

① Exercise 1.3 a)

1.3 a) $y(t) \cdot w^T(t) \cdot x(t) < 0$

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

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

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

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

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

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

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

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

\therefore Hence proved, $y(t) w^T(t) x(t) < 0$] same for the other case

Exercise 1.3(b)

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

Point : Use (1.3).

\therefore (1.1) said that with $w(t+1)$ our $x(t)$ will be correctly classified.

$\therefore y(t) = +1$ then $\text{sign}(w^T(t+1) x(t)) = +1$
as it is correctly classified.
 $x(t)$

$\therefore \text{sign}(w^T(t+1) x(t))$ to be $=+1$

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

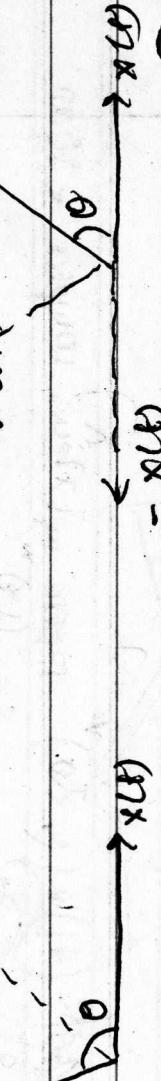
$$\therefore \text{As } y(t) = +1 \quad \therefore \boxed{y(t) * w^T(t+1) * x(t) > 0}$$

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

already which means,

$$\boxed{y(t) * w^T(t+1) * x(t) > y(t) w^T(t) x(t)}$$

$$y = +1 \quad w^T(t+1) \nearrow \begin{matrix} w^T(t+1) \\ x(t) \end{matrix}$$



Exercise 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 dat set :-

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 dat set:- positions of values for 'X' or 'O'

i.e first_box, second_box, third_box,
fourth_box, fifth_box, sixth_box, seventh_box,

→ State value of position (b)
eighth_box, ninth_box, haveWon = {0, 1} = 0 for winning
and 1 for losing.

and each box will have { 'X', '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 then.

dataset:-

→ movie_id, isAdult, isHorror, isAction, isRomantic,
isDrama, isComedy, budget, actors, director,
actors, length_of_movie, rating, language,
runtime etc.

d) Learning to play music :-

Reinforcement Learning → As we have to learn by checking if output is proper music
and if it's adhering to music notes or how.

Dataset :- Bandwidth, frequency, bass, contrast,
audio attributes like chroma, Tonnetz, roll off
etc.

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

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

Dataset :- Age, income, members in family,
education, number of credit cards, credit scores,
no. of loans, asset value etc.

Exercise 4-9

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

Now we estimate for which ϵ $P(D \leq \epsilon) \approx 0.8$.

i. By Chebyshev inequality,

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

$$|D - \mu| > \epsilon$$

$$\begin{aligned} \text{Squaring both sides, } P((D - \mu)^2 > (\epsilon)^2) \\ (0.1 - 0.9)^2 > (\epsilon)^2 \end{aligned}$$

$$(0.1 - 0.9)^2 > (\epsilon)^2 \Rightarrow (0.8)^2 > (\epsilon)^2$$

So we can say for $\epsilon \leq 2e^{-\frac{2\epsilon^2 N}{\sigma^2}}$, $\epsilon = 0.8$.

$$\Rightarrow P[|D - \mu| > \epsilon] \leq 2e^{-\frac{2(0.8)^2 (10)}{\sigma^2}} \leq 2e^{-\frac{2(0.8)^2 (10)}{(0.1)^2}} \leq 5.52 \times 10^{-6}$$

$$\Rightarrow P(D \leq 0.1) \leq P[|D - \mu| > \epsilon] \leq 5.52 \times 10^{-6}$$

$$\Rightarrow 5.52 \times 10^{-6} \leq P[|D - \mu| > \epsilon] \leq 5.52 \times 10^{-6}$$

$$\leftarrow \overbrace{(1 - p)(p)}^{P(D \leq 0.1)} \rightarrow$$

$$(1 - p)(p) + (0 - p)(p) \leq (1.0 \geq 0)$$

$$p(1-p) + p(0-p) \leq 1$$

$$p(1-p) = p(1-p) + p(0-p)$$

Now, we have given this, is, $x_2 = a x_1 + b$ and for
where $a = \text{slope}$ & $b = \text{intercept}$

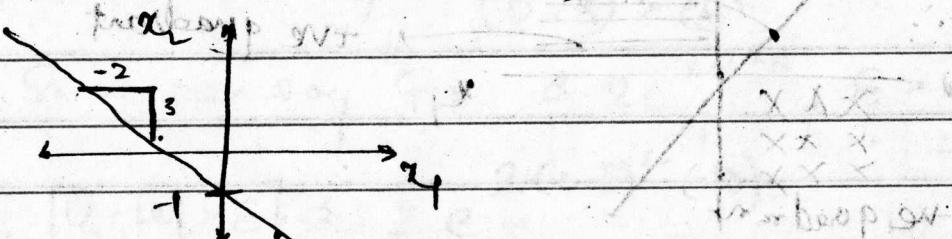
$$\therefore x_2 = -w_1/w_2 x_1 + (-w_0/w_2) = x$$

$$\therefore \begin{cases} a = -w_1/w_2 \\ b = -w_0/w_2 \text{ as } x_0 = 1 \end{cases}$$

Ans, part (a)

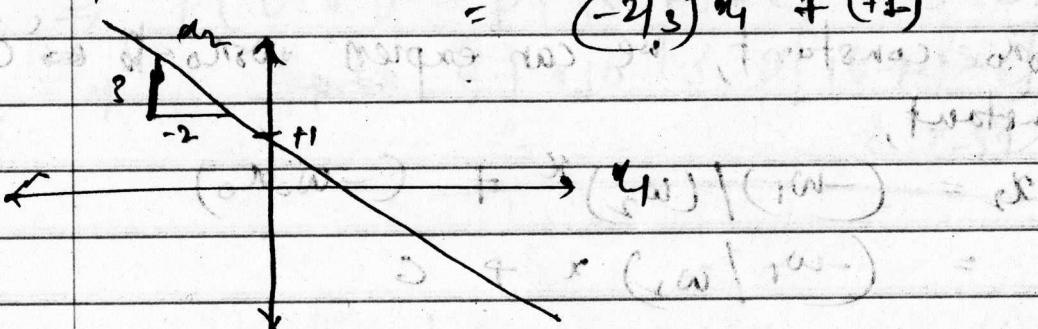
part (b) $\therefore w = [1, 2, 3]$

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



$$\therefore w = [-1, -2, 0]$$

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



and so there is a line

problem 1.11:

a) Supermarket:

		+	+1	-1
+1		0	1	
-1		10	0	

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

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

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

b) for C1A: matrix \Rightarrow

		+	+1	-1
+1		0	1000	
-1		1	0	

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

$$E_{in}(g) = \frac{1}{N} \left[\left(\sum_{\substack{y_{in}=+1 \\ y_{in}=-1}} (g(x_n) \neq +1) \right) x 1 \right. \\ \left. + \left(\sum_{\substack{y_{in}=+1 \\ y_{in}=-1}} (g(x_n) \neq -1) \right) x 1000 \right].$$