

L.3 a)

$$\text{sign } y(t) + w^T(t) \cdot x(t) < 0$$

Hint:  $x(t)$  is misclassified by  $w(t)$

$\therefore$  since  $x(t)$  is misclassified, that means.

$$\text{sign}(w^T(t) \cdot x(t)) = -1 \quad \& \quad y(t) = +1$$

$$\text{and } \text{sign}(w^T(t) \cdot x(t)) = +1 \quad \& \quad y(t) = -1$$

for  $\text{sign}(w^T(t) \cdot x(t))$  to be  $-1$ ,

we can refer  $\text{sign}(n) = \begin{cases} -1 & \text{if } n < 0 \\ 0 & \text{if } n = 0 \\ 1 & \text{if } n > 0 \end{cases}$

$$\therefore \text{sign}(n) = -1 \quad \therefore n < 0.$$

$$\therefore \text{sign}(w^T(t) \cdot x(t)) = -1$$

which implies,  $w^T(t) \cdot x(t) < 0.$

i. As we already have  $y(x) = +1$  for  $\text{sign}(w^T(t) \cdot x(t)) = +1$

Then  $y(t) = w^T(t) \cdot x(t) = (+1) * (\text{sum number less than } 0)$

$$\Rightarrow (+1) * (m < 0) \quad \text{where } m = \text{some number} < 2020.$$

$$\therefore (+1) * (m < 0) = \text{output} < 0.$$

Hence proved,  $y(t) w^T(t) x(t) < 0$

same for the other case

(b) Show that  $y(t) w^T(t+1) x(t) > y(t) w^T(t) x(t)$

Hint : Use (1.3).

∴ It is said that with  $w(t+1)$  our  $n(t)$  will be correctly classified.

∴  $y(t) = +1$  then ~~sign~~ sign  $(w^T(t+1) n(t)) = +1$   
as it is correctly classified.  
 $(n(t))$

∴ for sign  $(w^T(t+1) n(t))$  to be  $=+1$

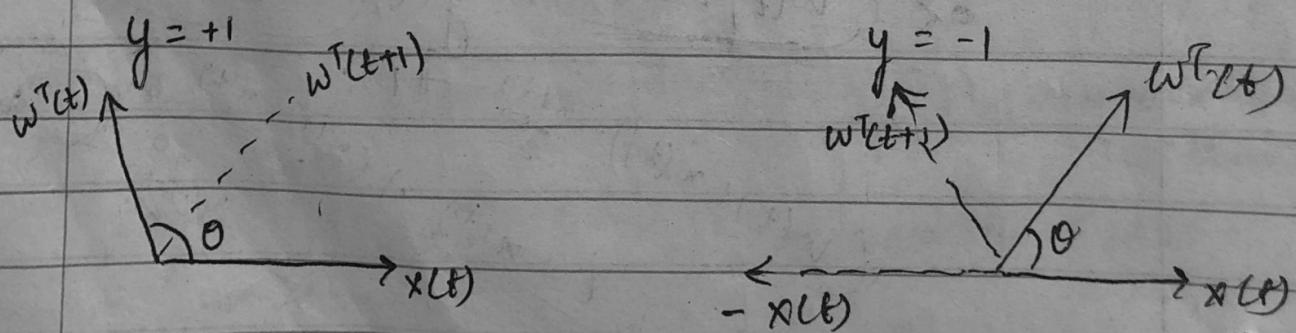
$w^T(t+1) n(t) > 0$ , from  $\text{sign}(n) = \begin{cases} -1 & \text{if } n < 0 \\ 0 & \text{if } n = 0 \\ 1 & \text{if } n > 0 \end{cases}$

∴ As  $y(t) = +1$   $\therefore [y(t) + w^T(t+1) * x(t) > 0]$

As we have proved if 1.3(a)  $y(t) w^T(t) x(t) < 0$

already which means,

$[y(t) + w^T(t+1) * x(t) > y(t) w^T(t) x(t)]$



## Problem 1.6

a) Recommending a book to a user in an online bookstore :-

→ Supervised learning → As we have to have labelled purchase history of user to recommend the book.

→ We know what user has purchased in past & hence for every feature set ' $x$ ' we have ' $y$ ' as output whether "purchased" or not.

∴ Training datset :-

book author, no. of pages, type of book like its genre, ebook or hardcopy, graphical content or not, etc.

b) Playing a tic-tac-toe! - (Reinforcement learning)

- We do not have output of a player's win/loss given few steps.

→ The model has to learn whether the performed set of steps lead to a winning output or losing one, hence reinforcement learning.

Training datset:- positions of values for 'X' or 'O'

i.e first\_box, second\_box, third\_box,  
fourth\_box, fifth\_box, sixth\_box, seventh\_box,

→  $\text{haveWon} = \{0, 1\} = 0$  for winning  
and '1' for losing.

each box will have  $\{\text{'X'}, \text{'O'}\}$ , winningStep? where  
winning step = boolean i.e. this step performed winning  
step or losing one.

### c) Categorizing movies into different types:-

• It can be supervised or unsupervised based on the data we have,

∴ Supervised → If we have attributes in dataset for which we have 'movie type' as output attribute for each input vector.  
Unsupervised if, if do not have labelled output data for movie category then this becomes clustering problem which will be Unsupervised.

dataset:-  
→ movie\_id, isAdult, isHorror, isAction, isRomantic,  
isDrama, isComedy, budget, actors, director,  
actress, length\_of\_movie, rating, language,  
runtime etc.

d) Learning to play music :-

problem statement :- (1, 0) = incorrect, (0, 1) = correct, (1, 1) = redundancy

Reinforcement Learning → As we have to learn by checking if output is proper music

such as note prioritization, X adhering to music notes or how.

guitar bending note sixth and second note priorities

Dataset :- Bandwidth, frequency, bass, contrast, etc.  
audio attributes like chroma, Tonnetz, roll off  
etc. (egpt too difficult after solving prioritization)

e) Credit limit :- Deciding the maximum allowed debt for each bank customer :-

task of statistics over the bankrupcy

→ Supervised :- We have to have history of the customers so the bank has to based on that only we can predict the future customer's debt.

dataset :- Age, income, members in family, education, number of creditcards, creditscores, financials, no. of loans, assetvalue etc.

rewards, roles, features, planned, predicted, payment, partner, sibling, first, last, address, its signature

Exercise 1.8 :-

Given,  $\mu = 0.9 \Rightarrow$  probability of red marble in bin.

$$\therefore P(\text{red}) = 0.9$$

Let's say  $N \Rightarrow$  No of samples = 10.

$$\text{To find: } P(D \leq 0.1) \Rightarrow P = \frac{\text{No. of red}}{\text{Total sample}} = \frac{\text{red}}{N}$$

$$P(\text{marble} \neq \text{red}) = 1 - 0.9 = 0.1$$

$$P(\text{marble} = \text{red}) = 0.9$$

$$\therefore P(D \leq 0.1) = P\left(\frac{\text{red}}{N} \leq 0.1\right) = P(\text{red} \leq 0.1 \times N)$$

$$= P(\text{red} \leq 1) = P(\text{red} = 0) + P(\text{red} = 1)$$

$\therefore P(\text{red} = 0) =$  each sample has probability of marble being not red.

$$\therefore P(\text{not red}) = P(\text{not red})^10 = 1 \times 10^{-10}$$

$$\therefore P(\text{red} = 1) = 0.9 \times 0.9$$

$$= (0.9)^{10} \times (0.1)^0$$

$$= 9 \times 10^{-9}$$

$$\therefore P(D \leq 0.1) = P(\text{red} = 0) + P(\text{red} = 1)$$

$$= 1 \times 10^{-10} + 9 \times 10^{-9}$$

$$= 0.1 \times 10^{-9} + 9 \times 10^{-9} = \underline{\underline{9.1 \times 10^{-9}}}$$

## Exercise 1-9

$\mu = 0.9$ . We have to bound  $P(0 \leq 0.1)$ .

c. By Hoeffding Inequality,

$$P[|D - \mu| > \epsilon] \leq 2e^{-\frac{\epsilon^2 N}{2}} \quad \text{for any } \epsilon > 0$$

$$\therefore |D - \mu| > \varepsilon$$

: Squaring 2 sides,  $(D-pe)^2 > (E)^2$

$$(0.1 - 0.9)^2 \rightarrow (6)$$

So we can say for  $\leq 2e^{-2\epsilon^2 N}$ ,  $\epsilon = 6.8$ .

$$\Rightarrow P[|U - \mu| > \varepsilon] \leq 2e^{-2(0.8)^2(10)} \\ \leq (2)e^{-2(6.4)} \\ \leq 5.52 \times 10^{-6}$$

$$\Rightarrow p(0 \leq \theta \cdot 1) < p[|\theta - \mu| > \epsilon] \leq 5.82 \times 10^{-6}$$

$$\rightarrow \text{prob} = 9.1 \times 10^{-9} \leq p[|D - M| \geq 8] \leq 3.52 \times 10^{-6}$$

$\leftarrow f(+\infty) (-1) \cdot (p+q) \cdot (-)$

$$(1 + \log p)q + (0 + \log n)q + (1 - 0 \geq 0)q$$

$$P = \exp(-\frac{1}{2} \chi^2)$$

$$P_{\text{S1} \times \text{L.P}} = P_{\text{S1} \times \text{P}} + P_{\text{S1} \times \text{L.P}}$$

problem 1.2 (part 1) if will work over the whole

$$h(x) = \text{sign}(w^T x)$$

where  $w = [w_0, w_1, w_2]^T$

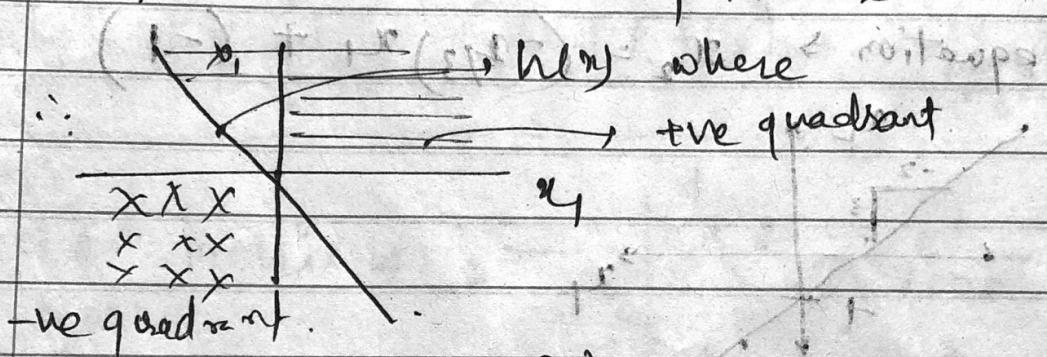
$$x = (1, [x_1, x_2])^T$$

(b) for x, sign

As we know,  $\text{sign}(x) = \begin{cases} + & \text{where } x > 0 \\ 0 & x = 0 \\ -1 & x < 0 \end{cases}$

$\therefore h(x) = +1$  which means points are in +ve quadrants.

$h(x) = -1$  which means points are in -ve quadrants.



Also,  $h(x) = \text{sign}(w^T x) = \text{sign}(w_0 x_0 + w_1 x_1 + w_2 x_2)$

when  $w^T x = 0$  we say all points are not line

(1) &  $w$  is perpendicular to  $\vec{x}$

$\therefore w_0 x_0 = \text{constant}$ , we can express  $w_0 x_0 = k$  ~~or~~ some constant,

$$\begin{aligned} x_2 &= (-w_1)/(w_2)x_1 + (-w_0/w_2) \\ &= (-w_1/w_2)x_1 + c \end{aligned}$$

This is a equation of line.

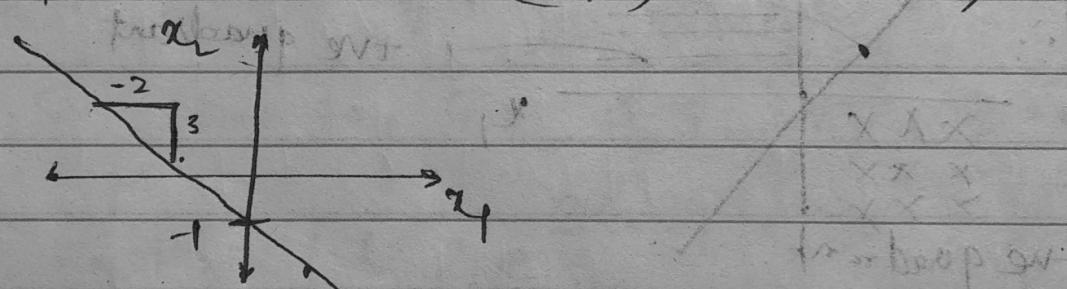
Now, we have given this, is,  $x_2 = ax_1 + b$ ! (read off)  
 where  $a = \text{slope}$  &  $b = \text{intercept}$  (w)

$$\therefore x_2 = -\frac{\omega_1}{\omega_2}x_1 + (-\omega_0) \rightarrow$$

$$\therefore \begin{cases} a = -\frac{\omega_1}{\omega_2} \\ b = -\omega_0 \text{ as } x_0 = 1 \end{cases} \quad \text{Ans, part a)}$$

part (b) v.t. i)  $\omega = [1, 2, 3]^T$

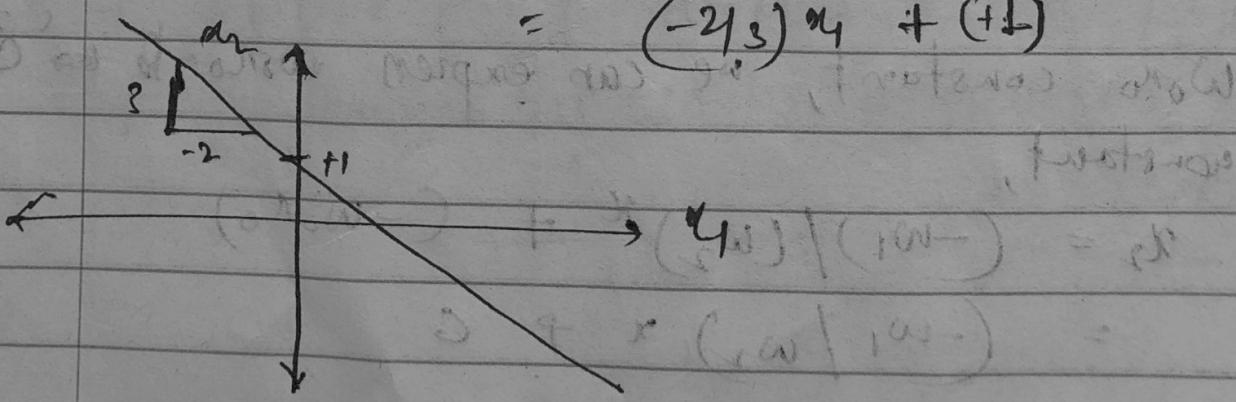
$$\therefore \text{equation} \Rightarrow x_2 = \left(-\frac{2}{3}\right)x_1 + (-1)$$



$$ii) \omega = [-1, -2, -3]^T$$

$$\therefore \text{equation} \Rightarrow x_2 = \left(-\frac{-1}{-3}\right)x_1 + (-(-1))$$

$$= \left(\frac{2}{3}\right)x_1 + (+1)$$



problem 1.11:

a) Supermarket:  $\therefore$  matrix  $\Rightarrow$

$$\begin{array}{c|cc} & +1 & -1 \\ \hline +1 & 0 & 1 \\ -1 & 10 & 0 \end{array}$$

$$\therefore E_{in}(g) = \frac{1}{N} \sum_{n=1}^N \delta(h(x_n), f(x_n))$$

$$\therefore E_{in}(g) = \frac{1}{N} \sum_{n=1}^N [y_n \neq g(x_n)]$$

$$E_{in}(g) = \frac{1}{N} \left[ \left( \sum_{y_n=+1} (g(x_n) \neq +1) \right) \times 10 + \left( \sum_{y_n=-1} (g(x_n) \neq -1) \right) \times 1 \right]$$

b) for CIA: matrix  $\Rightarrow$

$$\begin{array}{c|cc} & +1 & -1 \\ \hline +1 & 0 & 1000 \\ -1 & 1 & 0 \end{array}$$

$$E_{in}(g) = \frac{1}{N} \sum_{n=1}^N [y_n \neq g(x_n)]$$

$$E_{in}(g) = \frac{1}{N} \left[ \left( \sum_{y_n=+1} (g(x_n) \neq +1) \right) \times 1 + \left( \sum_{y_n=-1} (g(x_n) \neq -1) \right) \times 1000 \right]$$