

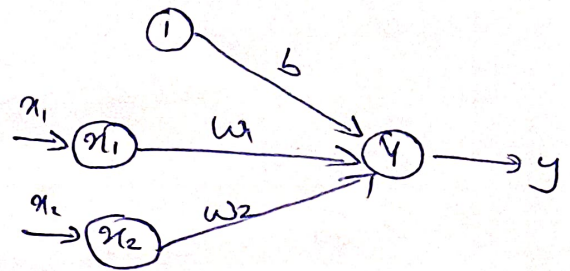
# Soft Computing

- Implement OR function using the perceptron training algorithm with binary input and bipolar target

Ans

Truth Table:

$x_1$	$x_2$	$t$
1	1	1
1	-1	1
-1	1	1
-1	-1	-1



Initial weight  $w_1 = w_2 = 0$   $b = 0$   $\alpha = 1$

$$y = \begin{cases} 1 & \text{if } y_{in} > 0 \\ 0 & \text{if } y_{in} = 0 \\ -1 & \text{if } y_{in} < 0 \end{cases}$$

EPOCH-1

where;  $y_{in} = w_1 x_1 + w_2 x_2 + b$

Pattern 1:  $x_1 = 1$   $x_2 = 1$   $t = 1$

$$y_{in} = (1 \times 0) + (1 \times 0) + 0 = 0 \quad y_{in} = 0 \Rightarrow y = 0$$

$t = 1, y = 0$ : error: update required ( $t \neq y$ )

$$w_1(\text{new}) = w_1(\text{old}) + \alpha t x_1$$

$$w_1(\text{new}) = 0 + (1 \times 1 \times 1) = 1$$

$$w_2(\text{new}) = 0 + (1 \times 1 \times 1) = 1$$

$$b(\text{new}) = 0 + 1 \times 1 = 1 \quad b(\text{new}) = b(\text{old}) + \alpha t$$

Pattern 2 ( $x_1 = 1$   $x_2 = -1$   $t = 1$ )

$$y_{in} = (1 \times 1) + (-1 \times 1) + 1 = 1$$

$$y_{in} = 1 \Rightarrow y = 1$$

$t = 1, y = 1$   $t = y$  (no update)  $w_1 = w_2 = b = 1$

Pattern 3:  $x_1 = -1$   $x_2 = 1$   $t = 1$

$$y_{in} = (-1 \times 1) + (1 \times 1) + 1 = 1$$

$$y = 1, t = 1, y = t \text{ (no update)}$$

Pattern 4: ( $x_1 = -1$   $x_2 = -1$   $t = -1$ )

$$y_{in} = (-1 \times 1) + (-1 \times 1) + 1 = -1$$

$$y_{in} < 0 \Rightarrow y = -1$$

$t = -1, y = -1$   $t = y$  & no update



## EPOCH 2

All outputs are correct, so no wt change. Training stop  
Final weights:  $w_1=1, w_2=1, b=1$

2. Explain the Training algorithm used for a perceptron network with single output classes.

The Perceptron algorithm proposed by Frank Rosenblatt is a supervised learning Algo used to train single layered perceptron.

Steps in Training:

Step 0: Initialization:

Initialize all wt to small random value or zero

Initialize bias  $b$  and also choose learning rate  $\alpha$  ( $0 < \alpha \leq 1$ )

Step 1: Perform step 2-6 until final stopping condition is reached

Step 2: Perform step 3-5 for each training pair indicated by  $s_i$

Step 3: The input layer containing input unit is applied with identity activation  $f^n$   $x_i = s_i$

Step 4: Calculate the o/p of network

$$y_{in} = b + \sum_{i=1}^n x_i w_i$$

$$Y = f(y_{in}) = \begin{cases} 1 & \text{if } y_{in} \geq \theta \\ 0 & \text{if } -\theta \leq y_{in} < \theta \\ -1 & \text{if } y_{in} < -\theta \end{cases}$$

Step 5: weight and bias adjustment

if  $y \neq t$   $w_{new} = w_{old} + \alpha t x_i$ ,  $b_{new} = b_{old} + \alpha t$

else  $w_{new} = w_{old}$ ,  $b_{new} = b_{old}$

Step 6: Train the network until there is not wt change, else start from 2.

