

Q.01: Give Examples of
a) vector space b) finite dimensional vector space
c) infinite dimensional vector space.

a) vector space:

Set: 1) $X = \mathbb{R}$ over \mathbb{R}

2) $X = \text{Complex no's}$ over complex numbers

3) $X = \text{set of complex no's}$ over complex numbers

4) $C[0,1]$ on $[0,1]$

5) $X = P[0,1]$ over $[0,1]$

6) $X = P_n[0,1]$ & degree $\leq n$

7) $X = \{0\} : K = \mathbb{R}$

8) set of all $m \times n$ matrices usual matrix addition over K .

9) $Ax = 0$ [set of all homogeneous eqns]

b) Finite dimensional vector space

1) $\{0\}$

$$\dim(\{0\}) = 0$$

2) $P_n[0,1]$

$$\dim(P_n[0,1]) = n+1$$

Basis = $(1, x, x^2, \dots, x^n)$

3. Standard Basis of \mathbb{R}^n

$$x = k_1(1, 0, 0, \dots, 0) +$$

$$k_2(0, 1, 0, \dots, 0) +$$

$$\dots$$

$$k_n(0, 0, \dots, 1)$$

c) Infinite dimensional vector space.

1. set of continuous functions

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

2. ~~continuous~~ $C[0,1]$

$$\dim(C[0,1]) = \infty$$

3. $P[0,1] \& P[-1,1]$

$$\dim(P[0,1]) = \infty, \dim(P[-1,1]) = \infty$$

Basis = $(1, x, x^2, \dots, \infty)$

4. Set of all non convergent sequences.

5. All the differential fn's from \mathbb{R} to it self

3. Consider the following matrix

$$x = \begin{bmatrix} 3 & 4 \\ 6 & 2 \end{bmatrix}$$

a) Find the rank of x on the basis of rank of x , discuss the existence of the inverse of x .

Sol: $x = \begin{bmatrix} 3 & 4 \\ 6 & 2 \end{bmatrix} \Rightarrow |x| = 6 - 24 = -18; \boxed{\text{Rank}(x) = 2}$

As $|x| \neq 0 \rightarrow$ non-singular matrix.

$$A^{-1} = \frac{\text{adj} A}{|A|} \Rightarrow x^{-1} = \frac{\text{adj } x}{|x|} \quad \begin{array}{l} \text{non-singular} \\ \text{matrix} \end{array}$$

\uparrow

exists

$|x| \neq 0$

\therefore For all non-singular matrix inverse exists

b) Find the condition number

$$\boxed{\text{Condition number} = \|A\|_1 \|A^{-1}\|_1}$$

Infinity norm:

$$\text{Condition number} = \|A\|_{\infty} \cdot \|A^{-1}\|_{\infty}$$

$\|A\|_{\infty} \Rightarrow$ Row sum matrix.

$$\|A\|_{\infty} = 8; \quad \|A^{-1}\|_{\infty} = 12$$

$$\therefore \text{Condition number} = \|A\|_{\infty} \cdot \|A^{-1}\|_{\infty} \\ = 8 \times \frac{1}{2} = 4.$$

since ~~$\|A\|_{\infty}$~~

$$\max \left\{ \begin{array}{l} \rightarrow [3 \ 4] \rightarrow 3+4=7 \\ \rightarrow [6 \ 2] \rightarrow 6+2=8 \end{array} \right.$$

$$\max(7, 8) = 8$$

$$x = \begin{bmatrix} 3 & 4 \\ 6 & 2 \end{bmatrix} \quad x^{-1} = \frac{1}{-18} \begin{bmatrix} 2 & -4 \\ -6 & 3 \end{bmatrix}$$

One norm:

$$\text{Condition number} = \|A\|_1 \|A^{-1}\|_1$$

$\|A\|_1 \Rightarrow$ column sum matrix.

$$\|A\|_1 = \left[\begin{array}{c} \downarrow \\ 3 \\ 6 \end{array} \quad \begin{array}{c} \downarrow \\ 4 \\ 2 \end{array} \right]$$

$$\|A\|_1 = \max(9, 6) \\ = 9$$

$$\|A^{-1}\|_1 = \left[\begin{array}{c} \downarrow \\ \frac{2}{-18} \\ -\frac{6}{-18} \\ \downarrow \\ \frac{3}{-18} \\ \downarrow \\ \frac{1}{-18} \end{array} \right] \quad \left[\begin{array}{c} \downarrow \\ \frac{-4}{-18} \\ \frac{3}{-18} \\ \downarrow \\ \frac{1}{-18} \\ \downarrow \\ \frac{7}{-18} \end{array} \right]$$

$$\|A\|_1 \cdot \|A^{-1}\|_1 = 9 \times \frac{8}{18} = 4$$

$$\boxed{\text{Condition number} = 4}$$

6.

Give an Example of Monotonically increasing and decreasing function

sol: Monotonic function:

let f be a function defined on closed interval $[a, b]$. Then:

① f is said to be monotonically increasing on $[a, b]$

if for $x_1, x_2 \in [a, b]$, $x_1 > x_2$

$$x_1 > x_2$$

$$f(x_1) \geq f(x_2)$$

② f is said to be monotonically decreasing on $[a, b]$

if $x_1, x_2 \in [a, b]$

$$x_1 > x_2$$

$$f(x_1) \leq f(x_2)$$

③ f is said to be monotonic if f is either monotonically increasing (or) decreasing.

④ f is said to be strictly monotonically increasing on $[a, b]$

if for $x_1, x_2 \in [a, b]$

$$x_1 > x_2$$

$$f(x_1) > f(x_2)$$

⑤ f is said to be strictly monotonically decreasing on $[a, b]$

if for $x_1, x_2 \in [a, b]$

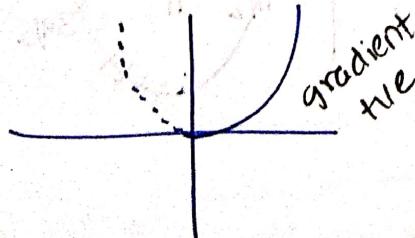
$$x_1 > x_2$$

$$f(x_1) < f(x_2)$$

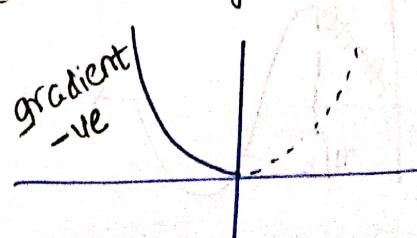
⑥ f is said to be strictly monotonic if

f is either strictly monotonically increasing

(i) Strictly monotonically decreasing.

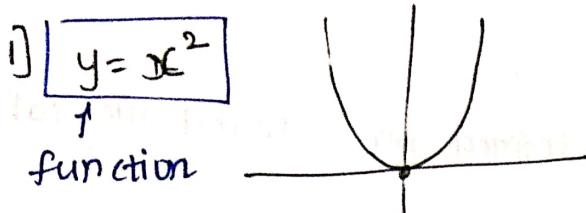


monotonically increasing



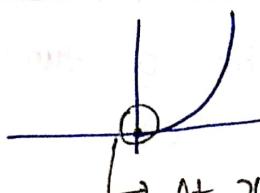
monotonically decreasing

Monotonic increasing:



monotonically increasing

$$y = x^2; \quad x > 0$$



At $x=0 \Rightarrow$ stationary point

$$x_1 = 4 > 0 \quad x_2 = 2 > 0$$

$$f(x_1) = 16 > f(x_2) = 4$$

Neither increasing nor decreasing at $x=0$

2) $f(x) = \log x; \quad x > 0$

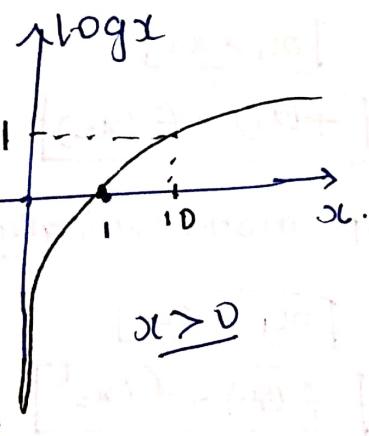
$$x_1 = 1 > 0 \quad x_2 = 10 > 0$$

$$x_2 > x_1$$

$$f(x_1) = \log(1) = 0$$

$$f(x_2) = \log(10) = 1$$

$$f(x_2) > f(x_1)$$



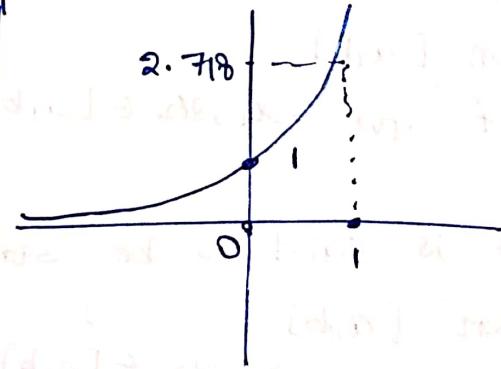
3) $f(x) = e^x$

$$x_1 = 0 \quad x_2 = 1$$

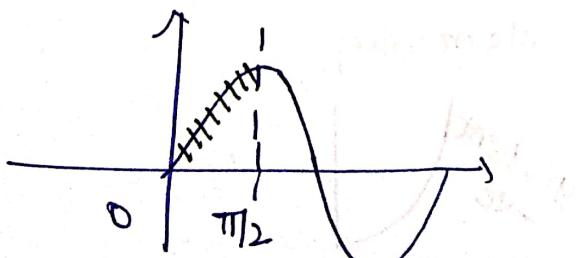
$$x_2 > x_1$$

$$f(x_1) = e^0 = 1 \quad f(x_2) = e^1 = 2.718$$

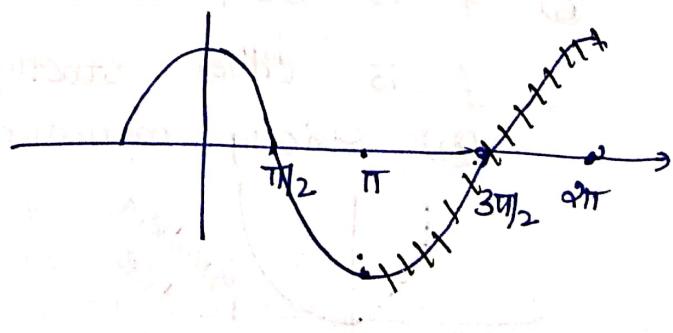
$$f(x_2) > f(x_1)$$



4) $f(x) = \sin x \quad [0 \text{ to } \pi/2]$



5) $f(x) = \cos x \quad [\pi/2 \text{ to } 2\pi]$



6] polynomial with positive coefficients.

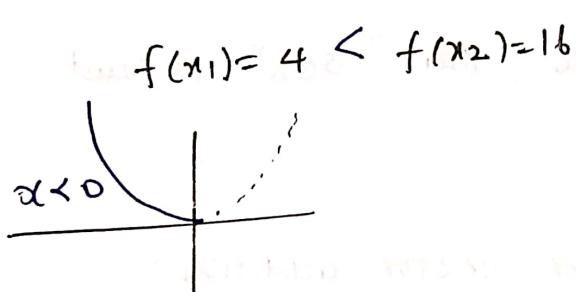
$$P_n = a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n \quad \text{for } x > 0$$

7] $f(x) = x \cdot \ln(x) \quad \forall x \in (1/e, \infty)$

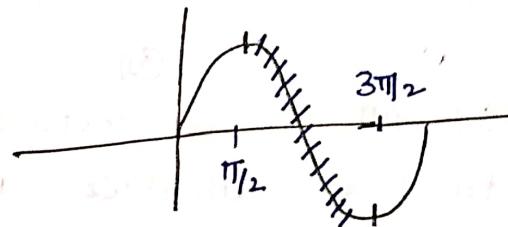
Monotonic decreasing fns:

1) $y = x^2$: $x < 0$

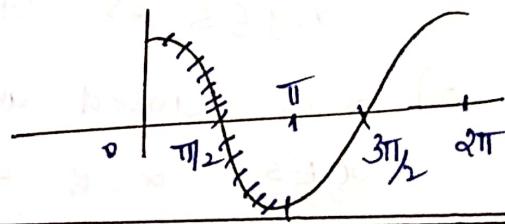
$$x_1 = -2 > x_2 = -4$$



2) $f(x) = \sin x \quad x \in [\pi/2, 3\pi/2]$



3) $f(x) = \cos x \quad x \in [0, \pi]$



4) $f(x) = x \cdot \ln(x)$

$$\forall x \in (0, 1/e)$$

Q.7. Find the projection of x onto $V = \text{span}\{u, v, w\}$

$$x = (-1, 1, 1, 2)^T \quad u = (0, 2, -2, 1)^T \quad v = (1, 3, 0, 1)^T \quad w = (1, 1, 1, 1)^T$$

Sol: The projection of y onto $R(x)$.

$p(x)$ is a vector in $R(x)$ which is the smallest distance to y . That is

$$p(x) = \arg \min_{y' \in R(x)} \|y - y'\|$$

$$\begin{aligned} \|x - u\| &= \sqrt{(-1)^2 + (-1)^2 + (3)^2 + (1)^2} \\ &= \sqrt{1+1+9+1} = \sqrt{12} \end{aligned}$$

$$\begin{aligned} \|x - v\| &= \sqrt{(-2)^2 + (-2)^2 + (1)^2 + (1)^2} \\ &= \sqrt{4+4+1+1} = \sqrt{10} \end{aligned}$$

$$\begin{aligned} \|x - w\| &= \sqrt{(-2)^2 + (0)^2 + (0)^2 + (1)^2} \\ &= \sqrt{4+1} = \sqrt{5} \end{aligned} \quad \therefore \text{projection of } x \text{ onto } V$$

$$w = (1, 1, 1, 1)^T$$

(Q.08) Explain the term subspace of a vector space, span of a set of vectors.

Sol: (i) Subspace of a vector space:

→ If R^n is a vector space over R, then a subset S of R^n is said to be a subspace of vector space R^n : if S is also a vector space over field R.

(ii)

→ If R^n is a vector space, then SGR^n is said to be a subspace if

i) $0 \in S$

2) S is closed under vector addition:
 $x, y \in S \Rightarrow x+y \in S$

3) S is closed under scalar multiplication
 $\alpha \in S, \alpha \in R \Rightarrow \alpha \cdot x \in S$

(ii) Span of Set of Vectors:

→ If $S = \{v_1, v_2, \dots, v_k\}$ is a set of vectors in a vector space V, the span of S is the set of all linear combinations of the vectors in S.

$$\text{span}(S) = \{ c_1 v_1 + c_2 v_2 + \dots + c_k v_k \mid \forall c_i \in R \}$$

→ If every vector in a given vector space can be written as a linear combination of vectors in a given set S, then S is called a spanning set of vector space.

a] Range space of matrix transformation

$A: \mathbb{R}^n \rightarrow \mathbb{R}^m$ a subspace of \mathbb{R}^m

sol: $y = w \cdot x$: $y \in \mathbb{R}^m$
 $w \cdot x \in \mathbb{R}^{m \times n}$

let $x \in \mathbb{R}^n$ $x \cdot w = y$ $x: \mathbb{R}^n \rightarrow \mathbb{R}^m$

$R(x)$ [where $x: \mathbb{R}^n \rightarrow \mathbb{R}^m$].

$y_1, y_2 \in R(x)$.

since y_1, y_2 are $R(x)$, they should have preimages.

$y_1 = xw^1 \in \mathbb{R}^m; y_2 = xw^2 \in \mathbb{R}^m$ w^1 is preimage of y_1
 w^2 is preimage of y_2 .

$$\alpha y_1 + \beta y_2 = \alpha xw^1 + \beta xw^2$$

$$w^1, w^2 \in \mathbb{R}^n$$

$$\text{and } = x[\alpha w^1 + \beta w^2]$$

where $\alpha w^1 + \beta w^2 = w \in \mathbb{R}^n$

$$\therefore \alpha y_1 + \beta y_2 = x \cdot w \in \mathbb{R}^m$$

∴ Range space of matrix transformation

$A: \mathbb{R}^n \rightarrow \mathbb{R}^m$ a subspace of \mathbb{R}^m

b) All vectors $(v_1, v_2, v_3)^T$ in \mathbb{R}^3 with $v_1 - v_2 + 2v_3 = 0$
a subspace of \mathbb{R}^3

let $(v_1, v_2, v_3)^T \in \mathbb{R}^3$ & $(u_1, u_2, u_3)^T \in \mathbb{R}^3$
 $v_1 - v_2 + 2v_3 = 0$

$$\alpha v + \beta u \rightarrow (\alpha v_1 + \beta u_1, \alpha v_2 + \beta u_2, \alpha v_3 + \beta u_3)$$

$$z_1 - z_2 + 2z_3 = \alpha v_1 + \beta u_1 - [\alpha v_2 + \beta u_2] + \alpha \cdot (\alpha v_3 + \beta u_3)$$

$$= \alpha(v_1 - v_2 + 2v_3) + \beta(u_1 - u_2 + 2u_3) = 0$$

∴ $(v_1, v_2, v_3)^T$ forms a subspace of \mathbb{R}^3

with $v_1 - v_2 + 2v_3 = 0$

c) All vector $(v_1, v_2)^T$ in \mathbb{R}^2 $v_1 \geq v_2$ a subspace of \mathbb{R}^2

sol: let $(v_1, v_2)^T$ & $(u_1, u_2)^T$
 $v_1 \geq v_2$ $u_1 \geq u_2$.

 $\alpha v + \beta u \Rightarrow (\alpha v_1 + \beta u_1, \alpha v_2 + \beta u_2)$

$$v_1 - v_2 \geq 0 \quad \alpha v_1 + \beta u_1 \geq \alpha v_2 + \beta u_2$$

$$u_1 - u_2 \geq 0 \quad \alpha(v_1 - v_2) \geq \beta(u_2 - u_1)$$

$$\underbrace{\alpha(v_1 - v_2)}_{v_1 - v_2 \geq 0} \geq \underbrace{-\beta(u_1 - u_2)}_{u_1 - u_2 \geq 0} \Rightarrow \text{satisfied.}$$

\therefore $(v_1, v_2)^T$ in \mathbb{R}^2 , $v_1 \geq v_2$ is a subspace of \mathbb{R}^2 .

d) $v = (v_1, v_2, v_3)^T \in \mathbb{R}^3$ with v_1, v_2, v_3 positive

let $(v_1, v_2, v_3)^T$ & $(u_1, u_2, u_3)^T$
 $v_1 > 0 \quad v_2 > 0 \quad v_3 > 0 \quad u_1 > 0 \quad u_2 > 0 \quad u_3 > 0$

$$(\alpha v + \beta u) \Rightarrow (\alpha v_1 + \beta u_1, \alpha v_2 + \beta u_2, \alpha v_3 + \beta u_3)$$

$$\alpha v_1 + \beta u_1 > 0 \quad \alpha v_2 + \beta u_2 > 0 \quad \alpha v_3 + \beta u_3 > 0$$

$$\begin{array}{lll} B_1 \subset & u_1 > 0 & \alpha v_1 + \beta u_1 > 0 \\ & u_1 > 0 & B_1 \subset v_2 > 0 \\ & u_1 > 0 & u_2 > 0 \\ & & B_1 \subset v_3 > 0 \\ & & u_3 > 0 \end{array}$$

$\therefore v = (v_1, v_2, v_3)^T \in \mathbb{R}^3$ with v_1, v_2, v_3 positive
is subspace of \mathbb{R}^3 .

e) $v = [v_1, v_2, v_3]^T \in \mathbb{R}^3$ with positive $v_1 - v_2 + v_3 = k$

$$\text{let } [v_1, v_2, v_3]^T \quad \& \quad [u_1, u_2, u_3]^T$$

$$v_1 - v_2 + v_3 = k \quad u_1 - u_2 + u_3 = k$$

$$\alpha v + \beta u \Rightarrow (\alpha v_1 + \beta u_1, \alpha v_2 + \beta u_2, \alpha v_3 + \beta u_3)$$

$$\alpha v_1 + \beta u_1 = \alpha v_2 - \beta u_2 + \alpha v_3 + \beta u_3$$

$$= \alpha(v_1 - v_2 + v_3) + \beta(u_1 - u_2 + u_3) = \beta k + \alpha k$$

$$\alpha v + \beta u = k(v + u) \neq k \cdot v$$

$v = (v_1, v_2, v_3)^T \in \mathbb{R}^3$ with $v_1 - v_2 + v_3 = k$ where k is a constant is not a subspace of \mathbb{R}^3 .

Q. 9:

Consider $xw = y$ where x is a 7×5 matrix.

a] Is x onto justify your answer?

$$[y]_{7 \times 1} = [x]_{7 \times 5} [w]_{5 \times 1}$$

$$N = 7$$

$$n+1 = 5$$

$N > n+1$ Since no of equations are more than no of unknowns.

∴ therefore there exists some y 's which doesn't have solution.
∴ x is not onto.

b) what is the maximum dimension the Range of x can have? Justify your answer.

sol: given x is a 7×5

$$y = xw \Rightarrow [y]_{7 \times 1} = [x]_{7 \times 5} [w]_{5 \times 1}$$

$x: \mathbb{R}^5 \rightarrow \mathbb{R}^7$ & x is not onto.

$$y = w_0 \cdot x_0 + w_1 x_1 + w_2 x_2 + w_3 x_3 + w_4 x_4$$

y can be expressed as unique combination

$$\text{of } S = \{x_0, x_1, x_2, x_3, x_4\}$$

∴ S forms the Basis

and max dimension of $R(x) = 5$

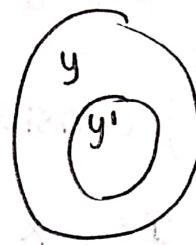
c) If it is possible to find w , what should be the rank of x ? In this case is x one to one? Justify your ans?

Sol: As x is not onto.

→ we have to find y' i.e projection of y on to $R(x)$

such that

$$P(y) = \min_{y' \in R(x)} \|y - y'\|$$



In this case matrix equation

$$xw = P(y)$$

→ The preimage of $P(y)$ onto the range space of x is to be found out.

→ Given if it is possible to w

→ For getting a unique preimage in this case

$$x : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^N$$

dimension of $R(x)$ should be $n+1$.

That is columns of x should be Linearly Independent

∴ x is one-one.

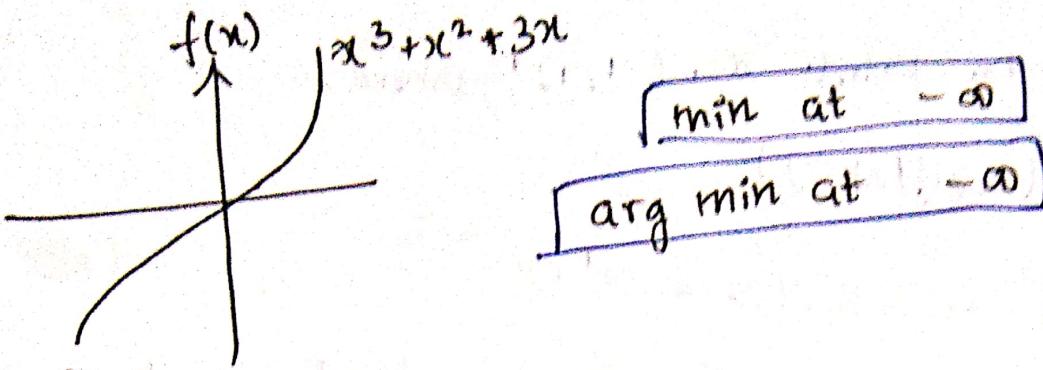
$$\text{Rank}(x) = 5.$$

10. Find $\arg \min f(x)$ and $\min f(x)$ where

a) $f(x) = x^3 + x^2 + 3x$, $x \in \mathbb{R}$.

Sol: $f'(x) = 3x^2 + 2x + 3 = 0$

$$x = -\frac{1}{3} \pm \frac{\sqrt{-2}}{2} i \Rightarrow \text{stationary point is complex}$$



b] $f(x) = \sin x + \cos x, x \in \mathbb{R}.$

$$f'(x) = \cos x - \sin x = 0 \Rightarrow \begin{aligned} \cos x &= \sin x \\ \tan x &= 1 \Rightarrow x = \frac{\pi}{4}. \end{aligned}$$

$$f''(x) = -\sin x + \cos x \quad x = [4n+1] \cdot \frac{\pi}{4}$$

$$\begin{aligned} -\sqrt{1^2 + 1^2} &\leq \sin x + \cos x \leq \sqrt{1^2 + 1^2} \\ -\sqrt{2} &\leq \sin x + \cos x \leq \sqrt{2} \end{aligned}$$

$$\min F(x) = -\sqrt{2} \text{ for } \arg \min F(x) = [4n+1] \cdot \frac{\pi}{4}.$$

c] $f(x) = \exp(\|x\|^2), x \in \mathbb{R}^2.$

$$f'(x) = \frac{d}{dx} \exp(\langle x, x \rangle) = e^{\|x\|^2} \cdot 2x = 0 \quad x=0$$

$$f''(x) = 2x e^{\|x\|^2} \cdot 2x + 2 \cdot e^{\|x\|^2}$$

$$f''(0) = 0 + 2 > 0$$

$\therefore x=(0,0)$ is $\arg \min (f(x))$

$$f(x)=1 \text{ is min.}$$

Q11

a] ∇f at the point $x = (-1, 1)^T$ where

$$f(x) = \sin(\cos(\|x\|^2))$$

$$\|x\|^2 = \langle x, x \rangle = x^T \cdot x$$

$$\frac{d}{dx} \|x\|^2 = \frac{d}{dx} \langle x, x \rangle = \frac{d}{dx} (x^T \cdot x) = \frac{d}{dx} x^T \cdot x + \frac{d}{dx} x^T \cdot x \\ = 2x.$$

$$\nabla f'(x) = \frac{d}{dx} [\sin(\cos(x^T \cdot x))]$$

$$= \cos(\cos(x^T \cdot x)) [-\sin(\cos(x^T \cdot x))] \cdot \frac{d}{dx} [x^T \cdot x]$$

$$\nabla f(x) = \cos(\cos(x^T \cdot x)) [-\sin(\cos(x^T \cdot x))] \cdot 2x.$$

$$x^T x = [-1, 1] \begin{bmatrix} -1 \\ 1 \end{bmatrix} = 1 + 1 = 2$$

$$x^T x = 2$$

$$= \underbrace{\cos(\cos(2)) [-\sin(2)]}_{-0.8316} \cdot 2 \cdot \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$= -1.663 \begin{bmatrix} -1 \\ 1 \end{bmatrix} \Rightarrow \boxed{\nabla f(x) = \begin{bmatrix} 1.663 \\ -1.663 \end{bmatrix}}$$

b) $\nabla_w f(w)$ where $f(w) = \exp(-\|xw - y\|^2)$

$$f(w) = \frac{1}{2} \|xw - y\|^2$$

$$\frac{d}{dw} f(w) = \nabla_w f(w) = \frac{1}{2} \cdot \langle xw - y, xw - y \rangle$$

$$= \frac{1}{2} \left[\langle (xw, xw) \rangle - \langle xw, y \rangle - \langle y, xw \rangle + \langle y, y \rangle \right]$$

$$J = \frac{1}{\alpha} \cdot \left[(xw)^T \cdot (xw) - (xw)^T y - y^T xw + y^T y \right]$$

$$\cancel{J} = \frac{1}{\alpha} \cdot [w^T x^T x w - w^T x^T y - y^T x w + y^T y]$$

$$\nabla_w J = \frac{1}{\alpha} \cdot \left[\frac{d}{dw} w^T (x^T x w) + \frac{d}{dw} (x^T x w)^T w - \frac{d}{dw} w^T (x^T y) \right. \\ \left. - \frac{d}{dw} (x^T y)^T \cdot w + \frac{d}{dw} (y^T y) \right]$$

$$= \frac{1}{\alpha} \cdot [x^T x w + x^T x w - x^T y - x^T y]$$

$$\nabla_w J = x^T x w - x^T y$$

$$f(w) = \exp(-\|xw - y\|^2) = \exp\left(-\frac{1}{\alpha} \cdot \|xw - y\|^2\right)$$

$$= \exp(-2J) = \exp$$

$$\nabla_w f(w) = \exp(-2J) \times -2 \times \nabla(J)$$

$$= \exp(-2J) \times -2 \cdot [x^T x w - x^T y]$$

$$\cancel{\nabla_w f(w)} \quad \boxed{\nabla_w f(w) = -2[f(w)]^2 \nabla_w J}$$

$$\text{where } J = \frac{1}{\alpha} \cdot \|xw - y\|^2$$

function:

There should not be multiple outputs of y

for single x , need not be every element

Element

is, set -B has

ONE-ONE:

a single preimage in set -A

ONTO: Every element has at least one preimage

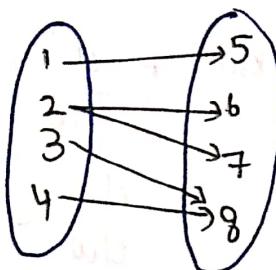
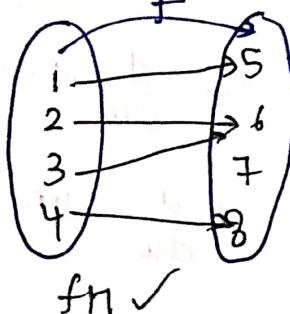
in A CODOMAIN = RANGE \rightarrow ONTO

CONCEPTS

Function:

There should not be multiple outputs of y for single x . Then given Relation is a function

Ex:



As single x i.e. 2 has multiple outputs in y i.e. (6, 7)

Note:

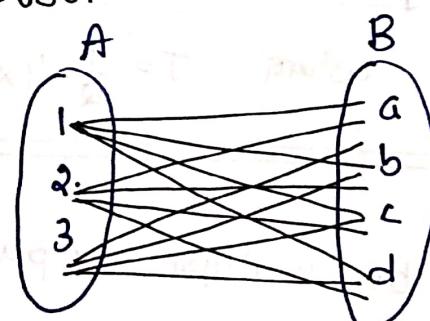
SETS \rightarrow Cartesian product \rightarrow Relation \rightarrow Function

SET: ~~Well~~ Well defined collection of objects

Cartesian Product: $A \times B$

$$\{(a,b) : a \in A, b \in B\}$$

All possible ordered pairs from A and B



$$A \times B = \{(1,a), (1,b), (1,c), (1,d), (2,a), (2,b), (2,c), (2,d), (3,a), (3,b), (3,c), (3,d)\}$$

Relation: $R: A \rightarrow B ; R \subseteq B$

Relation: $R: A \rightarrow B$

$R \subseteq B$

Ex: $A = \{1, 2, 3\}$ $B = \{a, b, c, d\}$
 R : all ordered pairs beginning with an odd no
 $\{(1, a), (1, b), (1, c), (1, d)\}$
 $(3, a), (3, b), (3, c), (3, d)\}$
∴ Relation happens to be a part of cartesian product

Domain: Ex: $A = \{1, 2, 3, 5\}$ $B = \{4, 6, 9\}$

set builder form $\rightarrow R: \{(x, y) : \text{diff b/w } x \text{ and } y \text{ is odd}$
 $x \in A \text{ and } y \in B\}$

sol:

Domain: set of all 1st elements in the ordered pair.

Domain: $\{1, 2, 3, 5\}$

Roaster form

$\{(1, 4), (1, 6), (2, 9)\}$
 $(3, 4), (3, 6), (5, 4)$
 $(5, 6)\}$

Range: set of all second elements in the order pair

Range: $\{4, 6, 9\}$

Co-Domain: codomain is always the second set

Codomain: B

Ex: $R: \{(x, y) : y = x + \frac{6}{x} \text{ where } x, y \in \mathbb{N} \text{ & } x < 6\}$

Domain: $\{1, 2, 3\}$

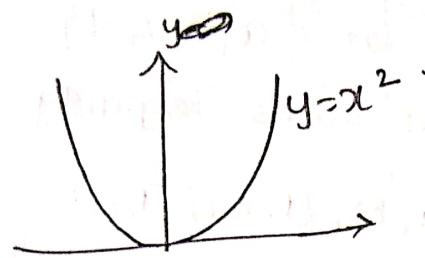
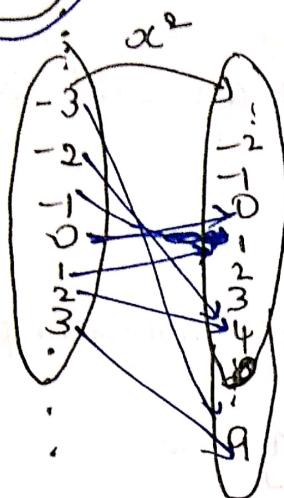
$\{(1, 7), (2, 5), (3, 5)\}$

Range: $\{4, 5, 7\}$

Codomain: \mathbb{N}

Q13

a) $f: \mathbb{R} \rightarrow \mathbb{R}; f(x) = x^2$



→ There are no multiple outputs of y for single x .
 $\therefore f(x) = x^2$ is a **fn.**

→ Domain = \mathbb{R}

→ Range = \mathbb{R}^+

→ Codomain = \mathbb{R}

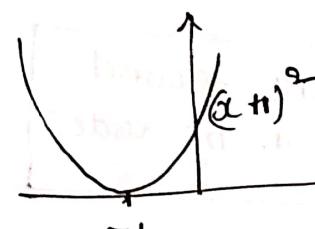
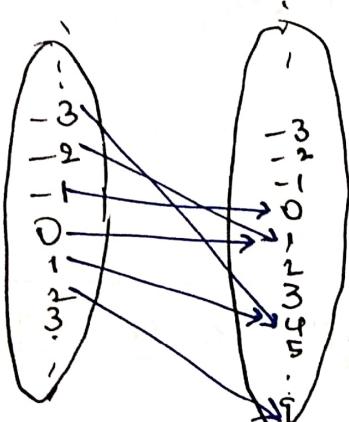
i.e. $\{x^2 | x \in \mathbb{R}, \forall x \in \mathbb{R}\}$

Not one-one **Not onto**

b) $f: \mathbb{R} \rightarrow \mathbb{R}; f(x) = (x+1)^2$

Not one-one

Not onto



→ There are no multiple outputs of y for single x .

$f(x) = (x+1)^2$ is a **function**

→ Domain = \mathbb{R}

Range = \mathbb{R}^+

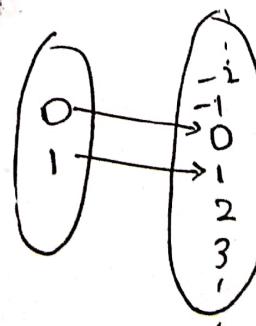
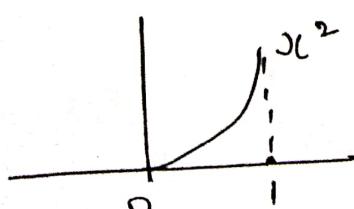
→ Codomain = \mathbb{R} .

i.e. $\{(x+1)^2 | x \in \mathbb{R}, \forall x \in \mathbb{R}\}$

c) $f: [0, 1] \rightarrow \mathbb{R}$

$f(x) = x^2$ **one-one**

Not onto



→ There are no multiple outputs for single x .

$f: [0, 1] \rightarrow \mathbb{R}$ is a function

domain: $[0, 1]$

Range: $[0, 1]$

Codomain: \mathbb{R}

- d] $f: [0,1] \rightarrow R$, where $f(x) = (x+3)^2$
-
- \rightarrow There are no multiple O.P.'s for single x
 $\therefore f(x) = (x+3)^2$ is a function
- \rightarrow Domain = $[0,1]$
 \rightarrow Range = $\{9, 16\}$
 \rightarrow codomain = R
- \rightarrow One-one
 \rightarrow Not onto

- e] $f: R \rightarrow R$, $f(x) = \sqrt{x}$.
-
- \rightarrow There are multiple y's for single x.
 $\therefore f(x) = \sqrt{x}$ is not a function.

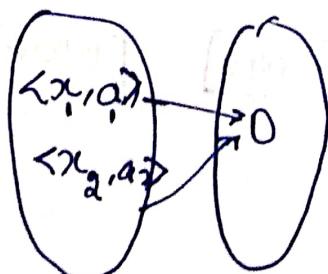
- f] $f: R^2 \rightarrow R$, $f(x) = \langle x, a \rangle$, $a \in R^2$.

Sol: $\langle x, a \rangle = x^T a$
 As there are no multiple outputs for a fixed i.p. It is a function.

Domain = R^2

Range = R

codomain = R



for different combinations of $\langle x, a \rangle$ will also lead to a single o.p. So not one-one

$\rightarrow \langle x, a \rangle$ every element in R is mapped onto

g) $f: \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ where $f(x, y) = \langle x, y \rangle$

As $f(x, y) = \langle x, y \rangle = x^T y$

There are no multiple o/p's for single i/p.
 $\therefore f(x, y) = \langle x, y \rangle$ is a fn.

Domain: $\mathbb{R}^2 \times \mathbb{R}^2$

Range: \mathbb{R}

Codomain: \mathbb{R} $\langle x, y \rangle = ab - ab = 0$

Ex:

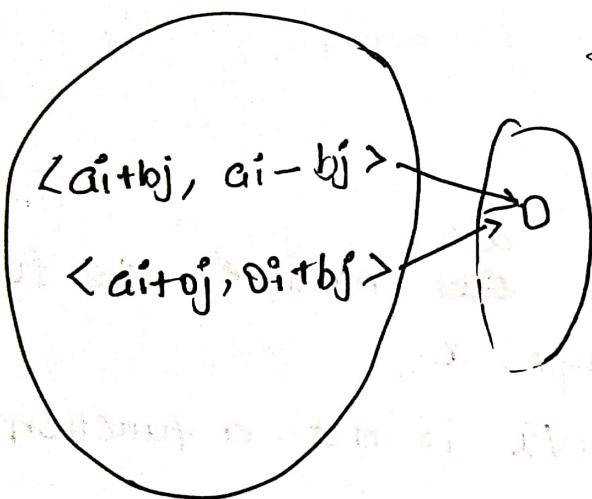
$$x = a_i + b_j \quad y = b_i - a_j$$

$$\langle x, y \rangle = 0$$

Not one-one.

As codomain = Range

[ONTO]



h) $f: \mathbb{R}^3 \times \mathbb{R} \rightarrow \mathbb{R}$ where $f(x) = \|x\|^2$

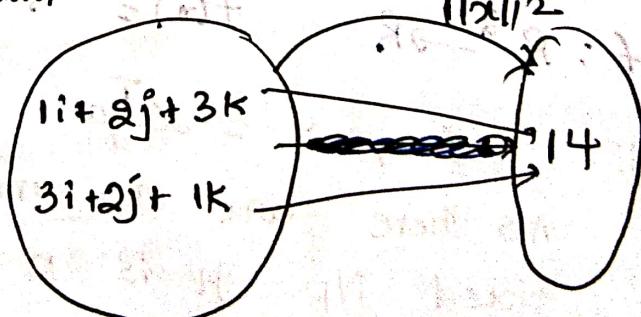
As $f(x) = \|x\|^2$

There are no multiple o/p's for fixed i/p.

Domain: \mathbb{R}^3

Range: \mathbb{R}^+

Codomain: \mathbb{R}



[Not one]

[Not onto]

\Rightarrow Codomain \neq Range.

$$i) f: \mathbb{R}^4 \times \mathbb{R}^4 \rightarrow \mathbb{R}, \quad f(x, y) = \exp(\|x - y\|^2)$$

Sol: $f(x, y) = \exp(\|x - y\|^2) \rightarrow$ function.

There are no multiple o/p's for single i/p.

Domain: $\mathbb{R}^4 \times \mathbb{R}^4$

Range: N \Rightarrow codomain \neq Range

Codomain: R

NOT ONTO

$$\text{Ex: } x_1 = (1, 2, 3) \quad y_1 = (3, 2, 1)$$

$$x_2 = (3, 2, 1) \quad y_1 = (1, 2, 3)$$

$$\|x_1 - y_1\|^2 = (2)^2 + (0)^2 + (-2)^2$$

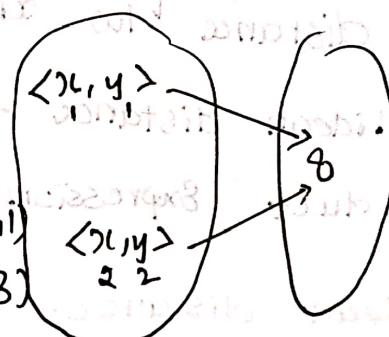
$$\begin{aligned} \|x_2 - y_2\|^2 &= (2)^2 + (0)^2 + (2)^2 \\ &= 8 \end{aligned}$$

$$(f(x_1) \neq f(x_2)) \Rightarrow \text{NOT ONTO}$$

$$\text{but } x_1 = (1, 2, 3) \quad y_1 = f(3, 2, 1)$$

$$y_1 = f(x_1)$$

$$\begin{aligned} x_2 &= (3, 2, 1) \\ y_2 &= f(x_2) \end{aligned}$$



NOT ONE-ONE

$$j) f: \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow M(2, \mathbb{R}), \quad f(x, y) = xy^T$$

where $M(2, \mathbb{R})$ is set of all 2×2 Real matrices

Sol: $f(x, y) = xy^T \rightarrow$ function.

There are no multiple o/p's for a single i/p

Domain = $(\mathbb{R}^2 \times \mathbb{R}^2)$

Range = codomain

Range = $M(2, \mathbb{R})$

ONTO function

Codomain = $M(2, \mathbb{R})$



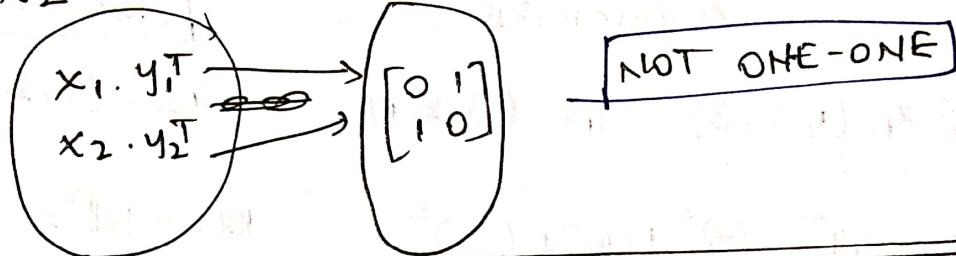
Ex:

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$x_1 \cdot y_1^T$

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

$x_2 \cdot y_2^T$



Q14: Find distance b/w $x_1^T = (1, 1, -2)$ and $x_2^T = (7, -3, 2)$
using Euclidean distance formulae, norm and
inner product expressions [write all relevant steps]

Sol: Euclidean distance.

$$x_1^T = (1, 1, -2) \quad x_2^T = (7, -3, 2)$$

$x_1 \quad x_2 \quad x_3$ $y_1 \quad y_2 \quad y_3$.

Euclidean distance $= \sqrt{(y_1 - x_1)^2 + (y_2 - x_2)^2 + (y_3 - x_3)^2}$

$$= \sqrt{(6)^2 + (-4)^2 + (4)^2}$$

$$= \sqrt{36 + 16 + 16} = \sqrt{68}$$

$$\text{norm} = \|x_2^T - x_1^T\|$$

$$x_2^T - x_1^T = (6, -4, 4)$$

$$\|x_2^T - x_1^T\| = \sqrt{6^2 + (-4)^2 + (4)^2} = \sqrt{68}$$

$$\begin{aligned}\text{Inner product} &= \langle x_1, x_2 \rangle = x_1^T x_2 \\ &= [1, 1, -2] \cdot \begin{bmatrix} 7 \\ -3 \\ 2 \end{bmatrix} \\ &= 7 - 3 - 4 = 0\end{aligned}$$

$$\langle x_1, x_2 \rangle = x_1^T \cdot x_2$$

$$\|x_1\|_2 = \sqrt{1^2 + 1^2 + (-2)^2} = \sqrt{6}$$

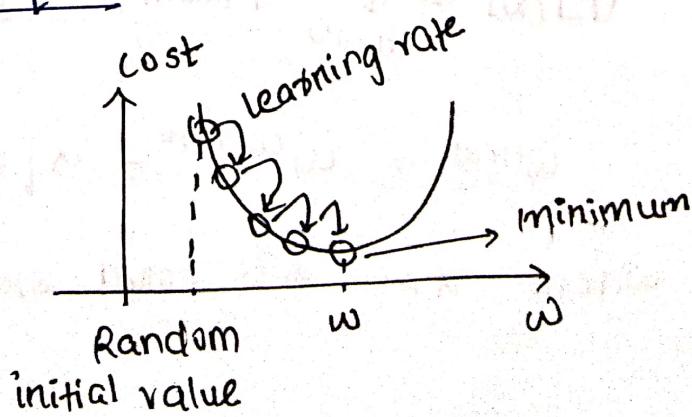
$$\|x_2\|_2 = \sqrt{(7)^2 + (-3)^2 + (2)^2} = \sqrt{62}$$

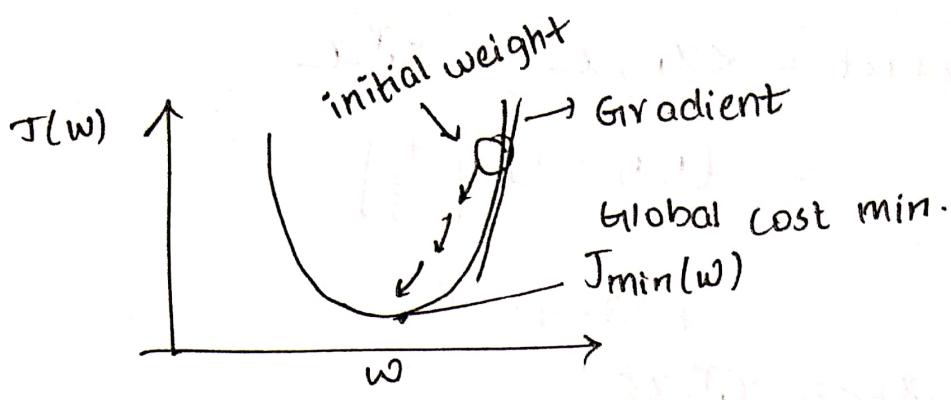
Q15: write a short notes on iterative techniques and gradient optimization approach.

- If the size of x is very large . Then it becomes computationally expensive to find x in direct method . so to avoid this we use iterative techniques to approach towards the solution & to find the parameter w .
- Initially By assuming ω calculating the error function by adjusting the hyper parameter or learning rate . we update the value of w in each iteration to reduce error.

Iterative Techniques:

Gradient descent:





→ The gradient vector can be interpreted as the "direction and the rate of fastest increase".

If the gradient of the function is non-zero at a point p , the direction of the gradient is in the direction in which function increases most quickly from p . and the magnitude of the gradient is the rate of increase in that direction.

→ The size of steps take to reach minimum or bottom is learning rate.

Gradient descent: If a real valued function $F(x)$ is defined and differentiable in a neighbourhood of point a , then $F(x)$ decreases fastest if one goes from a in the direction of negative gradient of F at a , $\nabla F(a)$

$$\nabla J(w) = \lim_{h \rightarrow 0} \frac{J(w+h) - J(w)}{h}$$

$$w_{\text{new}} = w^{\text{current}} - \alpha [\nabla J(w)]_{w^{\text{current}}}$$

where $\alpha > 0$ is called step length.

updation of w:

for applying Gradient descent, consider the following steps.

→ choose the initial value $w = (w_0, w_1, \dots, w_n)^T \in \mathbb{R}^{n+1}$.

Then repeatedly perform the update.

$$w := w - \alpha \cdot \nabla J$$

J is a function of w_0, w_1, \dots, w_n .

$$\nabla J = \left(\frac{\partial J}{\partial w_0}, \frac{\partial J}{\partial w_1}, \dots, \frac{\partial J}{\partial w_n} \right)^T$$

$$(w_0, w_1, w_2, \dots, w_n) := (w_0, w_1, \dots, w_n)^T - \alpha \cdot \left[\frac{\partial J}{\partial w_0}, \frac{\partial J}{\partial w_1}, \dots, \frac{\partial J}{\partial w_n} \right]^T$$

Batch Gradient Descent:

→ For updating the parameter, the algorithm looks at all every data point in the training set at every step and hence it is called Batch Gradient Descent.

→ In general, gradient descent does not guarantee a global minimum since J is a convex quadratic function, the algorithm converges to global minimum [assuming α is not too large]

Algorithm 1: Updation of w: Gradient Descent.

Initialize w

Iterate until convergence of

$$w := w + \alpha \cdot \sum_{i=1}^N (y_i - f(x_i)) \cdot x_i$$

Algorithm 2: updating of w : Gradient descent:

Initialize w .

Iterate until convergence {

$$w_j := w_j + \alpha \sum_{i=1}^N (y_i - f(x_i)) x_{ij}, j = 0, 1, \dots, n$$

}

Stochastic Gradient descent:

→ The online version of gradient descent

called stochastic gradient descent.

→ In contrast to batch gradient descent technique,

stochastic gradient process only takes one

training point at each step. Hence when

N becomes large, that is for large data sets,

stochastic gradient descent is more

computationally efficient than batch gradient descent.

Algorithm 3: updating w using Stochastic Gradient

descent.

Iterate until convergence {

for $i = 1, 2, \dots, N$ {

$$w := w + \alpha (y_i - f(x_i)) x_i$$

}

}

Algorithm 4: updating w using Stochastic Gradient descent.

Iterate until convergence {

for $i = 1, 2, \dots, N$ {

$$w_j := w_j + \alpha (y_i - f(x_i)) x_{ij}, j = 0, 1, 2, \dots, n$$

}

}

Q2:

$$(i) \{(1, -1), (2, 11), (4, 19)\} y = f(x_i, y_i)$$

$$(ii) x_1 = (2, 3, 2)^T, y_1 = 7$$

$$x_2 = (4, -5, 5)^T, y_2 = 28$$

$$x_3 = (-3, 7, -2)^T, y_3 = -19.$$

a) write $f(x_i)$, $i=1, 2, \dots, N$

$$f(x_i) = w_0 \cdot x_{i0} + w_1 \cdot x_{i1} + w_2 \cdot x_{i2} + \dots + w_n \cdot x_{in} = w^T \cdot x_i$$

$$\text{where } x_i = (1, x_{i1}, x_{i2}, \dots, x_{in})^T \in \mathbb{R}^{n+1}$$

$$w = (w_0, w_1, w_2, \dots, w_n)^T \in \mathbb{R}^{n+1}$$

Here

$w_0, w_1, w_2, \dots, w_n$ are the unknown parameters

$$(i) f(x_i) = w^T \cdot x_i$$

$$(ii) f(x_i) = w^T \cdot x_i$$

$$w_0 + w_1(1) = -1$$

$$w_0 + w_1(2) + w_2(3) + w_3(2) = 7$$

$$w_0 + w_1(2) = 11$$

$$w_0 + w_1(4) + w_2(-5) + w_3(5) = 28$$

$$w_0 + w_1(4) = 19$$

$$w_0 + w_1(-3) + w_2(7) + w_3(-2) = -19$$

b) Express data in $xw=y$

$$(i) \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \end{bmatrix} = \begin{bmatrix} -1 \\ 11 \\ 19 \end{bmatrix}$$

$$(ii) \begin{bmatrix} 1 & 2 & 3 & 2 \\ 1 & 4 & -5 & 5 \\ 1 & -3 & 7 & -2 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \\ w_2 \\ w_3 \end{bmatrix} = \begin{bmatrix} 7 \\ 28 \\ -19 \end{bmatrix}$$

c) set of vectors that spans Range of x .

$$(i) \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 4 \end{bmatrix} \Rightarrow R_2 = R_2 - R_1 \Rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ 0 & 3 \end{bmatrix} \Rightarrow R_3 = R_3 - 3R_2$$

$$\Rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$\dim = 2 \quad \{(1, 1), (1, 2), (1, 4)\}$$

∴ vectors

$$\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 4 \\ 1 \end{pmatrix}$$

spans range of X

$$(ii) \begin{bmatrix} 1 & 2 & 3 & 2 \\ 1 & 4 & -5 & 5 \\ 1 & -3 & 7 & -2 \end{bmatrix} \Rightarrow R_2 = R_2 - R_1 \Rightarrow \begin{bmatrix} 1 & 2 & 3 & 2 \\ 0 & 2 & -8 & 3 \\ 1 & -3 & 7 & -2 \end{bmatrix}$$

$$R_3 = 2R_3 + 5R_2 \Rightarrow \begin{bmatrix} 1 & 2 & 3 & 2 \\ 0 & 2 & -8 & 3 \\ 0 & 0 & -32 & 7 \end{bmatrix} \quad \boxed{\text{dim} = 3.}$$

As number of unknowns 4 (w_0, w_1, w_2, w_3)

But number of non-zero rows in Echelon form = 3

\therefore The set of vectors are L.D

$\therefore \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 4 \\ -3 \\ 7 \end{pmatrix}, \begin{pmatrix} 3 \\ -8 \\ 1 \\ -2 \end{pmatrix}, \begin{pmatrix} 2 \\ 3 \\ 5 \\ 1 \end{pmatrix} \right\}$ doesn't span Range of x .

$\Rightarrow \{(1, 2, 3, 2), (1, 4, -5, 5), (1, -3, 7, -2)\}$
doesn't span range of x .

$\Rightarrow \{(1, 0, 0, 0), (0, 1, 0, 0), (0, 0, 1, 0), (0, 0, 0, 1)\}$
spans Range(x) $\in \mathbb{R}^{4+1} = \mathbb{R}^4$.

d) Find the dimension of Range(x)

(i) $\boxed{\text{dim} = 2}$

\hookrightarrow No of non-zero

Rows in Echelon form = 2

(ii) $\boxed{\text{dim} = 3}$

\hookrightarrow No of non-zero

Rows in Echelon form = 3

$$e] \text{ (i)} \quad \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \end{bmatrix} = \begin{bmatrix} -1 \\ 11 \\ 19 \end{bmatrix}$$

$$\begin{array}{ccc} x \downarrow & w \uparrow & = \begin{array}{c} 4 \\ \uparrow \\ 3 \times 1 \end{array} & N=3 \\ 3 \times 2 & 2 \times 1 & \downarrow & n+1=2 \\ N \times (n+1) & (n+1) \times 1 & N \times 1 & \end{array}$$

$\therefore N > (n+1) \Rightarrow$ over determined system

Apply Least square Approximation

i.e Right pseudo inverse

To find w .

$$w = (x^T x)^{-1} x^T y \rightarrow \text{Right pseudo inverse}$$

By least square Approximation

$$x^T x = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 4 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 4 \end{bmatrix} = \begin{bmatrix} 3 & 7 \\ 7 & 21 \end{bmatrix}$$

$2 \times 3 \quad 3 \times 2 \quad 2 \times 2$

$$(x^T x)^{-1} = \frac{1}{14} \cdot \begin{bmatrix} 21 & -7 \\ -7 & 3 \end{bmatrix}$$

$$(x^T x)^{-1} \cdot x^T = \frac{1}{14} \cdot \begin{bmatrix} 21 & -7 \\ -7 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 4 \end{bmatrix}$$

$$= \frac{1}{14} \begin{bmatrix} 14 & 7 & -7 \\ -4 & -1 & 5 \end{bmatrix}$$

$$(x^T x)^{-1} \cdot x^T \cdot y = \frac{1}{14} \cdot \begin{bmatrix} 14 & 7 & -7 \\ -4 & -1 & 5 \end{bmatrix} \begin{bmatrix} -1 \\ 11 \\ 19 \end{bmatrix}$$

$$= \frac{1}{14} \cdot \begin{bmatrix} -70 \\ 88 \end{bmatrix} = \begin{bmatrix} -5 \\ 6.28 \end{bmatrix}$$

$$w = \begin{bmatrix} -5 \\ 6.28 \end{bmatrix}$$

$$w_0 = -5$$

$$w_1 = 6.28$$

Hyperplane:

Regression :

$$y_i = w^T \cdot x_i + w_0$$

Classification:

Decision boundary

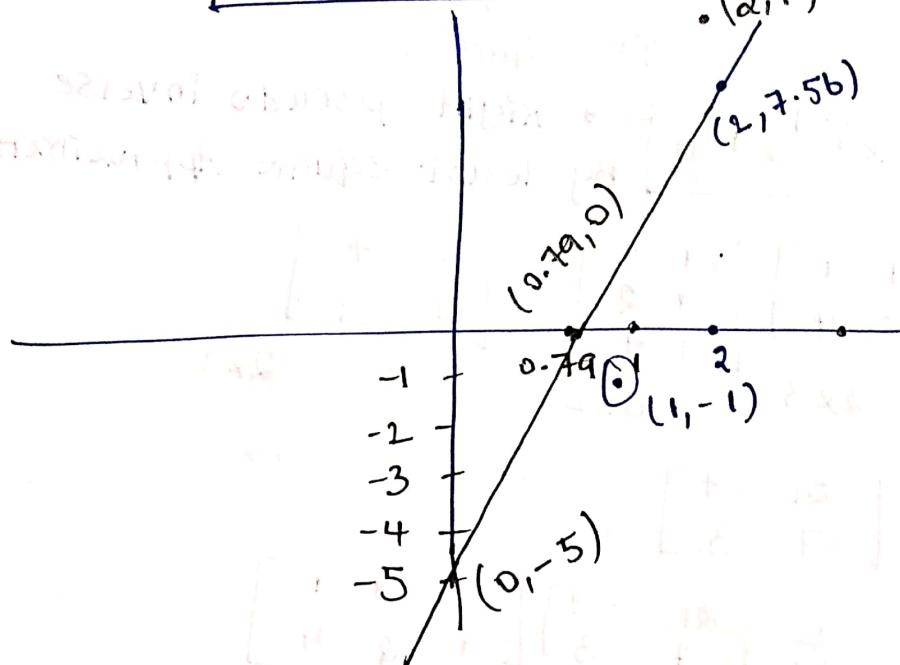
$$w^T \cdot x_i + w_0 = 0$$

For the given Regression problem

Hyperplane

$$y_i = w^T \cdot x_i + w_0 \quad \cdot (3, 19)$$

$$y_i = 6 \cdot 28 \cdot x_i - 5 \rightarrow \text{hyper plane}$$



f) Report values of parameter

$$w_0 = -5$$

$$w_1 = 6 \cdot 28$$

g) projection of y onto Range of x

$$y' = x \cdot w$$

$$= \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} -5 \\ 6 \cdot 28 \end{bmatrix} =$$

$$y' = \begin{bmatrix} 1 \cdot 28 \\ 7.56 \\ 20.12 \end{bmatrix} = w \cdot x$$

projection of
y onto
Range of x.

Q5:

consider the following set of points

$$\{(x_i, y_i) : \{(-1, 0), (2, 2), (4, 3)\}\}$$

Is that possible to find hyperplane that generates data justify your answer.

Sol:

$$X = \begin{bmatrix} -1 & -1 \\ 1 & 2 \\ 4 & 3 \end{bmatrix} \quad w = (X^T X)^{-1} X^T y$$

3×2

$$N=3 \quad (n+1)=2$$

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

$N > (n+1) \Rightarrow$ overdetermined system

solved by least square approximation

[Right pseudo inverse]

$$X^T X = \begin{bmatrix} 1 & 1 & 4 \\ -1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 4 \\ -1 & 2 & 3 \end{bmatrix}^{-1} \begin{bmatrix} 1 & -1 \\ 2 & 3 \end{bmatrix}$$

$$X^T X = \begin{bmatrix} 3 & 5 \\ 5 & 21 \end{bmatrix}$$

$$(X^T X)^{-1} = \frac{1}{38} \begin{bmatrix} 21 & -5 \\ -5 & 3 \end{bmatrix}$$

$$(X^T X)^{-1} \cdot X^T = \frac{1}{38} \begin{bmatrix} 21 & -5 \\ -5 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ -1 & 2 & 4 \end{bmatrix}$$

$$= \frac{1}{38} \begin{bmatrix} 26 & 11 & 1 \\ -8 & 1 & 7 \end{bmatrix}$$

$$w = (x^T x)^{-1} x^T y$$

$$= \frac{1}{38} \begin{bmatrix} 26 & 11 & 1 \\ -8 & 7 & 13 \end{bmatrix} \begin{bmatrix} 0 \\ 2 \\ 3 \end{bmatrix}$$

$$w = \frac{1}{38} \begin{bmatrix} 25 \\ 23 \end{bmatrix} \Rightarrow w = \begin{bmatrix} 0.657 \\ 0.605 \end{bmatrix}$$

Hyperplane $\rightarrow y = w^T x_i + w_0$

for given Regression pblm

$$y = 0.605 x_1 + 0.657$$

Hyperplane for the given pblm

Q 04:

Consider the following data

$$w_0 + 2w_1 - w_2 = 1$$

$$w_0 - w_1 + 2w_2 = 1$$

$$w_0 + 3w_1 + 4w_2 = 1$$

a) Express $y = [1, 1, 1]^T$ as a linear combination

of vectors. Are those vectors linearly independent?

Is it possible to find the unknown

unknown parameters of eqn? Justify your

answer?

Sol:

$$\begin{bmatrix} 1 & 2 & -1 \\ 1 & -1 & 2 \\ 1 & 3c & c \end{bmatrix} \rightarrow R_2 = R_2 - R_1$$

$$R_3 = R_3 - R_1$$

$$\begin{array}{rcl} 1 & -1 & 2 \\ -1 & -2 & -1 \\ \hline R_2 \rightarrow 0 & -3 & 3 \end{array} \quad \begin{array}{rcl} 1 & 3c & c \\ -1 & -2 & -1 \\ \hline R_3 \rightarrow 0 & 3c-2 & c+1 \end{array} \Rightarrow \begin{bmatrix} 1 & 2 & -1 \\ 0 & -3 & 3 \\ 0 & 3c-2 & c+1 \end{bmatrix}$$

$$R_3 = 3 \times R_3 + (3c-2)R_2$$

$$0 \quad 0 \quad 3[3c-2] \quad 3c+3$$

$$0 \quad -3[3c-2] \quad 9c-6$$

$$\begin{bmatrix} 1 & 2 & -1 \\ 0 & -3 & 3 \\ 0 & 0 & 12c-3 \end{bmatrix}$$

Conclusion: given vectors are **linearly independent** for $c \neq 1/4$

Linearly Independent

→ If the given

vectors are

Linearly independent

i.e $c \neq 1/4$

Then we **can't find**
the unknown parameters

$[w_0, w_1, w_2]$

because c is unknown

Linearly Dependent

then $c = 1/4$.

Echelon form of x is.

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

→ There exists more
than one solution

→ $N < (M+1)$

under determined system

→ unknown parameters

can be found by

Lagrangian formulation

$$w = x^T (x x^T)^{-1} \cdot y$$

b) Find:

No of attributes = 2

No of data points = 3

$$x_1 = (2, -1), y_1 = 1$$

$$x_2 = (-1, 2), y_2 = 1$$

$$x_3 = (3c, c), y_3 = 1$$

c) Find the hyper plane that generates the data using least square regression, by applying direct method.

Sol: Linearly Independent:

→ If the given vectors are

Linearly independent i.e $c \neq 1/4$

→ As the c is unknown

we can't find the unknown parameters

$$[w_0, w_1, w_2]^T$$

→ So hyperplane is not possible

if $[w_0, w_1, w_2]^T$ are unknown

b/c hyperplane

$$y_i = w^T x_i + w_0$$

Linearly Dependent:

→ If the given vectors are

Linearly Dependent

Then $c = 1/4$

$$\left[\begin{array}{ccc|c} 1 & 2 & -1 & 1 \\ 1 & -1 & 2 & 1 \\ 1 & 3c & c & 1 \end{array} \right] \Rightarrow R_2 = R_2 - R_1 \quad R_3 = R_3 - R_1 \Rightarrow \left[\begin{array}{ccc|c} 1 & 2 & -1 & 1 \\ 0 & -3 & 3 & 0 \\ 0 & 3c-2 & c+1 & 0 \end{array} \right]$$

Augmented matrix

$$\left[\begin{array}{ccc|c} 1 & 2 & -1 & 1 \\ 0 & -3 & 3 & 0 \\ 0 & 0 & 12c-3 & 0 \end{array} \right] \leftarrow R_3 = 3R_3 + (3c-2)R_2$$

Linearly dependent

$$12c-3=0 \Rightarrow c=1/4.$$

$$\left[\begin{array}{ccc} 1 & 2 & -1 \\ 0 & -3 & 3 \end{array} \right] \left[\begin{array}{c} w_0 \\ w_1 \\ w_2 \end{array} \right] = \left[\begin{array}{c} 1 \\ 0 \end{array} \right]$$

$$n+1=3$$

$$N=2$$

$$(n+1) > N \Rightarrow$$

under determined system

↳ solved by

Lagrangian formulation

(§1)

Minimum Normed solution

i.e. left pseudo inverse.

→ under determined

cannot be solved

by least square approximation

Lagrangian formulation (§1) Minimum Normed solution:

left pseudo inverse

$$w = X^T (X X^T)^{-1} y$$

$$X = \begin{pmatrix} 1 & 2 & -1 \\ 0 & -3 & 3 \end{pmatrix} \quad y = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$x = \begin{bmatrix} 1 & 2 & -1 \\ 0 & -3 & 3 \end{bmatrix} \quad y = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$w = x^T (x x^T)^{-1} \cdot y$$

$$x x^T = \begin{bmatrix} 1 & 2 & -1 \\ 0 & -3 & 3 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & -3 \\ 0 & -1 & 3 \end{bmatrix} = \begin{bmatrix} 6 & -9 \\ -9 & 18 \end{bmatrix}$$

$$(x x^T)^{-1} = \frac{1}{27} \begin{bmatrix} 18 & 9 \\ 9 & 6 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 6 & -3 \\ 0 & -1 & 3 \end{bmatrix}$$

$$x^T (x x^T)^{-1} = \frac{1}{27} \cdot \begin{bmatrix} 1 & 2 & 0 \\ 0 & -3 & 3 \end{bmatrix} \begin{bmatrix} 18 & 9 \\ 9 & 6 \end{bmatrix} = \begin{bmatrix} 18 & 9 \\ 9 & 6 \end{bmatrix}$$

$$x^T (x x^T)^{-1} = \frac{1}{27} \begin{bmatrix} 18 & 9 \\ 9 & 0 \\ 9 & 9 \end{bmatrix}$$

$$x^T (x x^T)^{-1} y = \frac{1}{27} \begin{bmatrix} 18 & 9 \\ 9 & 0 \\ 9 & 9 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\frac{1}{27} \begin{bmatrix} 18 \\ 9 \\ 9 \end{bmatrix} = \begin{bmatrix} 2/3 \\ 1/3 \\ 1/3 \end{bmatrix} = w$$

$$w_0 = 2/3 \quad w_1 = 1/3 \quad w_2 = 1/3$$

$$\text{Hyperplane} \Rightarrow y_i = w^T x_i + w_0$$

$$y_i = w_1 x_1 + w_2 x_2 + w_0$$

using

$$y_i = 1/3 x_1 + 1/3 x_2 + 2/3$$

Hyper plane

Legrangian Formulation

Q12

consider the following Regression problem

$$\{(\mathbf{x}_i, y_i)\} \text{ where } i=1, 2, 3, 4$$

$$\text{where } \mathbf{x}_1 = (-1, 1)^T, \mathbf{x}_2 = (0, 2)^T, \mathbf{x}_3 = (1, 1)^T, \mathbf{x}_4 = (2, 1)^T$$

$$\text{and } y_1 = -1, y_2 = 1, y_3 = 2, y_4 = 0. \text{ let } \mathbf{x}\mathbf{w} = y$$

If it is possible to find hyperplane that generates data

Sol:

$$\begin{bmatrix} 1 & -1 & 1 \\ 1 & 0 & 2 \\ 1 & 1 & 1 \\ 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} w_0 \\ w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \\ 2 \\ 0 \end{bmatrix}$$

$$N=4 \quad (n+1) = 3$$

$N > (n+1) \Rightarrow$ over determined system

↳ parameter can be found by least square regression

Right pseudo inverse $\mathbf{w} = (\mathbf{x}^T \mathbf{x})^{-1} \cdot \mathbf{x}^T \cdot \mathbf{y}$

Hyper plane $y_i = \mathbf{w}^T \mathbf{x}_i + w_0$

a)

$$\mathbf{x}(\text{domain}) = \mathbb{R}^3$$

$$\mathbf{x}(\text{Range}) = \mathbb{R}^4$$

$$\mathbf{x}: \mathbf{w} \rightarrow \mathbf{y}$$

$$\mathbf{x}: \mathbb{R}^3 \rightarrow \mathbb{R}^4$$

\rightarrow As $N > (n+1) \Rightarrow$ there exists some y 's

which doesn't have solution

x is not onto

$$s = \left\{ \left(\begin{array}{|c|} \hline 1 \\ \hline \end{array} \right), \left(\begin{array}{|c|} \hline -1 \\ \hline 0 \\ \hline 2 \\ \hline \end{array} \right), \left(\begin{array}{|c|} \hline 1 \\ \hline 1 \\ \hline \end{array} \right), \left(\begin{array}{|c|} \hline 1 \\ \hline 2 \\ \hline \end{array} \right) \right\}$$

\rightarrow As the given vectors uniquely represented

x is one-one

b) dimension of Range(\mathbf{x}) = $n+1 = 3$

x is uniquely represented $\rightarrow x$ is one-one

Basis and dimension $R(x) = n+1$

(iii) Basis of $R(x) = n+1 = 3$

$$S = \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right\}$$

(iv) Write the expression to find the projection of y onto the Range of X

$$y' = w x$$

$N > (n+1) \Rightarrow$ overdetermined system

↳ least square approx

Right pseudo inverse

$$w = (x^T x)^{-1} x^T y$$