 $A - A^T$, $A^T - A$ } skew symmetric.

Idempotent matrix: $A^2 = A$

Involutory matrix: $A^2 = I_n$

Nilpotent matrix: $A^m = O_n$

Index of Nilpotent

Orthogonal matrix: $AA^T = A^T A = I$

$$A^{-1} = A^T$$

Note:

- If $AB = A$ & $BA = B$ then A & B are idempotent

- If A & B are orthogonal then AB and BA are also orthogonal matrices.

Note:

The number of different $n \times n$ symmetric matrices with each element being 0 or 1 is $2^{\frac{n(n+1)}{2}}$

Note:

Property	Addition	Multiplication
Commutative	✓	✗
Associative	✓	✓
Distributive	-	✓

Additive Identity $\rightarrow 0$

Additive Inverse $\rightarrow -A$

Multiplicative Identity $\rightarrow I$

Multiplicative Inverse $\rightarrow A^{-1}$

The product of two matrices can be the null matrix while neither of them is the null matrix.

$$A \begin{bmatrix} 0 & 2 \\ 0 & 0 \end{bmatrix}^k \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} kA$$

If $AB = 0$ then BA may or may not be 0

Note:

Sum of odd numbers

$$1+3+5+7+\dots+(2n-1) = n^2$$

Sum of even numbers

$$2+4+6+8+\dots+2n = n^2+n$$

Determinants:

- Det $A = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$

- $\begin{vmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \\ a_3 & b_3 & c_3 \end{vmatrix} = \begin{vmatrix} a_1 & b_1 \\ a_2 & b_2 \\ a_3 & b_3 \end{vmatrix} \begin{vmatrix} c_1 \\ c_2 \\ c_3 \end{vmatrix}$

Sign $= (-1)^{i+j}$

$$= a_1(b_2c_3 - b_3c_2) - b_1(c_2a_3 - c_3a_2) + c_1(a_2b_3 - a_3b_2)$$

- Singular matrix: $|A| = 0$

Properties of determinants:

- If A and B are two square matrices of same order then $|AB| = |A||B|$

- $|A^m| = |A|^m$ ($m=2,3,4,\dots$)

- If $|A| \neq 0$ then $|A^{-1}| = \frac{1}{|A|}$

- $|A| = |A^T|$

- If every element of a row (column) of a determinant of A is zero then $|A| = 0$

Ex:- $\begin{vmatrix} 1 & 2 & 3 \\ 0 & 0 & 0 \\ 4 & 5 & 6 \end{vmatrix} = 0$ (or) $\begin{vmatrix} 1 & 0 & 4 \\ 2 & 0 & 5 \\ 3 & 0 & 6 \end{vmatrix} = 0$

- If any two rows (columns) are identical then the value of determinant is zero.

$\begin{vmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 4 & 5 & 6 \end{vmatrix} = 0$ (or) $\begin{vmatrix} 1 & 1 & 4 \\ 2 & 2 & 5 \\ 3 & 3 & 6 \end{vmatrix} = 0$

- If any two rows (columns) of a determinant are interchanged then sign of det changes

- $|kA_{mn}| = k^n |A_{mn}|$

- $\begin{vmatrix} a+\lambda_1 & b+\lambda_2 & c+\lambda_3 \\ d & e & f \\ g & h & i \end{vmatrix} = \begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} + \begin{vmatrix} \lambda_1 & \lambda_2 & \lambda_3 \\ d & e & f \\ g & h & i \end{vmatrix}$

- The det of upper Δ^{lu} , lower Δ^{lu} , diagonal, scalar matrix is equal to the product of its diagonals.

- If A is an orthogonal matrix then $|A| = \pm 1$

- If A is an idempotent matrix then $|A| = 0$ or 1
- $|I_n| = 1$
- The det of skew sym matrix of odd order is zero.

Minor of an element (M_{ij})

If $A = (a_{ij})$ is a square matrix of order n then the minor of an element a_{ij} in A is the det of a square matrix that remains after deleting i th row and j th col of A .

Ex: $A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \rightarrow M_{11} = \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix}; M_{22} = \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix}$

Cofactor of an element:

$$A_{ij} = (-1)^{i+j} M_{ij}$$

Note:

If $A = (a_{ij})$ is a square matrix of order n then

$$\sum_{j=1}^n a_{ij} M_{ij} = |A| \quad (\text{or}) \quad \sum_{i=1}^n a_{ij} M_{ij} = |A|$$

$$\sum_{j=1}^n a_{ij} M_{kj} = 0 \quad (\text{or}) \quad \sum_{i=1}^n a_{ij} M_{ik} = 0$$

Adjoint matrix:

If B is Cofactor matrix of A then

$$\text{adj}(A) = B^T = (\text{Cofactor matrix})^T$$

Q. $A = \begin{bmatrix} 1 & -2 & -1 \\ 2 & 3 & 1 \\ 0 & 5 & -2 \end{bmatrix}$ Find $\text{Adj}(A)$

$$|A| = -2$$

$\text{Adj } A$:

$$\begin{bmatrix} 3 & 1 & 2 & 3 \\ 5 & -2 & 0 & 5 \\ 2 & -1 & 1 & -2 \\ 3 & 1 & 2 & 3 \end{bmatrix}$$

(or)

starting from middle row middle element

$$\text{Adj}(A) = \begin{bmatrix} -11 & -9 & 1 \\ 4 & -2 & -3 \\ 10 & -5 & 1 \end{bmatrix}$$

Properties of adjoint matrix:

- $A \cdot \text{adj}(A) = \text{adj}(A) \cdot A = |A| I_n$
- $\text{Adj}(0) = 0$; $\text{Adj}(I) = I$
- If D is a diagonal matrix then $\text{adj}(D)$ is also a diagonal matrix.

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & m & 0 \\ 0 & 0 & n \end{bmatrix} \quad \text{adj}(A) = \begin{bmatrix} mn & 0 & 0 \\ 0 & ln & 0 \\ 0 & 0 & lm \end{bmatrix}$$

$$\text{Adj}(A^T) = (\text{adj}(A))^T$$

$$\text{Adj}(AB) = \text{adj}(B) \cdot \text{adj}(A)$$

$$|A| = 0 \text{ then } |\text{adj}(A)| = 0$$

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

$$|\text{adj}(\text{adj } A)| = |A|^{(n-1)^2}$$

$$\text{adj}(\text{adj } A) = |A|^{n-2} \cdot A$$

- If A is symmetric then $\text{adj}(A)$ is also symmetric

$$\text{If } A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \text{ then } \text{adj}(A) = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

Inverse of a matrix:

If A is non-singular matrix then

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

If $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ then $A^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

Properties of Inverse of a square matrix:

- If the inverse of a square matrix A exists then it is unique.
- $(AB)^{-1} = B^{-1}A^{-1}$
- $(ABC)^{-1} = C^{-1}B^{-1}A^{-1}$
- $(A^T)^{-1} = (A^{-1})^T$
- If $AB = AC$ then $B = C$
- If $BA = CA$ then $B = C$
- $(kA)^{-1} = \frac{1}{k} A^{-1}$
- $(I_n)^{-1} = I_n$
- If $A = \begin{bmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{bmatrix}$ then $A^{-1} = \begin{bmatrix} \frac{1}{a} & 0 & 0 \\ 0 & \frac{1}{b} & 0 \\ 0 & 0 & \frac{1}{c} \end{bmatrix}$

Note:

- Num of mulⁿ of $A_{m \times n} B_{n \times p} = mnp$
- Num of Addⁿ of $A_{m \times n} B_{n \times p} = mp(n-1)$

Rank of a matrix:

- A non negative integer ' r ' is said to be rank of matrix A if
 - there exist atleast one non-zero minor of order ' r '
 - All minors of order ' $(r+1)$ ' if they exist are zeros.
- then $\rho(A) = r$

(or)

- The number of non-zero rows in the Echelon form of a matrix can also be observed as Rank of the matrix.

Echelon form:

$$A = \left[\begin{array}{ccc|c} a_1 & b_1 & c_1 & d_1 \\ 0 & b_2 & c_2 & d_2 \\ 0 & 0 & c_3 & d_3 \end{array} \right]_{3 \times 4} \quad B = \left[\begin{array}{ccc|c} 0 & a & b & c \\ 0 & 0 & d & e \\ 0 & 0 & 0 & 0 \end{array} \right] \quad C = \left[\begin{array}{ccc|c} a & b & c & d \\ 0 & 0 & e & f \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \quad D = \left[\begin{array}{ccc} a & b & c \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{array} \right]$$

$\rho(A) = 3 \quad \rho(B) = 2 \quad \rho(C) = 2 \quad \rho(D) = 1$

- Num of leading zero's is increasing from top to bottom like row number is increasing
- All zero rows are present below the non-zero rows only i.e. at the bottom of the matrix.

Any matrix satisfying the above conditions can be observed in Echelon form.

Properties of Rank of a matrix:

- $1 \leq \rho(A) \leq \min(m, n)$
- $\rho(O)_{m \times n} = 0$
- $\rho(I_n) = n$
- $0 \leq \rho(AB) \leq \min\{\rho(A), \rho(B)\}$
- If $|A| \neq 0$ then $\rho(A) = n$ (n rows are independent)
- If $|A| = 0$ then $\rho(A) < n$ (some rows are dependent)
- $\rho(A) = \rho(A^T)$
- $\rho(A+B) \leq \rho(A) + \rho(B)$
- $\rho(A-B) \geq \rho(A) - \rho(B)$

Linearly dependent
 $|A| = 0$

If A has rank ' r ' then interchanging of rows in A doesn't affect the rank ($R_i \leftrightarrow R_j$) $\rho(A) = r$

If A has rank ' r ' and if a row is multiplied by non-zero element then rank will not be affected. ($R_i \rightarrow kR_i$ ($k \neq 0$) $\Rightarrow \rho(A) = r$)

If $R_i \rightarrow kR_i + R_j \Rightarrow \rho(A) = r$

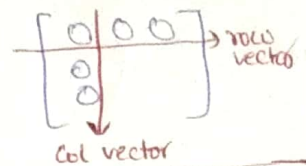
To find rank of a given matrix then we must exclusively apply Row ops or Column operations but not both.

The rank of a diagonal matrix is equal to the num of non-zero diagonal elements.

Note:

In echelon form of a matrix the rows which are made zero rows are dependent and the remaining non-zero rows are independent rows.

Num of independent rows / cols / vector of A = rank of A



System of Linear Equations:

Non-Homogeneous system

$AX = B$ → Constant col vector
Coeff matrix → var vector

Note:

- (i) $P(A) = P(A|B) = n \rightarrow$ only one sol
- (ii) $P(A) = P(A|B) < n \rightarrow$ Infinite sols
- (iii) $P(A) \neq P(A|B) \rightarrow$ No sol

↓
Inconsistent sol

Homogeneous system

$AX = 0$

Note:

- (i) $P(A) = n \rightarrow$ unique sol (zero sol) ($x_1 = x_2 = \dots = 0$)
- (ii) $P(A) < n \rightarrow$ Infinite sols
(zero + non-zero)
trivial ← non-trivial

* If A is $n \times n$ then

- If $|A| \neq 0 \Rightarrow P(A) = n \Rightarrow$ unique (zero) sol
- If $|A| = 0 \Rightarrow P(A) < n \Rightarrow \infty$ (non-zero) sol

If $P(A) = r$ in $AX = 0$ then

Num of independent sols = $n - r$

LU Decomposition:

Ex: $A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 3 & 3 \\ -3 & 0 & 2 \end{bmatrix}$

$= \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -3 & -4 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$

In L:
Now,

$2 = a(1) + 1(c) + 0(b)$
 $\therefore a = 2$

$-3 = b(1) + c(0) + 1(b)$
 $b = -3$

$-10 = b(2) + c(d)$
 $-10 + 6 = cd \Rightarrow cd = 4$
 $c = 4, d = -1$

In U:

$3d = 2(2) + 1(d) + 0(b)$
 $d = -1$

$3 = 2(1) + 3e + 3f$
 $3e + 3f = 1$
 $3 = 2(1) + e$
 $e = 1$

$2 = 1(-5) + 0e$
 $2 = -3(1) + 4(1) + f$
 $f = 1$

$\therefore L = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -3 & -4 & 1 \end{bmatrix} U = \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$

Eigen values and Eigen vectors :

- These properties are applied only on any square matrix
- To find Eigen values of a given matrix $A_{n \times n}$, First we consider the characteristic Equation $|A - \lambda I| = 0$

The roots of this Equations are called characteristic roots | latent roots | prop values | Eigen values

ie $(A) \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \lambda \begin{pmatrix} x \\ y \\ z \end{pmatrix}$

$$(A - \lambda I) \begin{pmatrix} x \\ y \\ z \end{pmatrix} = 0 \Rightarrow |A - \lambda I| = 0 \text{ (Singular)}$$

\downarrow
roots \Rightarrow Eigen values

- To find Eigen vectors of the matrix A then

$$(A - \lambda I) X = 0$$

At diff values of λ and get non-zero sols only and those non-zero sols are called Eigen vectors of A

Properties of Eigen values & Eigen vectors :

- Sum of Eigen values of A = Trace(A)
sum of diagonals of A
- Product of Eigen values of A = det A
- If $A_{n \times n}$ has n distinct eigen values then a set of 'n' independent Eigen vectors of $A_{n \times n}$ exist
- The Eigen values of upper Δ / lower Δ / diagonal matrix are given by its Principal diagonal elements
- If $A_{n \times n}$ has repeated Eigen values then atmost a set of n independent Eigen vectors of $A_{n \times n}$ exist
- If $A = \begin{bmatrix} -2 & 3 & 4 \\ 0 & 4 & 5 \\ 0 & 0 & 3 \end{bmatrix}$ $|A| = 24$
 $\lambda = -2, 4, 3$

(i) $A + 2I \Rightarrow 0, 6, 5$

(ii) $A^3 \Rightarrow -8, 64, 27$

(iii) $5A \Rightarrow -10, 20, 15$

(iv) $A^{-1} \Rightarrow -\frac{1}{2}, \frac{1}{4}, \frac{1}{3}$

(v) $\text{Adj}(A) \Rightarrow -12, 6, 8$

(vi) $A^T \Rightarrow -2, 4, 3$

$$\begin{aligned} A \cdot A^{-1} &= I \\ \lambda \cdot (?) &= 1 \\ \downarrow \\ \frac{1}{\lambda} \end{aligned}$$

$$\begin{aligned} A \cdot \text{Adj}(A) &= |A| \cdot I \\ \lambda \cdot \begin{pmatrix} ? \\ ? \\ ? \end{pmatrix} &= |A| \cdot 1 \\ \downarrow \\ \frac{|A|}{\lambda} \end{aligned}$$

- The Eigen vectors of any real symmetric matrix are orthogonal to each other

ie $x_i^T x_j = 0 \quad \forall i \neq j$

$e_i = \frac{1}{\sqrt{1+1+1}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$

$\lambda = 1$

- If λ is Eigen value of orthogonal matrix then $\frac{1}{\lambda}$ is also an Eigen value to the same matrix.
- If $\begin{bmatrix} a+ib \\ a-ib \end{bmatrix}$ is an Eigen value of A then $\begin{bmatrix} a-ib \\ a+ib \end{bmatrix}$ is also an Eigen value of A .
- The Eigen values of a skew symmetric matrix is either zero or purely imaginary.
- '0' is an Eigen value iff $|A| = 0$

Eigen vectors :

- Eigen vectors of $\begin{bmatrix} A \text{ and } A^m \\ A \text{ and } A^{-1} \\ A \text{ and } KA \end{bmatrix}$ are same
- Eigen vectors of $\begin{bmatrix} A \text{ and } A^T \end{bmatrix}$ are not same
- Eigen vectors of symmetric matrix are orthogonal.

Algebraic multiplicity

- If an Eigen value λ of a square matrix A of order n is repeated m times then the number ' m ' is called algebraic multiplicity of an Eigen value λ .

IF $\lambda = \underbrace{1, 1, 1, 1}_4, 2$

AM of 1 is 4

Geometric multiplicity :

- If ' p ' num of linearly independent Eigen vectors of matrix A of order ' n ' to an Eigen value ' λ ' then p is called Geometric multiplicity of λ ie

$$\begin{aligned} P &= n - r \\ P &= (\text{num of var}) - P(A - \lambda I) \end{aligned}$$

Cayley Hamilton theorem :

Every square matrix satisfies its own characteristic Equation.

Ex:- If $\lambda^3 - 12\lambda^2 + 36\lambda - 32 = 0$ is CE then

A/c to CH theorem $\boxed{A^3 - 12A^2 + 36A - 32 = 0}$

Applications :

- To find higher powers of matrix A
- To find inverse of matrix A

X — The End — X