

# Deep Learning



## Perceptron Learning Algorithm

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# McCulloch-Pitts (MCP) Neuron (1943)

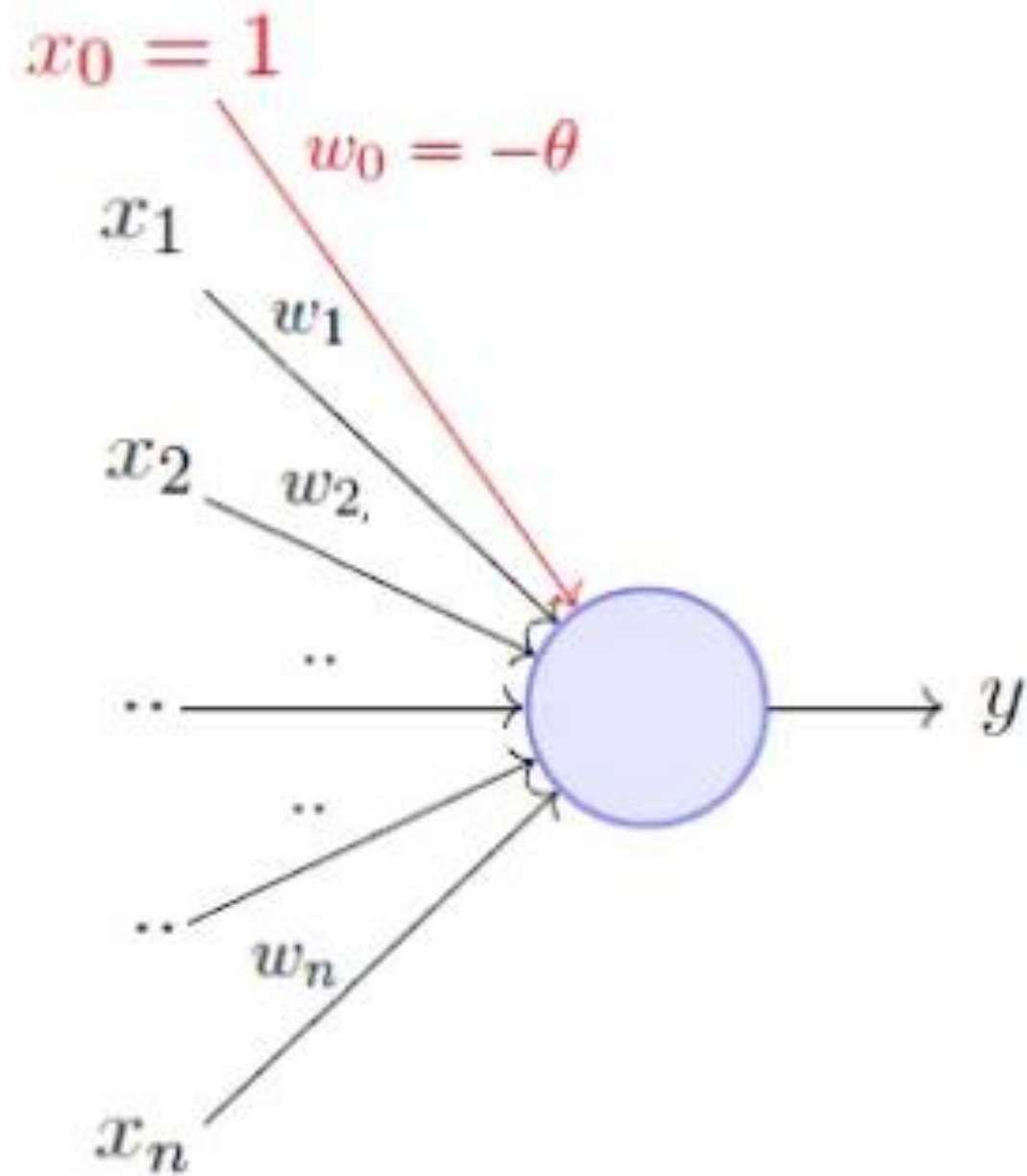
- **Drawbacks of MCP Neuron?**
  - **No Non-Boolean Inputs**
  - **Hard-coded Threshold (No Learning)**
  - **No Weightage for Inputs**
  - **Non-linearly Separable not Possible – Eg: XOR**

# The Perceptron: Learning Begins (1957)

- **Designer:** Frank Rosenblat
- Focused on drawbacks of MCP Neuron
  - **No Non-Boolean Inputs**
  - **Hard-coded Threshold (No Learning)**
  - **No Weightage for Inputs**
  - **Non-linearly Separable not Possible – Eg: XOR**

# The Perceptron: Learning Begins (1957)

- **Designer:** Frank Rosenblat
- Perceptron (Left) and MCP (Right)



$$y = 1 \quad \text{if} \quad \sum_{i=1}^n w_i * x_i - \theta \geq 0$$

$$= 0 \quad \text{if} \quad \sum_{i=1}^n w_i * x_i - \theta < 0$$

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where,  $x_0 = 1$  and  $w_0 = -\theta$

# The Perceptron Learning Algorithm

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- Mathematically,

$$y = f(Z) = \begin{cases} 1 & \text{if } Z \geq 0 \\ 0 & \text{if } Z < 0 \end{cases}$$

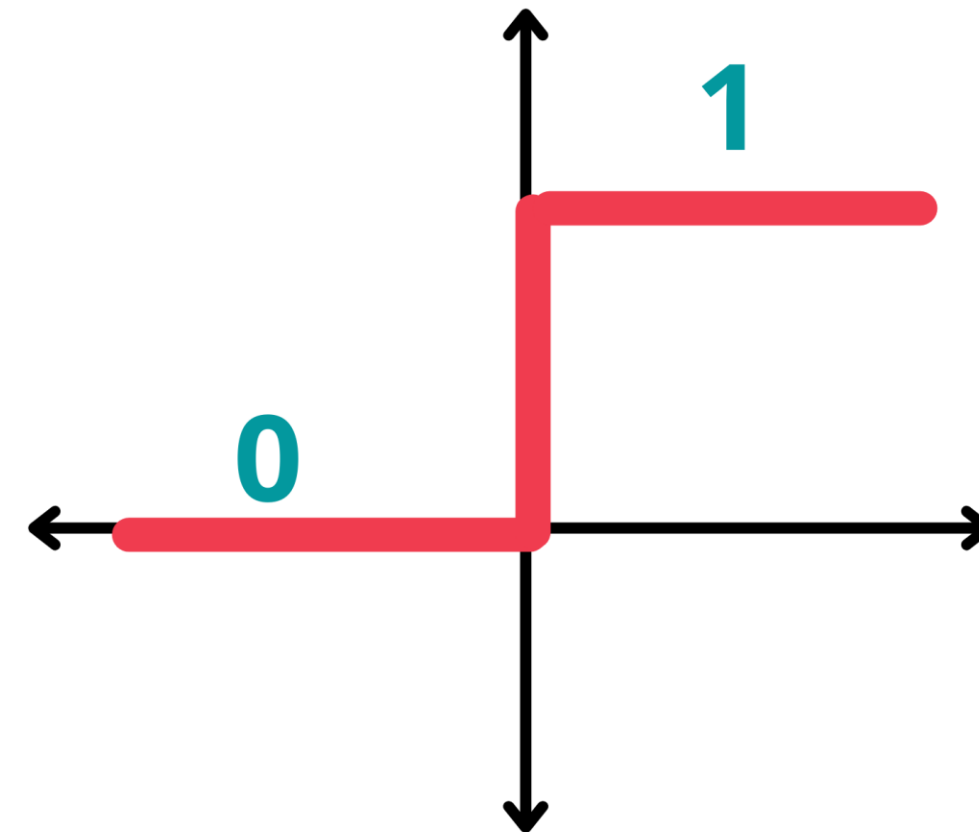


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- Sharp change at 0  $\Rightarrow$  Step function



# The Perceptron Learning Algorithm

**Algorithm:** Perceptron Learning

**Input:** Learning rate  $\alpha$ , Training dataset  $D$

**Output:** Learned weights  $w$

1. Initialize weights  $w$  to zero
2. Repeat until convergence
  - for each data point  $i$  in  $D$ 
    - if  $y_i$  misclassified
      1.  $w \leftarrow w + \alpha * (t_i - y_i) * x$
3. return  $w$

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Boolean AND gate

# References

1. **Goodfellow I, Bengio Y, Courville A, Bengio Y. Deep learning. Cambridge: MIT press; 2016 Nov 18.**

THANK YOU

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