

Perceptron Learning Convergence Theorem:-

Let $x(n)$ be the input vectors & $w(n)$ be the weight vectors, b is bias, y be the actual response & d be the desired output & η be the learning rate parameter, $0 < \eta \leq 1$.

- ① Initili-Initialization. Set $w(0) = 0$. Then perform the following computations for time step $n = 1, 2, \dots$
- ② Activation: At time step n , activate the perceptron by applying continuous valued input vectors $x(n)$ & desired response $d(n)$.
- ③ Computation of Actual response of perceptron

$$y(n) = \text{sgn}[w^T(n) x(n)]$$

signum function.

- ④ Adaption Adaptation of Weight ~~to~~ Vectors: update the weight vector of perceptron:

$$w(n+1) = w_j(n) + \eta [d(n) - y(n)] x(n)$$
$$b(n+1) = b(n) + \eta [d(n) - y(n)]$$

Where, $d(n) = \begin{cases} +1, & \text{if } x(n) \text{ belongs to class I} \\ -1, & \text{if } x(n) \text{ belongs to class II} \end{cases}$

⑤ Continuation: Increment timestep n by one & go back to ③.

Example: -

	x_1	x_2	class	d
x_1	0	2	C_1	1
x_2	0	1	C_1	1
x_3	1	0	C_2	-1
x_4	1	1	C_2	-1

① initial weight vector $w(0) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Iteration 1: $w^T(0) \cdot x(1) = \begin{bmatrix} 0 & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 2 \end{bmatrix} = 0$

Ex $\rightarrow w_{11} = 1, w_{12} = 1, b = -1$

$a = w_i^T x_i + b$

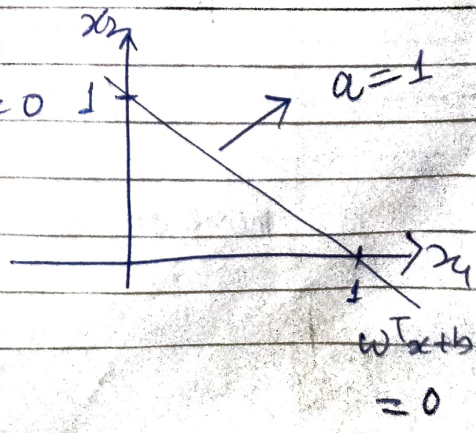
$1 \cdot x_1 + 1 \cdot x_2 + b = 0$


$x_1 + x_2 - 1 = 0$

To draw the line, we can find the points, where it intersects the x_1 & x_2

if $x_1 = 0, \Rightarrow 0 + x_2 - 1 = 0$
 $x_2 = 1$

if $x_2 = 0 \Rightarrow x_1 + 0 - 1 = 0$
 $x_1 = 1$





$$x = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

$$a = \text{hardlim}(\omega^T x + b)$$

$$z = \text{hardlim}\left(\begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 0 \end{bmatrix} - 1\right) = 1$$

output will be 1, for the region above and to the right of the decision boundary.

Ex →

0	0	0
0	1	0
● 1	0	0
1	1	1

① design the decision boundary.

② choose a weight vector that is orthogonal to the decision boundary.

