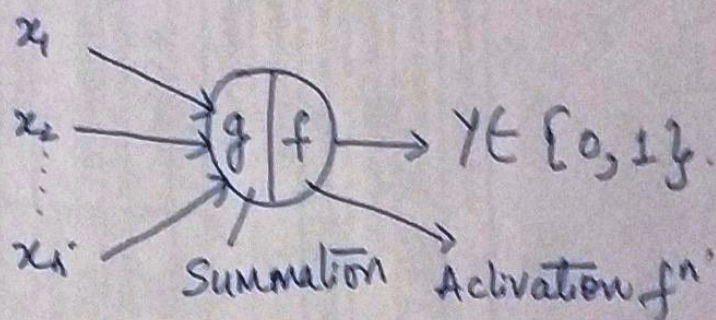


McCulloch Pitts Neuron / Thresholding Logic

- * Muculloch & pitts proposed a highly simplified Computational Model of neuron (1943).



g aggregates the input.

$$g(x) \Rightarrow \sum_{i=1}^n x_i \quad \text{--- (1)}$$

$$\Rightarrow x_1 + x_2 + \dots + x_n \quad \text{--- (2)}$$

function f takes decision based on aggregation (g).

function f is threshold function in this case.

$$y = 1 \quad \text{if } g(x) \geq \theta \quad \text{--- (3)}$$

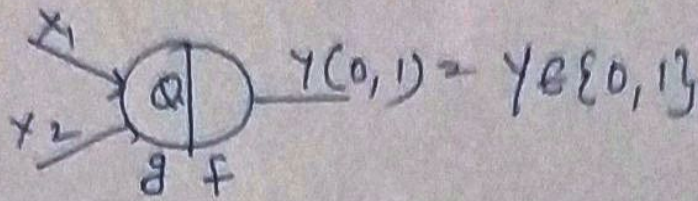
$$y = 0 \quad \text{if } g(x) < \theta \quad \text{--- (4)}$$

θ is called the thresholding parameter.

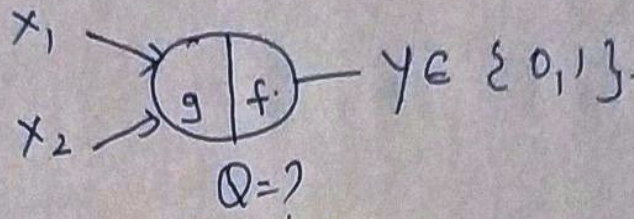
(1)

(1)

Let us implement some Boolean function
Using this McCulloch Pitts (MPP) neuron



Implement AND Gate



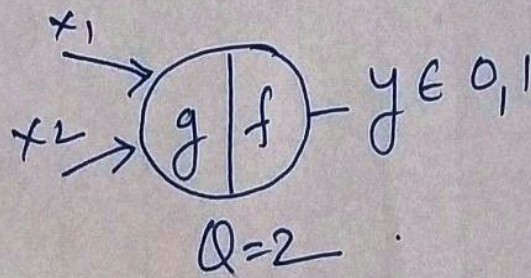
Truth Table for AND.

Data Set

x_1	x_2	y	$g = x_1 + x_2$
0	0	0	0
0	1	0	1
1	0	0	1
1	1	1	2

Target Op.

When both input are 1 we get $opp = 1$
So the $Q = 2$.

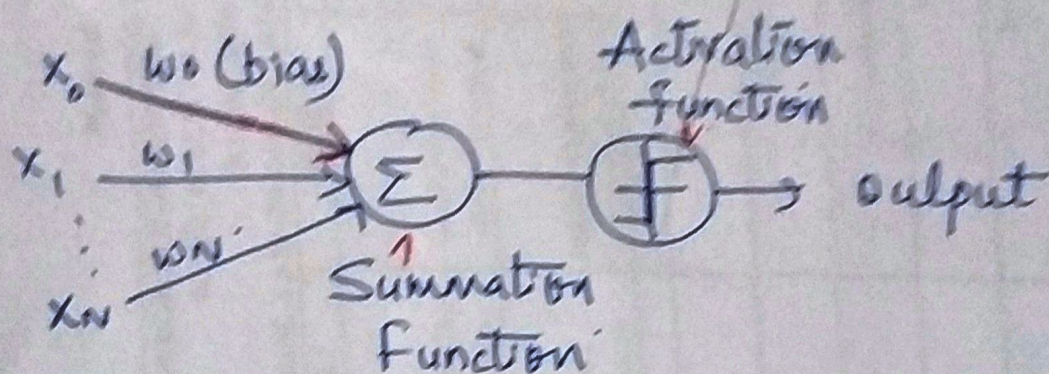
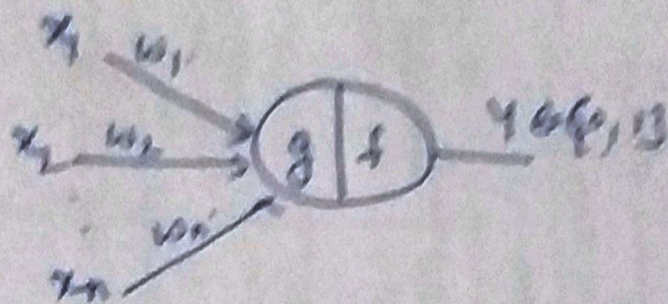


Single.

* McCulloch neuron can be used to represent
boolean function which are linearly
separable. (2)

Perceptron

Frank Rosenblatt, an American psychologist, proposed the classical perceptron model (1959).



Perceptron has 4 important Components

- ① Inputs - - - - - * Numerical data points, called input are fed into the neuron.
- ② Weight & Bias
- ③ Summation Function
- ④ Activation Function

* The weight of each input data indicates how important that input is predicting the outcomes.

→ The bias parameter allow you to adjust the activation function curve in such a way that a precise output is achieved

→ Summation Function → The product of the respective input & weight is taken. Adding all these products gives us the Weighted Sum.

$$\text{Weighted Sum} = \sum_{i=1}^n x_i w_i + \text{bias} \quad \text{--- (1)}$$

expand — eq (1)

$$= x_1 w_1 + x_2 w_2 + \dots + x_n w_n + \text{bias}$$

like ,

⇒ Activation function :→ To map the Weight Sum to the output.

for ex

$$\sum_{i=1}^n x_i w_i + \text{bias} \geq 0 \quad y = 1$$

$$\sum_{i=1}^n x_i w_i + \text{bias} < 0 \quad y = 0$$

P-2

Example

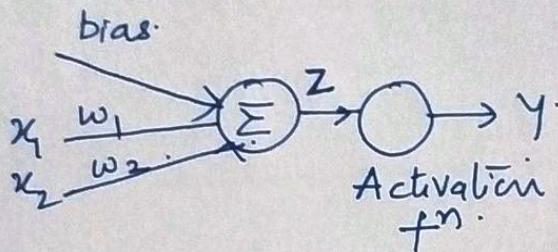
* Let us see how does a perceptron model reconstruct AND operation

x_1	x_2	O/p Label
0	0	0
0	1	0
1	0	0
1	1	1

Activation f^n

$$f(z) = \begin{cases} 1 & z \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

We Want



T

$$z = w_1 x_1 + w_2 x_2 + b = \text{Label} \quad (\text{we want})$$
$$\begin{aligned} w_1 \cdot 0 + w_2 \cdot 0 + b &= 0 & \text{--- (1)} \\ w_1 \cdot 0 + w_2 \cdot 1 + b &= 0 & \text{--- (2)} \\ w_1 \cdot 1 + w_2 \cdot 0 + b &= 0 & \text{--- (3)} \\ w_1 \cdot 1 + w_2 \cdot 1 + b &= 1 & \text{--- (4)} \end{aligned}$$

Can we suitably adjust weight & bias parameters so that all eqⁿ (2 to 5) are satisfied.

Yes choose $w_1 = 0.3$
 $w_2 = 0.3$ & $b = 0$