

UNIT - 1

Matrix

It is transformation from one space to another space

Eigen Value & Eigen Vector of Matrix

Let A be a square matrix. Then λ is called the Eigen value of A if $Ax = \lambda x$. Then x is called the corresponding Eigen vector to the eigen value λ .

$$|A - \lambda I| = 0$$

If A is $[3 \times 3]$ square matrix then λ (λ = Eigen value of matrix A) is root of equation

$$\lambda^3 - S_1\lambda^2 + S_2\lambda - S_3 = 0$$

where

S_1 = sum of principal diagonal elements

S_2 = sum of ~~matrix~~ minors containing diagonal elements
 (m_{11}, m_{22}, m_{33})

S_3 = determinate of A .

Properties of Eigen Values

- ① If $\lambda_1, \lambda_2, \dots, \lambda_n$ are the Eigen values of ~~values of~~ A then $\lambda_1^m, \lambda_2^m, \dots, \lambda_n^m$ are the Eigen values of A^m
- ② If $\lambda_1, \lambda_2, \dots, \lambda_n$ are the Eigen values of A , then $K\lambda_1, K\lambda_2, \dots, K\lambda_n$ are the Eigen values of KA where K is a constant
- ③ The sum of the Eigen values of A is equal to the diagonal entries of A (trace of A)
- ④ The Product of the Eigen of A is equal to $|A|$
- ⑤ A square matrix A and its transpose A^T have the same Eigen value
- ⑥ If $\lambda_1, \lambda_2, \dots, \lambda_n$ are Eigen values of square matrix A of order n Then $\frac{1}{\lambda_1}, \frac{1}{\lambda_2}, \dots, \frac{1}{\lambda_n}$ are eigen values of A^{-1} where ($\lambda_i \neq 0$)
- ⑦ If A is an upper triangular or a lower triangular matrix, then Eigen Values of A are its diagonal entries.

Cayley-Hamilton Theorem

Every square matrix satisfy its own characteristic equation

$$A_{2 \times 2} \Rightarrow A^2 - S_1 A - S_2 = 0$$

S_1 = trace of A

$S_2 = |A|$

$$A_{3 \times 3} \Rightarrow A^3 - S_1 A^2 + S_2 A - S_3 = 0$$

S_1 = trace of A

S_2 = sum of minors containing diagonal elements (m_{11}, m_{22}, m_{33})

$S_3 = |A|$

Similar matrix

Let A and B be two square matrices. The matrices A and B are said to be similar if there exists a non singular matrix P such that

$$A = P^{-1} B P$$

Similar matrices have the same Eigen values

Diagonal matrix

A square matrix A is said to be diagonal if $a_{ij} = 0, i \neq j$

Diagonalisation of a square matrix

Step I - obtain eigen values of matrix

Step II obtain Eigen Vectors

Step III form a modal matrix (M) with x_1 as first
 x_2 as second, x_3 as third column

Step IV Find m^{-1}

Step V The diagonal matrix D is given by
$$D = m^{-1} \cdot A \cdot m$$

$$D = \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix} \quad d_1, d_2, d_3 \text{ are eigenvalues of } A$$

Matrix A and D have same Eigen values

If A has n -distinct Eigen values then there must be n linearly independent Eigen vectors $x_1, x_2, x_3, \dots, x_n$

$$\text{Then } m = [x_1, x_2, x_3, \dots, x_n]$$

Orthogonal matrix

A real square matrix A is said to be an orthogonal matrix if $AA^T = A^TA = I$

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

Since A is a orthogonal matrix $A^T = A^{-1}$

Symmetric matrix

A real square matrix A is said to be symmetric if $A^T = A$

Some properties of orthogonal and symmetric matrices

- ① If A is orthogonal then A^T is also orthogonal
- ② If A is orthogonal then $|A| = \pm 1$
- ③ If λ is an eigen value of the orthogonal matrix A then $1/\lambda$ is also eigen value of A
- ④ If A and B orthogonal matrix then AB is also orthogonal
- ⑤ If A is orthogonal matrix then the magnitude of the eigen values of A is 1

6) Eigen values of real symmetric matrix are real

7) Eigen vectors corresponds to distinct Eigenvalues of a symmetric matrix are orthogonal vectors

Unit Vector

Let x be a non zero vector, then its unit vector is

$$x = (a, b, c)$$

$$\hat{x}_1 = \frac{x_1}{\sqrt{a^2 + b^2 + c^2}}$$

Diagonalisation by orthogonal Transformation

Step I : Find Eigen values $\lambda_1, \lambda_2, \dots, \lambda_n$

Step II Find the linearly independent Eigen-vectors
they are pairwise orthogonal i.e. $x_i \perp x_j \quad i \neq j$

Step III Calculate unit vector

Step IV Form modal matrix

$$\text{Step V} \quad N^T = N^{-1}$$

$$\text{Step 6} \quad N^{-1} A N = D$$

Quadratic forms

A homogeneous polynomial of second degree in any number of variables is called Quadratic form

Any quadratic form can be expressed in matrix form considering a quadratic

$$ax^2 + by^2 + cxy = \begin{bmatrix} x & y \end{bmatrix} \begin{bmatrix} a & c/2 \\ c/2 & b \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

Reduction to Canonical form

A quadratic equation which only contains only lies in squared terms is called canonical form

Step I : Find A

Step II :- Find Eigen Values & Eigen Vectors of A

Step III - check whether the Eigen Vectors are orthogonal to each other

Step IV : Construct modal matrix (P) & find normalised modal matrix N

Step I Find $N^T A N = D$

Step II Find $Y Q Y^T$ where $[Y_1, Y_2, Y_3]$

Nature of the Canonical form

The number of non-zero terms = r (rank of A)

The number of positive terms = S (Index of Q.F.)

The difference of number of positive & negative square terms = $2S - r$
(Signature of the Q.F.)

The number of positive Eigen values of A
= Index (P)

Definite & Indefinite quadratic forms

Let $Q = \mathbf{x}^T \mathbf{A} \mathbf{x}$ be a quadratic form

- ① Q is said to be positive definite if all eigenvalues of \mathbf{A} are positive
- ② Q is said to be negative definite if all eigenvalues are negative
- ③ Q is said to be positive semidefinite if all eigenvalues are non-negative and at least one of them is zero
- ④ Q is said to be negative semidefinite if all eigenvalues are non-positive and at least one of them is zero
- ⑤ Q is said to be indefinite in all other cases

The Quadratic form is said to be

- (i) Positive definite if $r=n, s=0$
- (ii) Positive semidefinite if $r \leq n, s=0$ $\left\{ \begin{array}{l} r = \text{no. of variables} \\ \text{in Q.F} \end{array} \right.$
- (iii) negative definite if $r=n, s=0$
- (iv) negative semidefinite if $r \leq n, s \geq 0$
- (v) Indefinite in all other cases

Principal minor is denoted by D

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ \vdots & & & & \ddots \\ a_{n1} & a_{n2} & a_{n3} & \cdots & a_{nn} \end{bmatrix}$$

$$D_1 = |a_{11}| = a_{11}$$

$$D_{33} = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

$$D_2 = (2 \times 2) \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}$$

- ① A Q.P is Positive definite if $D_i > 0$
- ② A Q.P is negative definite if $(-1)^i D_i > 0$
- ③ Q.P is positive semi definite if $D_i \geq 0$ and atleast one $D_i = 0$
- ④ Q.P is negative semi definite if $(-1)^i D_i \leq 0$ & atleast one $D_i = 0$
- ⑤ In all other cases, it is indefinite

UNIT - 2

Suppose $u = f(u, v)$

Then $\frac{\partial u}{\partial x}$ = Partial derivative of function u with respect to x alone by treating other variable as constant

$$\frac{\partial^2 u}{\partial x^2} = \frac{\partial}{\partial x} \left(\frac{\partial u}{\partial x} \right).$$

Homogeneous function

A function $f(x, y, z)$ is called a homogeneous function of degree if

$$f(tx, ty, tz) = t^n f(x, y, z)$$

Euler's theorem

If $f(x, y)$ is a homogeneous function of degree n

$$\text{then } n \frac{\partial f}{\partial x} + y \frac{\partial f}{\partial y} = nf$$

Total differentiation

Suppose

$$u = u(x, y, z) \quad \begin{cases} x = x(t) \\ y = y(t) \\ z = z(t) \end{cases}$$

Then total differentiation

$$\frac{du}{dt} = \frac{\partial u}{\partial x} \frac{dx}{dt} + \frac{\partial u}{\partial y} \frac{dy}{dt} + \frac{\partial u}{\partial z} \frac{dz}{dt}$$

Maxima & minima

$$y = f(x) \quad \frac{dy}{dx} = 0$$

Then condition for maxima & minima

If (a) is a extreme point

$$\left. \frac{d^2 y}{dx^2} \right|_{(a)} > 0 \quad \text{minima}$$

$$\left. \frac{d^2 y}{dx^2} \right|_{(a)} < 0 \quad \text{maxima}$$

Let $P = f(x, y)$

The condition for maxima or minima are

$$\frac{\partial f}{\partial x} = 0, \quad \frac{\partial f}{\partial y} = 0$$

If (a, b) is an extreme point then

$$\left. \frac{\partial f}{\partial x} \right|_{(a,b)} = 0, \quad \left. \frac{\partial f}{\partial y} \right|_{(a,b)} = 0$$

$$r = \frac{\partial^2 f}{\partial x^2}, \quad t = \frac{\partial^2 f}{\partial y^2}, \quad s = \frac{\partial^2 f}{\partial x \partial y}$$

If $rt - s^2 > 0$ & $r > 0$
 f attains maximum value at (a, b)

If $rt - s^2 > 0$ & $r < 0$
 f attains minimum value at (a, b)

If $rt - s^2 = 0$
 f either does not attain any extremal value

Constrained maxima or minima using Lagrange's
multiple method

Suppose $f = f(x, y)$ to be a function connected
by the constraint

$$\phi(x, y, z) = 0$$

Thus the auxiliary eqns

$$\frac{\partial F}{\partial x} = 0, \frac{\partial F}{\partial y} = 0, \frac{\partial F}{\partial z} = 0$$

where $F = f + \lambda \phi$

Step 1: construct $F = f + \lambda \phi$

Step 2: obtain $\frac{\partial F}{\partial x}, \frac{\partial F}{\partial y}, \frac{\partial F}{\partial z}$

Step 3: solve $\frac{\partial F}{\partial x} = 0, \frac{\partial F}{\partial y} = 0, \frac{\partial F}{\partial z} = 0$

for x, y, z

Taylor Series Expansion

Suppose $f = f(x, y)$

Then the Taylor Series Expansion of f about (a, b)

$$\begin{aligned} \text{is } f(x, y) &= f(a, b) + \frac{1}{1!} \left[\frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} \right]_{(a,b)}^{} f(a, b) \\ &+ \frac{1}{2!} \left[\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} \right]_{(a,b)}^{} f(a, b) + \frac{1}{3!} \left[\frac{\partial^3 f}{\partial x^3} + \frac{\partial^3 f}{\partial y^3} \right]_{(a,b)}^{} f(a, b) \end{aligned}$$

$$h = (x - a), \quad k = y - b$$

Jacobians

If u and v are functions of two independent variables x & y , then the Jacobian of u, v with respect to x, y is given by

$$\frac{\partial(u, v)}{\partial(x, y)} = \begin{vmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} \end{vmatrix}$$

Properties of Jacobians

1) If J_1 is Jacobians of u, v with respect to x, y & J_2 is another Jacobians of u, v with respect to u, v then

$$J_1 \times J_2 = 1$$

2) If u, v are function of r, s and r, s are function of x, y then

$$\frac{d(u,v)}{d(x,y)} = \frac{d(u,v)}{d(r,s)} \times \frac{d(r,s)}{d(x,y)}$$

UNIT - 3

The general form of a differential equation is

$$a_0 \frac{d^2y}{dx^2} + a_1 \frac{dy}{dx} + a_0 y = f(x)$$

Homogeneous differential Equation

$$a_0 \frac{d^2y}{dx^2} + a_1 \frac{dy}{dx} + a_0 y = 0$$

Complementary Function

$$D = \frac{d}{dx} \quad D^2 = \frac{d^2}{dx^2}$$

$$(a_0 D^2 + a_1 D + a_0 y) y = 0$$

The auxiliary Eqⁿ is

$$a_0 m^2 + a_1 m + a_0 = 0$$

Case(1) Roots are real but distinct $m_1 \neq m_2$

$$C.F = C_1 e^{m_1 x} + C_2 e^{m_2 x}$$

Case(II) Roots are real but equal $m_1 = m_2 = m$

$$C.P. = (C_1 + C_2 x) e^{mx}$$

Case(III) Roots are complex

$$m_1 = \alpha + i\beta \quad m_2 = \alpha - i\beta$$

$$C.P. = e^{\alpha x} (C_1 \cos \beta x + C_2 \sin \beta x)$$

Heterogeneous Function

Type-I

$$(a_0 D^2 + a_1 D + a_2) y = e^{\alpha x}$$

$$y = C.P. + P.I.$$

$$P.I. = \frac{e^{\alpha x}}{a_0 D^2 + a_1 D + a_2} \quad \text{Replace } D \text{ by } a$$

$$P(D) = a_0 D^2 + a_1 D + a_2$$

$$= \frac{e^{\alpha x}}{P(a)}$$

If $F(a) = 0$ then

$$P.I = \frac{x e^{ax}}{P'(a)} = \frac{x e^{ax}}{P'(a)}$$

If $F'(a) = 0$ then

$$P.I = \frac{x^2 e^{ax}}{P''(a)}$$

Type - 2

$$(a_0 D^2 + a_1 D + a_2) y = \sin(ax) \text{ or } \cos(ax)$$

$$P.I. = \frac{\sin(ax) / \cos(ax)}{a_0 D^2 + a_1 D + a_2} \quad D^2 \text{ by } -a^2$$

If Denominator is 0 then multiply x to Numerator & differentiate the denominator then make D^2 term in denominator

Type - 3

$$(a_0 D^2 + a_1 D + a_2) y = b_0 x^2 + b_1 x + b_2$$

$$P.I = \frac{b_0 x^2 + b_1 x + b_2}{a_0 D^2 + a_1 D + a_2} = \frac{1}{1 \pm P(D)}$$

$$\left\{ \begin{array}{l} \frac{1}{1-x} = 1 + x + x^2 + x^3 - \dots \\ \frac{1}{1+x} = 1 - x + x^2 - x^3 + x^4 - \dots \end{array} \right.$$

Type-4

$$P(D)y = e^{ax}, y$$

$$P.O.I = \frac{e^{ax}, y}{P(D)} = e^{ax} \left[\frac{y}{P(D+a)} \right] D \rightarrow D+a$$

Type-5

$$a x^2 \frac{d^2 y}{dx^2} + b x y + c y = P(x)$$

$$x = e^z, z = \log x$$

$$x D = D'$$

$$D' = \frac{d}{dz}$$

$$x^2 D^2 = D'(D'-1)$$

Legendre Type Equation

$$(ax+b)^n \frac{d^2y}{dx^2} + (ax+b) \frac{dy}{dx} + a_1 y = P(x)$$

Let $ax+b = e^z$, $z = \log(ax+b)$

$$(ax+b)D = aD'$$

$$(ax+b)^n = a^n D'(D-1)$$

Method variation of Parameters

$$a_0 \frac{d^2y}{dx^2} + a_1 \frac{dy}{dx} + a_2 y = P(x)$$

Step 1: Find C.P

$$C.P = C_1 f_1(x) + C_2 f_2(x)$$

Step 2: Find ω

$$\omega = \begin{vmatrix} f_1 & f_2 \\ f'_1 & f'_2 \end{vmatrix}$$

$$\text{Step 3: } P = - \int_{\omega} f_1 \cdot F(x) dx$$

$$Q = \int_{\omega} f_2 \cdot F(x) dx$$

$$\# P \cdot I = P \cdot f_1 + Q \cdot f_2.$$

$$\text{Step 4: } y = c \cdot P + P \cdot f_1 + Q \cdot f_2$$

UNIT - 4

Radius of Curvature

Forms of the curve

1) Cartesian form

$$y = f(x)$$

$$\rho = \frac{(1 + (y')^2)^{3/2}}{|y''|}$$

$$y' = \frac{dy}{dx} \quad y'' = \frac{d^2y}{dx^2}$$

2) Parameter form

$$x = x(t)$$

$$y = y(t)$$

$$\rho = \frac{(\dot{x}^2 + \dot{y}^2)^{3/2}}{\dot{x}\ddot{y} - \dot{y}\ddot{x}}$$

$$\dot{x} = \frac{dx}{dt} \quad \dot{y} = \frac{dy}{dt}$$

$$\ddot{x} = \frac{d^2x}{dt^2} \quad \ddot{y} = \frac{d^2y}{dt^2}$$

3) Polar form

$$\rho = \frac{r^2 f(\theta)^{3/2}}{r^2 + 2\dot{r}^2 - r\ddot{r}}$$

$$\dot{r} = \frac{dr}{d\theta}, \quad \ddot{r} = \frac{d^2r}{d\theta^2}$$

Centre of the Curve

$$\bar{x} = x - \frac{y_1(1+y_1^2)}{y_2} \quad y_1 = \frac{dy}{dx}$$

$$\bar{y} = y - \frac{(1+y_1^2)}{y_2} \quad y_2 = \frac{d^2y}{dx^2}$$

Evolutes of a curve

The locus of centre of curvature of a curve
is called evolutes

Step 1: Find (\bar{x}, \bar{y})

Step 2: Eliminate the parameter present in \bar{x}, \bar{y}

Equation of circle of curvature is

$$(x - \bar{x})^2 + (y - \bar{y})^2 = \rho^2$$

Parametric form

Ellipse

$$\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$$

$$x = a \cos \theta \quad y = b \sin \theta$$

Parabola

$$y^2 = 4ax$$

$$x = at^2 \quad y = 2at$$

Hyperbola

$$\frac{x^2}{a^2} - \frac{y^2}{b^2} = 1$$

$$x = a \sec \theta \quad y = b \tan \theta$$

Circle

$$x^2 + y^2 = a^2$$

$$x = a \cos \theta \quad y = a \sin \theta$$

Rectangular hyperbola

$$xy = c^2 \quad x = ct \quad y = \frac{c}{t}$$

Envelope

1 Parameter

Suppose $f(x, y, \alpha) = 0$ be family of curves and α is a parameter

Step 1:- Find $\frac{\partial f}{\partial \alpha}$

$$\frac{\partial f}{\partial \alpha} = 0 \quad \text{--- (1)} \quad \text{and} \quad f(x, y, \alpha) = 0 \quad \text{--- (2)}$$

Step 2: Eliminating α from (1) + (2)
we get envelope

Suppose the family of curve is expressed as
a quadratic equation in a parameter

$$\text{i.e. } Ax^2 + Bx + C = 0$$

Then eqn of Envelope is

$$B^2 - 4AC = 0$$

To Envelope when two parameters are given

Suppose $F(x, y, a, b) = 0 \quad \text{--- (1)}$

$$\phi(a, b) = 0 \quad \text{--- (2)}$$

Step 1:- Differentiate partially (1) w.r.t 'a' by
treating 'b' as function of 'a' and then call
the eqn as (3)

Step 2: Diff partially (2) w.r.t 'a' and call
the eqn as (4)

Step 3: Find $\frac{\partial \phi}{\partial a}$ from (4) and substitute in (3)

Step 4: Eliminate 'a' and 'b'

Gamma functions & Beta functions

Gamma function

$\Gamma(n) = \int_0^\infty x^{n-1} e^{-x} dx$ is called Gamma function

$\Gamma(n) = (n-1)!$ if n is integer

$\Gamma(n+1) = n \Gamma(n)$

$\int_0^\infty x^{n-1} e^{-ax} dx = \frac{\Gamma(n)}{a^n}$

$\Gamma(\frac{1}{2}) = \sqrt{\pi}$

Beta function

$$\beta(m,n) = \int_0^1 x^{m-1} (1-x)^{n-1} dx$$

$$\beta(m,n) = \beta(n,m)$$

$$\beta(m,n) = \int_0^{\pi/2} \sin^{2m-1}\theta \cos^{2n-1}\theta d\theta$$

$$\# \beta(m,n) = \frac{\Gamma(m), \Gamma(n)}{\Gamma(m+n)}$$

UNIT - 5

A set of numbers, $a_1, a_2, a_3, \dots, a_n$ such that to each positive integer n , the corresponding a_n of the set is called Sequence and it denoted by $\{a_n\}$

Limit of a sequence

A sequence converges to a limit (l) if given very positive number ϵ , there exist a positive small integer N such that

$$|a_n - l| < \epsilon \quad \forall n \geq N$$

Bounded Sequence

A sequence a_n is said to be bounded if $a_n \leq M \quad \forall n$

Monotonically increasing Sequence

A Sequence a_n is said to be monotonically increasing if $a_n \leq a_{n+1} \quad \forall n$

Monotonically decreasing Sequence

A sequence a_n is said to be monotonically decreasing
if $a_n \geq a_{n+1} \forall n$

Converging Series.

A sequence a_n is said to be converge to l
if $|a_n - l| < \epsilon \forall n \geq N$ where $\lim_{n \rightarrow \infty} a_n = l$

Comparison test

If $\sum u_n$ and $\sum v_n$ are two series and the second series $\sum v_n$ is convergent, $u_n \leq k v_n$ then $\sum u_n$ is convergent

Limit form

If $\sum u_n$ and $\sum v_n$ are two series and $\sum v_n$ is convergent and $\frac{u_n}{v_n}$ tends to a limit

other than zero as $n \rightarrow \infty$ then $\sum u_n$ is convergent.

If $\sum v_n$ is divergent & $\frac{u_n}{v_n}$ tends to a limit other than zero as $n \rightarrow \infty$ then $\sum u_n$ is divergent.

If two positive terms $\sum u_n$ and $\sum v_n$ be such that $\lim_{n \rightarrow \infty} \frac{u_n}{v_n} = \text{finite number} \neq 0$

then both series converge or diverge together

$$\sum v_n = \frac{1}{n^p}$$

$p > 1 \rightarrow \text{converges}$
 $p \leq 1 \rightarrow \text{diverges}$

D'Alembert's ratio test

If $\sum u_n$ is a positive term series such that

$$\lim_{n \rightarrow \infty} \frac{u_{n+1}}{u_n} = K \quad \text{then}$$

① series is convergent if $K < 1$

② series is divergent if $K > 1$

Raabe's test

If $\sum u_n$ is a positive term series

$$\lim_{n \rightarrow \infty} \left\{ n \left(\frac{u_{n+1}}{u_n} - 1 \right) \right\} = \ell$$

① Series is convergent if $\ell > 1$

② Series is divergent if $\ell < 1$

Cauchy's Root test

If $\{c_n\}$ is a positive term series such that

$$\lim_{n \rightarrow \infty} (c_n)^{\frac{1}{n}} = \ell$$

① series is convergent if $\ell < 1$

② series is divergent if $\ell > 1$