



training pattern.

$$w_i(\text{new}) = w_i(\text{old}) + \alpha t \cdot x_i;$$

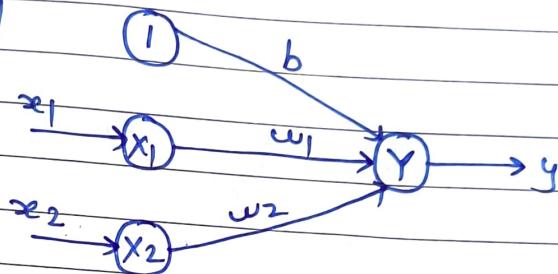
$$b(\text{new}) = b(\text{old}) + \alpha t.$$

Example No.1

1. Implement AND function using perceptron network for bipolar input and target

→ Truth table.

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



The n/w input pattern are presented to the n/w one by one, when all four input pattern are presented then one epoch is said to be completed.

The initial weights $\rightarrow w_1 = w_2 = 0$

The initial bias $\rightarrow b = 0$

The initial threshold $\rightarrow \theta = 0$

The initial learning rate $\rightarrow \alpha = 1$

Case①: For first i/p pattern

$$x_1 = 1, x_2 = 1, t = 1$$

$$w_1 = w_2 = b = 0.$$

Step ① : Net input $y_{in} = x_1 w_1 + x_2 w_2 + b$

$$= (1 \times 0) + (1 \times 0) + 0$$

$$= 0$$





Step ② : The output y is computed by applying activation over the net input calculated.

$$\begin{aligned} Y = F(y_{in}) &= 1 \text{ if } y_{in} > 0 \\ &= 0 \text{ if } y_{in} = 0 \\ &= -1 \text{ if } y_{in} < 0 \end{aligned}$$

Here, $\alpha_p \cdot y = 0$ since $y_{in} = 0$

Now, target (t) = 1
 $\therefore t \neq y$

Step ③ .

Hence, we need to update the weights and bias.

$$\begin{aligned} w_1(\text{new}) &= w_1(\text{old}) + \alpha t \cdot x_1 \\ &= 0 + 1 \cdot 1 \cdot 1 \\ &= 1 \end{aligned}$$

$$\begin{aligned} w_2(\text{new}) &= w_2(\text{old}) + \alpha t \cdot x_2 \\ &= 0 + 1 \cdot 1 \cdot 1 \\ &= 1 \end{aligned}$$

$$\begin{aligned} b(\text{new}) &= b(\text{old}) + \alpha t \\ &= 0 + 1 \cdot 1 \\ &= 1 \end{aligned}$$

Case ② Second pair of $s:t$

$$x_1 = 1 \quad x_2 = -1 \quad t = -1$$

$$w_1 = w_2 = b = 1$$

Step ① $\therefore y_{in} = x_1 \cdot w_1 + x_2 \cdot w_2 + b$

$$\begin{aligned} &= (1 \times 1) + (-1)(1) + 1 \\ &= 1 - 1 + 1 \\ &= 1 \end{aligned}$$



Step ② Apply activation function
here, o/p $y = 1$ since $y_{in} > 0$
Now, $t = -1$, $\therefore y \neq t$.

Step ③ Now update the weights.

$$\begin{aligned}\omega_1(\text{new}) &= \omega_1(\text{old}) + \alpha t \cdot x_1 \\ &= 1 + (1)(-1)(1) = 0\end{aligned}$$

$$\begin{aligned}\omega_2(\text{new}) &= \omega_2(\text{old}) + \alpha t \cdot x_1 \\ &= 1 + (1)(-1)(-1) = 2\end{aligned}$$

$$\begin{aligned}b(\text{new}) &= b(\text{old}) + \alpha t \\ &= 1 + (1)(-1) = 0\end{aligned}$$

case ③ Third input pair of s:t

$$x_1 = -1, x_2 = 1, t = -1$$

$$\omega_1 = 0, \omega_2 = 2, b = 0$$

Step ① Net I/p $y_{in} = x_1 \cdot \omega_1 + x_2 \cdot \omega_2 + b$

$$\begin{aligned} &= (-1)(0) + (1)(2) + 0 \\ &= 2\end{aligned}$$

Step ② Net I/p y_{fn} By applying activation function
here $\therefore y = 2$ since $y_{in} > 0$
Now, $t = 1$ and $\therefore y \neq t$

Step ③ Hence, updation of weights required,
 $\omega_1(\text{new}) = \omega_1(\text{old}) + \alpha t \cdot x_1$

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

$$\begin{aligned}\omega_2(\text{new}) &= \omega_2(\text{old}) + \alpha t \cdot x_2 \\ &= 2 + (1)(-1)(+1) = 1\end{aligned}$$



$$\begin{aligned}b(\text{new}) &= b(\text{old}) + \alpha t \\&= 0 + (1)(-1) = -1\end{aligned}$$

case④ Fourth pair of S:t

$$\begin{array}{lll}x_1 = -1 & x_2 = -1 & t = -1 \\w_1 = 1 & w_2 = 1 & b = -1\end{array}$$

$$\begin{aligned}\text{Step① Net i/p } y_{\text{in}} &= x_1 \cdot w_1 + x_2 \cdot w_2 + b \\&= (1)(-1) + (1)(-1) + (-1) \\&= -3\end{aligned}$$

Step② By applying activation function,

Here, o/p $y = -1$ Since $y_{\text{in}} < 0$

Now, $t = -1$ and $y = -1$

so $y = t$

\therefore No updation of weight and bias

$$w_1(\text{new}) = w_1(\text{old}) = 1$$

$$w_2(\text{new}) = w_2(\text{old}) = 1$$

$$b(\text{new}) = b(\text{old}) = -1$$

Here, first epoch is completed.

Similarly, we have to complete second epoch



Epochs	Inputs			Target t	Net i/p (yin)	Calculated o/p	Updated weights & bias		
	x_1	x_2	t				w_1	w_2	b
①	1	1	1	1	0	0	1	1	1
	1	-1	1	-1	1	1	0	2	0
	-1	1	1	-1	2	1	1	+1	-1
	-1	-1	1	-1	-3	-1	1	1	-1
②	1	1	1	1	1	1	1	1	-1
	1	-1	1	-1	-1	-1	1	1	-1
	-1	1	1	-1	-1	-1	1	1	-1
	-1	-1	1	-1	-3	-1	1	1	-1



Example No. 2

2. Implement OR function using perceptron networks for bipolar inputs and bipolar targets.

Truth table.

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

Case ① First input pair

$$x_1 = 1 \quad x_2 = 1 \quad t = 1$$

$$w_1 = w_2 = b = 0$$

$$\begin{aligned} \text{Step ① Net i/p } y_{in} &= x_1 \cdot w_1 + x_2 \cdot w_2 + b \\ &= (1) + (0) + (1)(0) + 0 \\ &= 0 \end{aligned}$$

Step ② Net i/p y_{in} - Apply activation function

Here, o/p $= y = 1$ Since $y_{in} > 0$

Now, $t = 1$ and $y_{in} = 1 \quad \therefore t = y$

Step ③ No need to weight updation.

$$w_1(\text{new}) = w_1(\text{old}) = 1$$

$$w_2(\text{new}) = w_2(\text{old}) = 1$$

$$b(\text{new}) = b(\text{old}) = 1$$

$$x_1 = -1 \quad x_2 = 1 \quad b = 1 \quad t = 1$$

$$\begin{aligned} \text{Case ③ } x_1 &= -1 \quad x_2 = 1 \quad b = 1 \quad t = 1 \\ w_1 = w_2 &= 1 \end{aligned}$$





Step ① Net I/p $y_{in} = x_1w_1 + x_2 \cdot w_2 + b$

$$= (1)(-1) + (1)(1) + 1$$
$$= 1$$

Step ② Apply activation function.

Here, $y=1$ since $y_{in} > 0$

now $t=1$ and $y=1$
 $\therefore t=y$

Step ③ No need updation of weights.

$$w_1(\text{new}) = w_1(\text{old}) = 1$$

$$w_2(\text{new}) = w_2(\text{old}) = 1$$

$$b(\text{new}) = b(\text{old}) = 1$$

Case ④ $x_1 = -1, x_2 = -1, t = -1$

$$w_1 = w_2 = b = 1$$

Step ① Net I/p $y_{in} = x_1w_1 + x_2 \cdot w_2 + b$

$$= (-1)(-1) + (-1)(1) + 1$$
$$= -1$$

Step ② Apply activation function

Here, $y=-1$ since $y_{in} < 0$

now, $t = -1, y = -1$

Step ③ No need to weight updation.

$$w_1(\text{new}) = w_1(\text{old}) = 1$$

$$w_2(\text{new}) = w_2(\text{old}) = 1$$

$$b(\text{new}) = b(\text{old}) = 1$$

\therefore Here first epoch is completed.



Similarly, we have to complete second epoch

Epochs	Inputs		Target	Net i/p y _{in}	Calculated o/p	Updated weights & bias		
	x ₁	x ₂	t			w ₁	w ₂	b
①	1	1	1	1	0	0	1	1
	1	-1	1	1	-1	1	1	1
	-1	1	1	1	-1	1	1	1
	-1	-1	1	-1	-1	1	1	1
②	1	1	1	1	3	1	1	1
	1	-1	1	1	1	1	1	1
	-1	1	1	1	1	1	1	1
	-1	-1	1	-1	-1	1	1	1

Conclusion:-

We have implemented perceptron network
for AND and OR logic.



1. What is confusion matrix?

→ A confusion matrix is a technique for summarizing the performance of a classification algorithm.

2. Give examples of supervised learning algorithms.

→ Some of the most common algo. in Supervised Learning include Support vector Machine (SVM), Logistic, Regressive, Naive Bayes, Neural Networks, k-nearest neighbour (KNN), and random forest.

3. Write a short note on perceptron network.

→ Perceptron is understood as an Artificial Neuron or neural n/w unit that helps to detect certain i/p data computations in business intelligence.

It is one of the best and simplest types of Artificial Neural Networks. However it is a supervised learning algo. of binary classifiers. Hence, we can consider it as a single-layer neural network with four main parameters.

i.e. i/p values, weights with four main parameters and bias, net sum and an activation function ✓