



# ARTIFICIAL NEURAL NETWORK

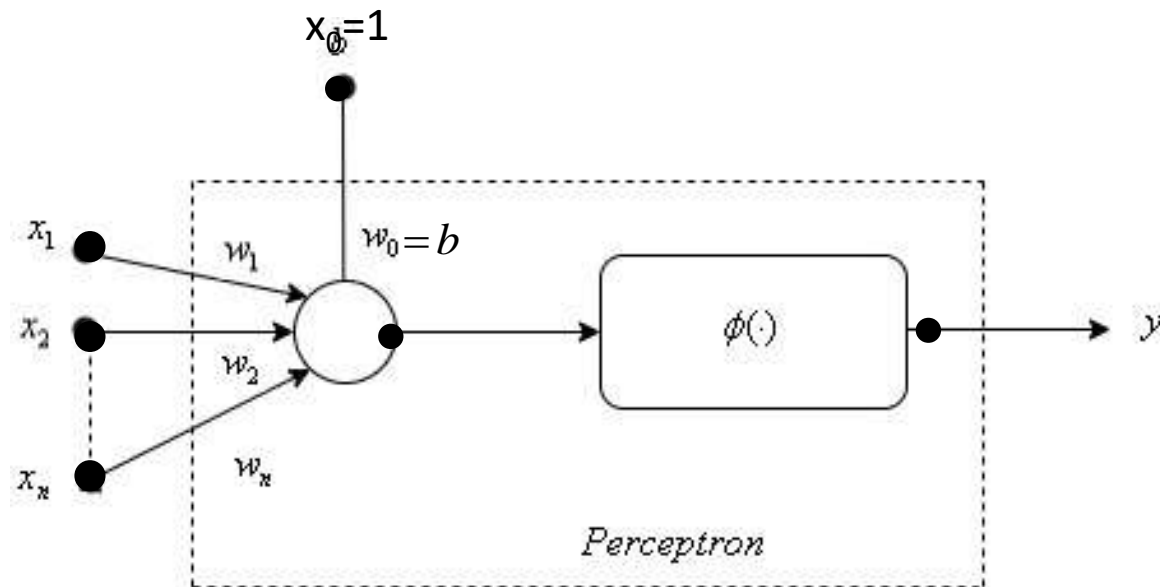
## Unit-2: Perceptron

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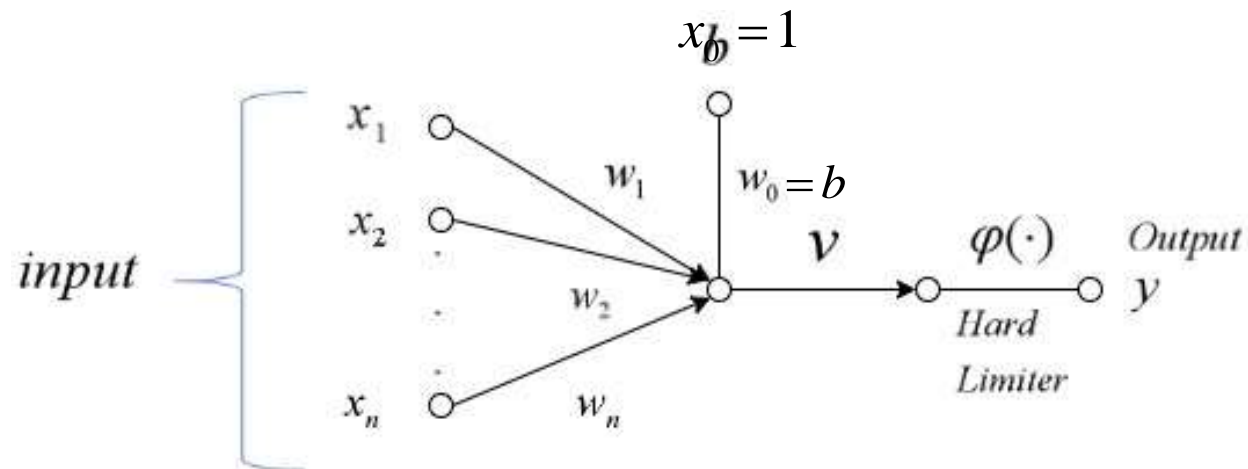
- **Perceptron:** It is a non-linear neuron model.
- The graphical representation of the perceptron is shown below:



# Artificial Neural Network

## Perceptron

- The signal flow graph of the perceptron is shown below:



- The summing node of the neuron model computes the linear combination of the inputs applied to its synapses and also incorporates an externally applied bias
- The resulting sum is applied to a hard limiter
- Accordingly neuron produces output

- From the model, the induced local field/hard limit input is

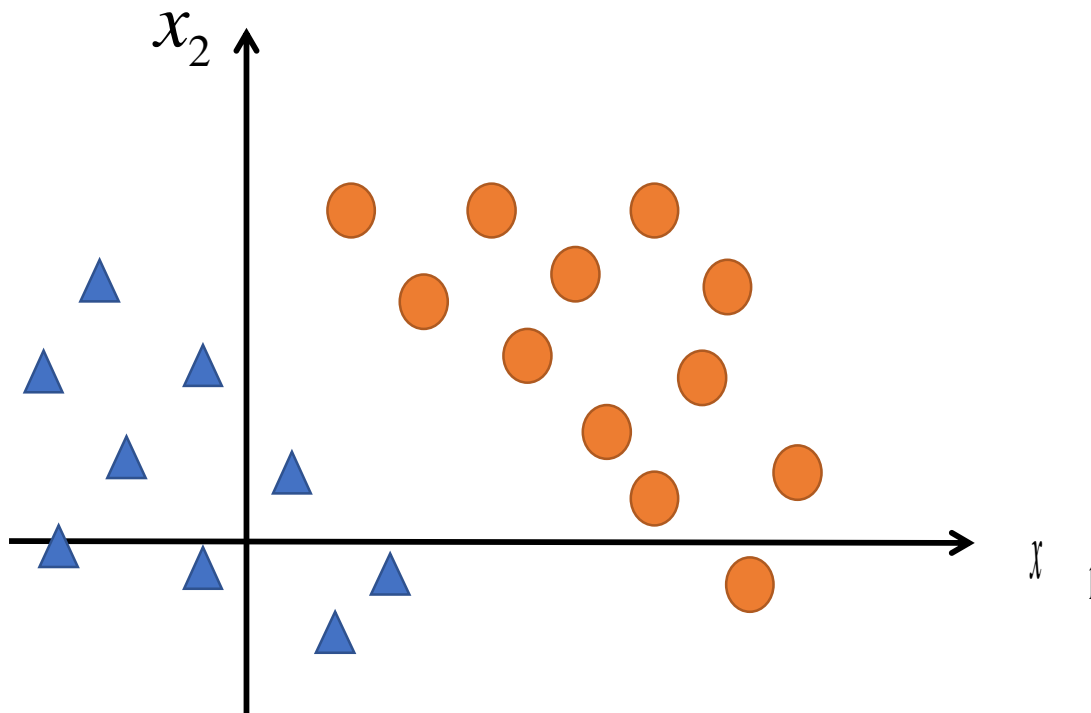
$$v = \sum_{i=1}^n w_i x_i + b$$

- The hardlimit function is defined as follows

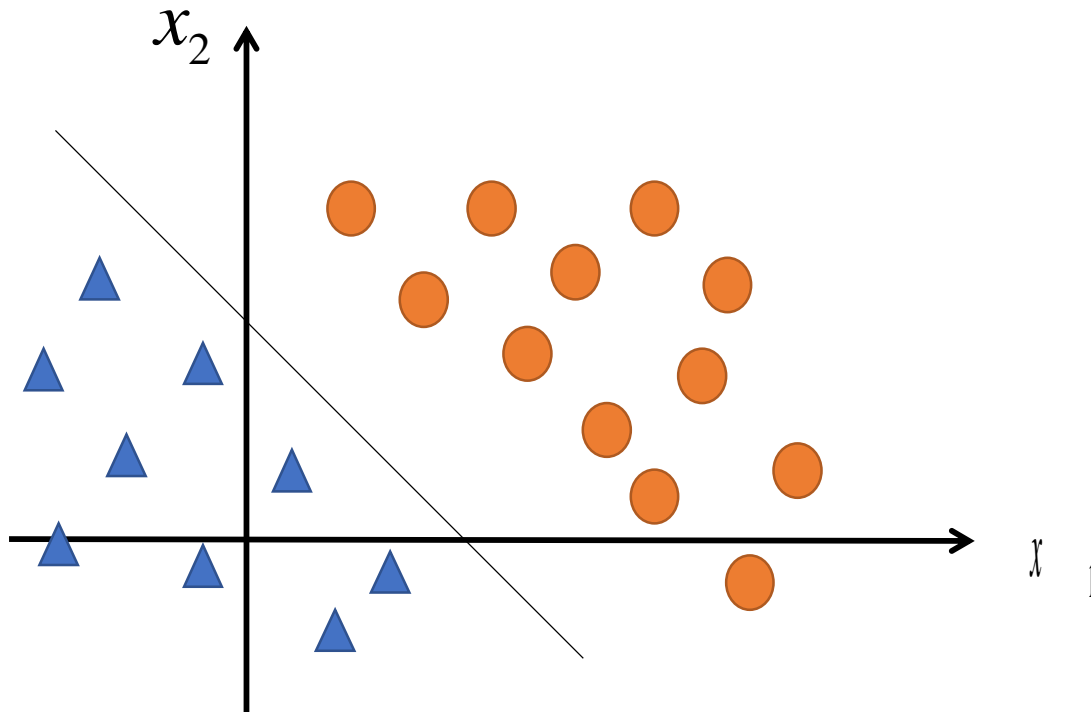
$$\varphi ( v ) = \begin{cases} 1 & v \geq 0 \\ 0 & v < 0 \end{cases}$$

- When do we use this perceptron ?
- Let us consider, we have 2 sets which is represented as in 2D- Plane as shown in the below figure

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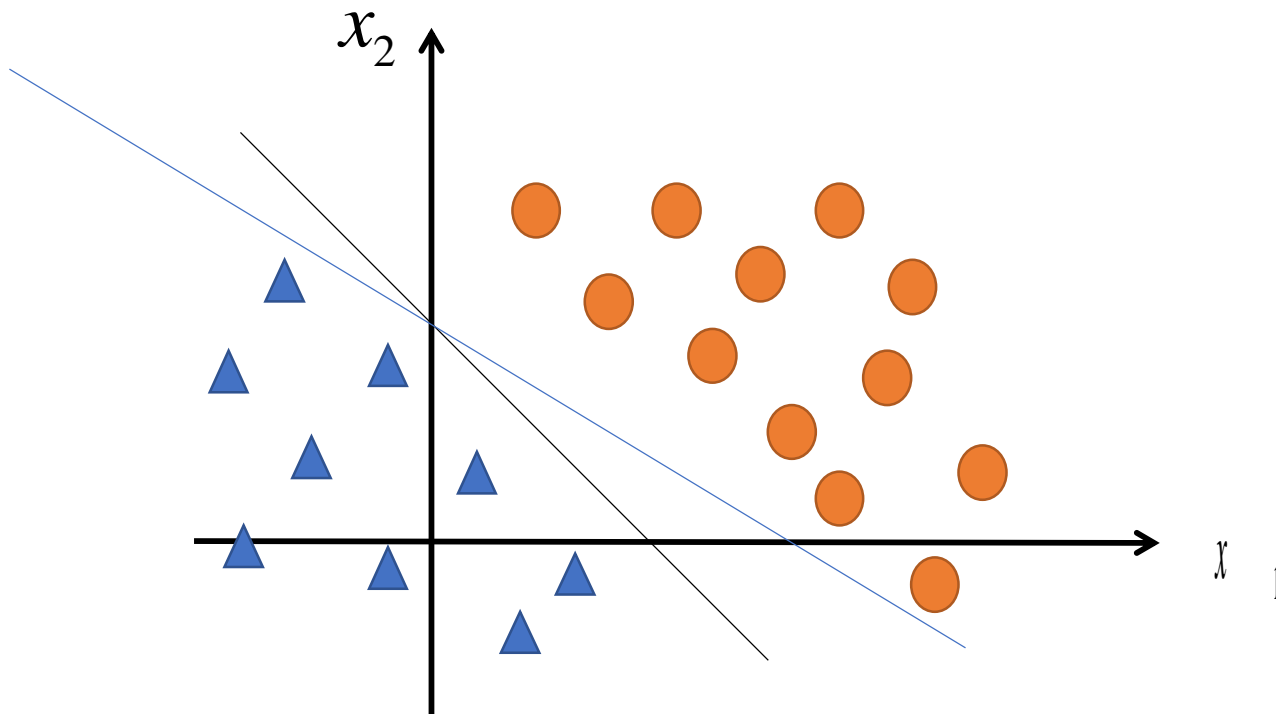


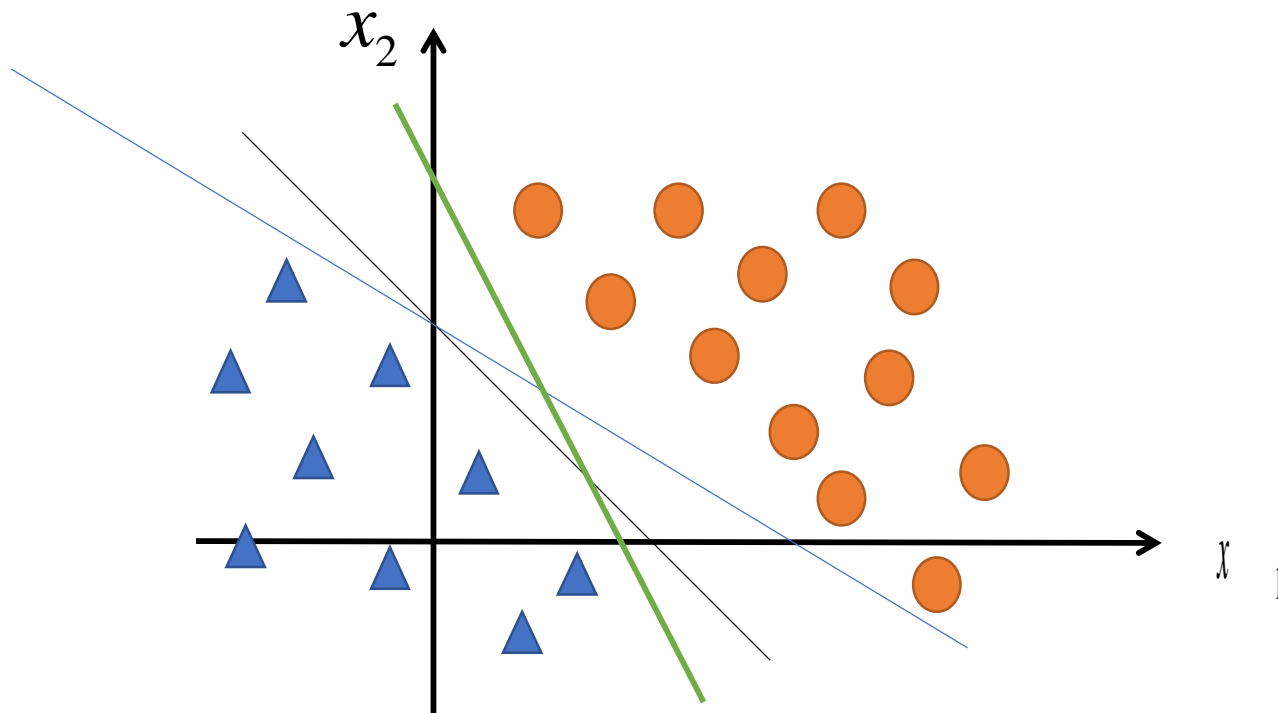
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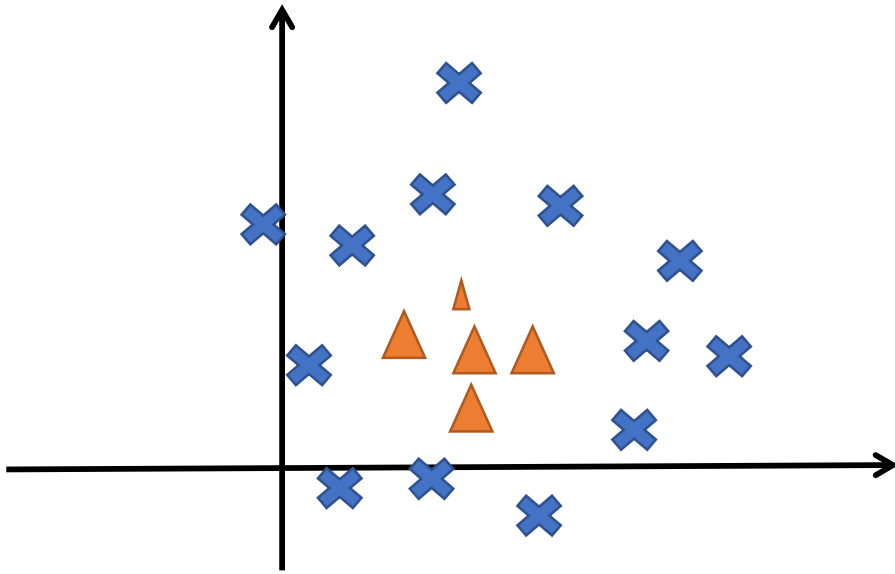
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- Let us consider, we have 2 sets which is represented as in 2D- Plane as shown in the below figure





- Therefore, if any 2 sets can be separated by line then it is referred as Linearly Separable sets otherwise it will be non-linear Separable Sets

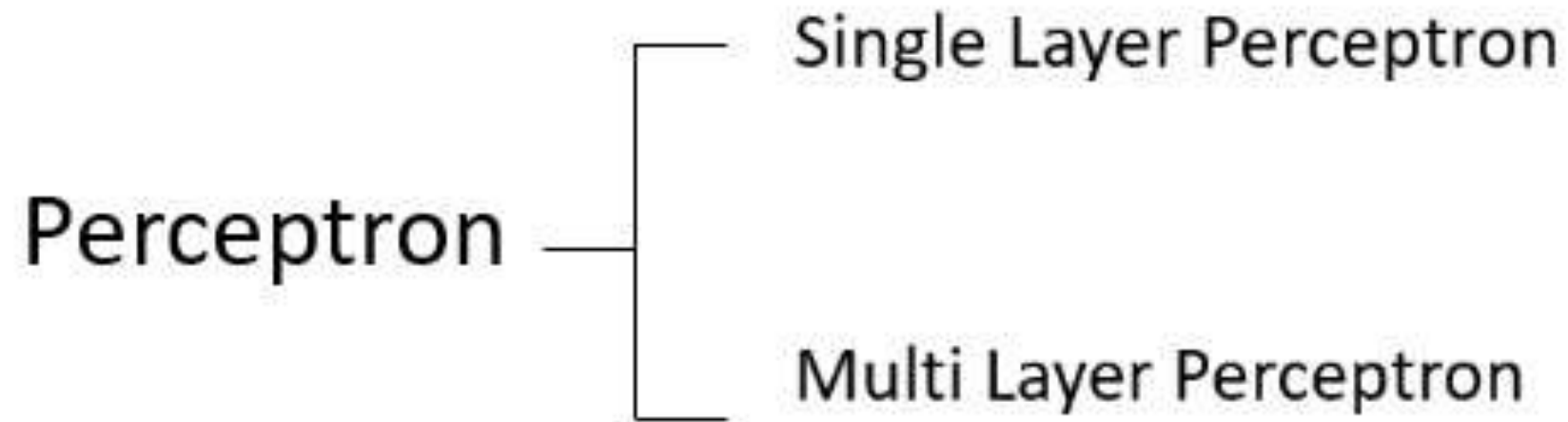
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# Artificial Neural Network

## Perceptron

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# Artificial Neural Network

## Single- Layer Perceptron

- Goal of the Single Layer Perceptron
  - Classify externally applied input into 2 classes
  - if  $v$  is greater than or equal to 0, then input  $X$  belongs to  $C_1$  otherwise, input  $X$  belongs to  $C_2$ .

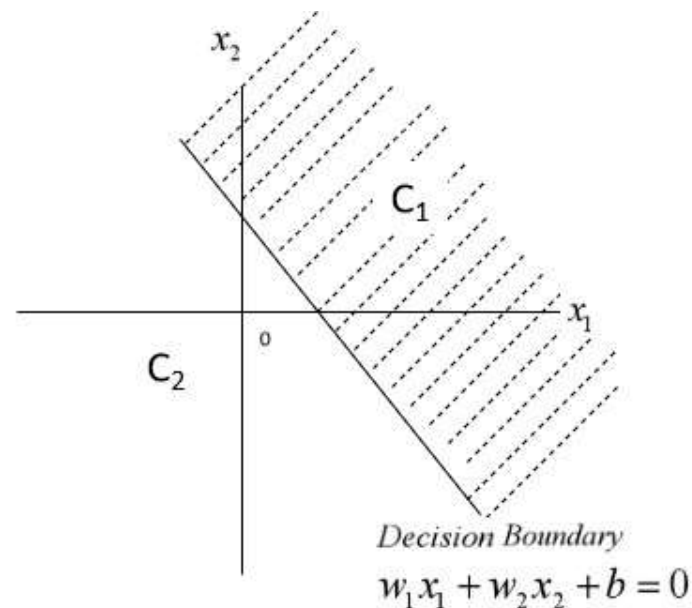


fig 1: Illustration of the hyperplane

# Artificial Neural Network

## Single-Layer Perceptron

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- Training the Single-Layer Perceptron:
- Rosenblatt's Algorithm/Perceptron Learning Rule:
  - It is a Supervised Learning.
  - This algorithm, learn the pattern/classes in finite number of steps.
  - It can be applied only for the linearly-separable classes

# Artificial Neural Network

## Single Layer Perceptron

- Single layer perceptron:  
Consider 2 input AND Logic Gate

x	y	z	
0	0	0	C <sub>2</sub>
1	0	0	
0	1	0	
1	1	1	
			C <sub>1</sub>

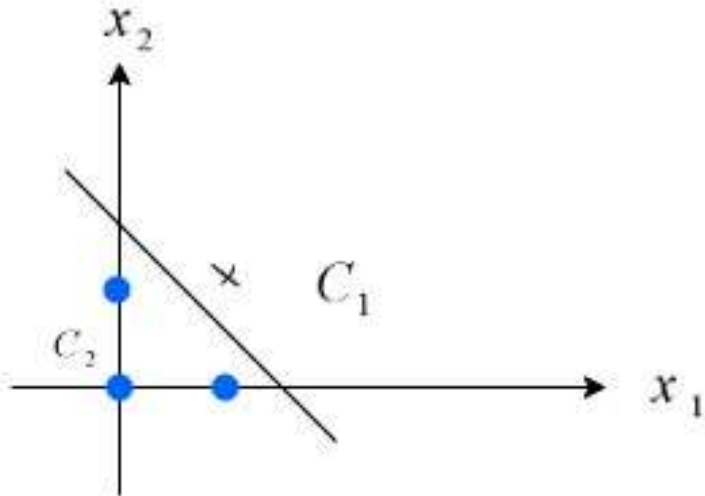


# Artificial Neural Network

## Single Layer Peceptron

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- Lets design the 2 input AND gate using single layer perceptron using Rosenblatt's Algorithm



# Artificial Neural Network

## Single- Layer Perceptron

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### Rosenblatt's algorithm:

Let  $W(1)$  be any initial choice of the weight vector and  $X(k)$  be any sequence in  $C_1 \cup C_2$

At the  $k$ th stage, Let  $W(k)$  be the weight vector

If  $X(k)$  is correctly classified, then no changes, i.e no updation of wieghts

Otherwise, updation in weights as follows

$$W(k+1) = W(k) + \begin{cases} -\eta X(k) & W^T(k)X(k) \geq 0 \text{ \& } X(k) \in C_2 \\ \eta X(k) & W^T(k)X(k) < 0 \text{ \& } X(k) \in C_1 \end{cases}$$

# Artificial Neural Network- Perceptron

## Reference

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- S. Haykin, (2003), “Neural Networks: A Comprehensive Foundation”, 2nd edition, Prentice Hall of India.





# THANK YOU

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