

## H.W - 2

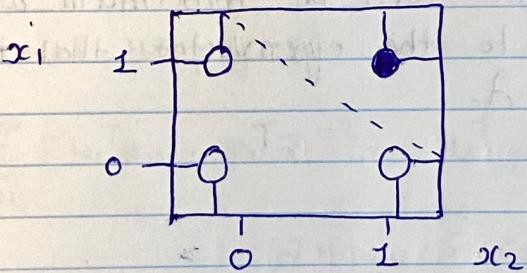
### (1) Perception Algorithm

→ Let's first define the truth table for the NAND function

→ Truth table:

$x_1$	$x_2$	Class	Label
0	1	Negative	1
1	1	Positive	0
1	0	Negative	1
0	0	Negative	1

→ From the truth table we can build follows



- → Positive class
- → Negative class

(1) From above figure we can say that for boolean NAND function, negative and positive class are linearly ~~separable~~ separable.

(2) As we know boolean function NAND is linearly separable we can use Perception Algorithm to obtain the linear decision boundary.

Let's assume our decision boundary is initially as follows.

$$x_1 + x_2 - \frac{1}{2} = 0$$

And our initial weights are as follows.

$$w_0 = -\frac{1}{2} = -0.5$$

$$w_1 = 1$$

$$w_2 = 1$$

⇒ Applying Perceptron Algorithm

ITERATION: 1

for  $f(0,1)$ , Target = -1

$$\begin{aligned}y &= w_0 + w_1 x_1 + w_2 x_2 \\&= -0.5 + (1)(0) + (1)(1) \\&= -0.5 + 1 \\&= 0.5\end{aligned}$$

Here  $y > 0$  and target = -1 so we have misclassification.

Hence we update our weights.

$$\begin{array}{c|c|c}w_0 = w_0^* + \text{Target} & w_1 = w_1^* + \text{Target} * x_1 & w_2 = w_2^* + \text{Target} * x_2 \\= -0.5 + (-1) &= 1 + (-1)(0) &= 1 + (-1)(1) \\= -1.5 &= 1 &= 1 - 1 = 0\end{array}$$

for  $f(1,1)$ , Target = +1

$$\begin{aligned}y &= w_0 + w_1 x_1 + w_2 x_2 \\&= -1.5 + (1)(1) + (0)(1) \\&= -1.5 + 1 \\&= -0.5\end{aligned}$$

Here  $y < 0$  and target = +1 so we have misclassification.

Hence we update our weights.

$$\begin{aligned} w_0 &= w_0^* + \text{Target} \\ &= -1.5 + 1 \\ &= -0.5 \end{aligned}$$

$$\begin{aligned} w_1 &= w_1^* + \text{Target} \cdot x_1 \\ &= 1 + (1)(1) \\ &= 2 \end{aligned}$$

$$\begin{aligned} w_2 &= w_2^* + \text{Target} \cdot x_2 \\ &= 0 + (1)(0) \\ &= 0 \end{aligned}$$

for  $f(1, 0)$  Target = -1

$$\begin{aligned} y &= w_0 + w_1 x_1 + w_2 x_2 \\ &= -0.5 + (2)(1) + (0)(0) \\ &= 1.5 \end{aligned}$$

Here  $y > 0$  and target = -1 so we have misclassification.

Hence we update our weights

$$\begin{aligned} w_0 &= w_0^* + \text{Target} \\ &= -0.5 - 1 \\ &= -1.5 \end{aligned}$$

$$\begin{aligned} w_1 &= w_1^* + \text{Target} \cdot x_1 \\ &= 2 + (-1)(1) \\ &= 1 \end{aligned}$$

$$\begin{aligned} w_2 &= w_2^* + \text{Target} \cdot x_2 \\ &= 1 + (-1)(0) \\ &= 1 \end{aligned}$$

for  $f(0, 1)$  Target = -1

$$\begin{aligned} y &= w_0 + w_1 x_1 + w_2 x_2 \\ &= -1.5 + (1)(0) + (1)(1) \\ &= -1.5 + 1 \\ &= -0.5 \end{aligned}$$

Here  $y < 0$  and target = -1 so have ~~not~~ correct classification

## ITERATION - 2

for  ~~$\oplus$~~   $f(0, 1)$  Target = -1

$$\begin{aligned} y &= w_0 + w_1 x_1 + w_2 x_2 \\ &= -1.5 + (1)(0) + (1)(1) \\ &= -1.5 + 1 \\ &= -0.5 \end{aligned}$$

Here  $y < 0$  and target = -1 so we have correct classification.

for  $f(1,1)$  Target = +1

$$\begin{aligned}y &= \omega_0 + \omega_1 x_1 + \omega_2 x_2 \\&= -1.5 + (1)(1) + (1)(1) \\&= -1.5 + 2 \\&= 0.5\end{aligned}$$

Here  $y > 0$  and target = +1 so we have correct classification.

for  $f(1,0)$  Target = -1

$$\begin{aligned}y &= \omega_0 + \omega_1 x_1 + \omega_2 x_2 \\&= -1.5 + (1)(1) + (1)(0) \\&= -1.5 + 1 \\&= -0.5\end{aligned}$$

Here  $y < 0$  and target = -1 so we have correct classification

for  $f(0,0)$  Target = -1

$$\begin{aligned}y &= \omega_0 + \omega_1 x_1 + \omega_2 x_2 \\&= -1.5 + (1)(0) + (1)(0) \\&= -1.5\end{aligned}$$

Here  $y < 0$  and target = -1 so we have correct classification

Since we have successfully classified all the points correctly  
our final weights for boolean NAND function is

$$\omega_0 = -1.5 \quad \omega_1 = 1 \quad \text{and} \quad \omega_2 = 1$$

And our decision boundary will be

$$-1.5 + x_1 + x_2 = 0$$

$$x_1 + x_2 = 1.5$$