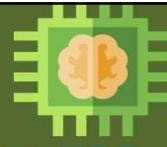


Elective Course

Course Code: CS4103

Autumn 2025-26

**Lecture #38**

Artificial Intelligence for Data Science

Week-11:**MACHINE LEARNING (Part VI)****Introduction to Neural Network****Course Instructor:****Dr. Monidipa Das**

Assistant Professor

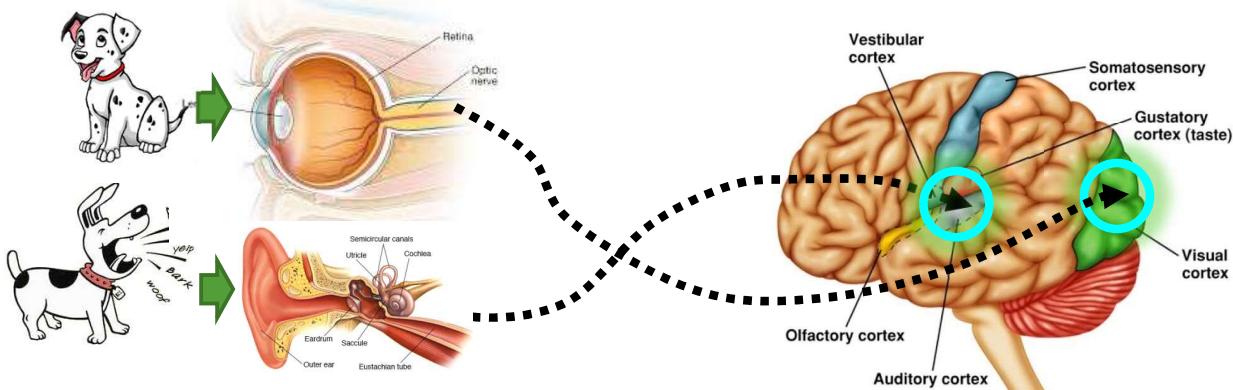
Department of Computational and Data Sciences

Indian Institute of Science Education and Research Kolkata, India 741246

Artificial Neural Networks (ANN_s) --- Origin



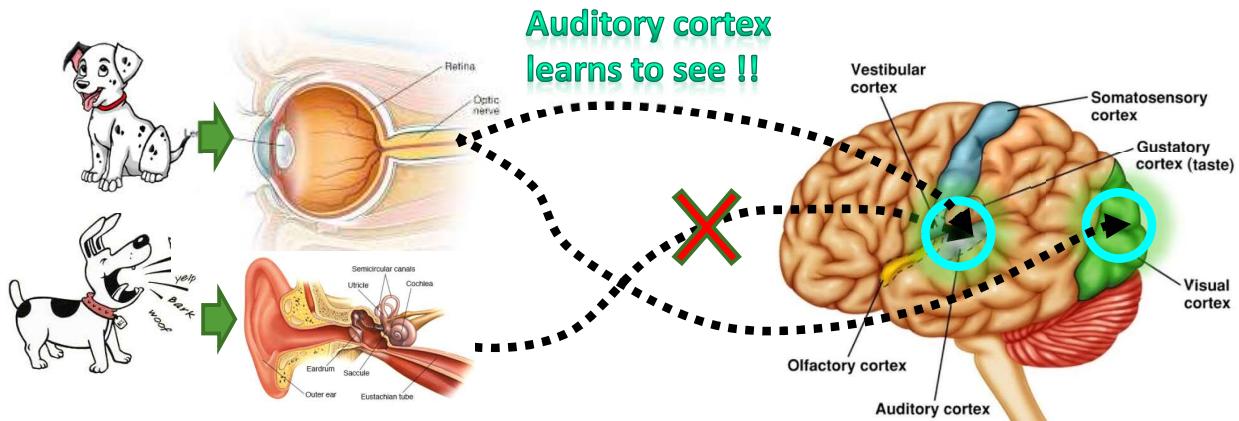
- ANNs- Information processing systems or algorithms that try to **mimic the brain**



Artificial Neural Networks (ANN_s) --- Origin

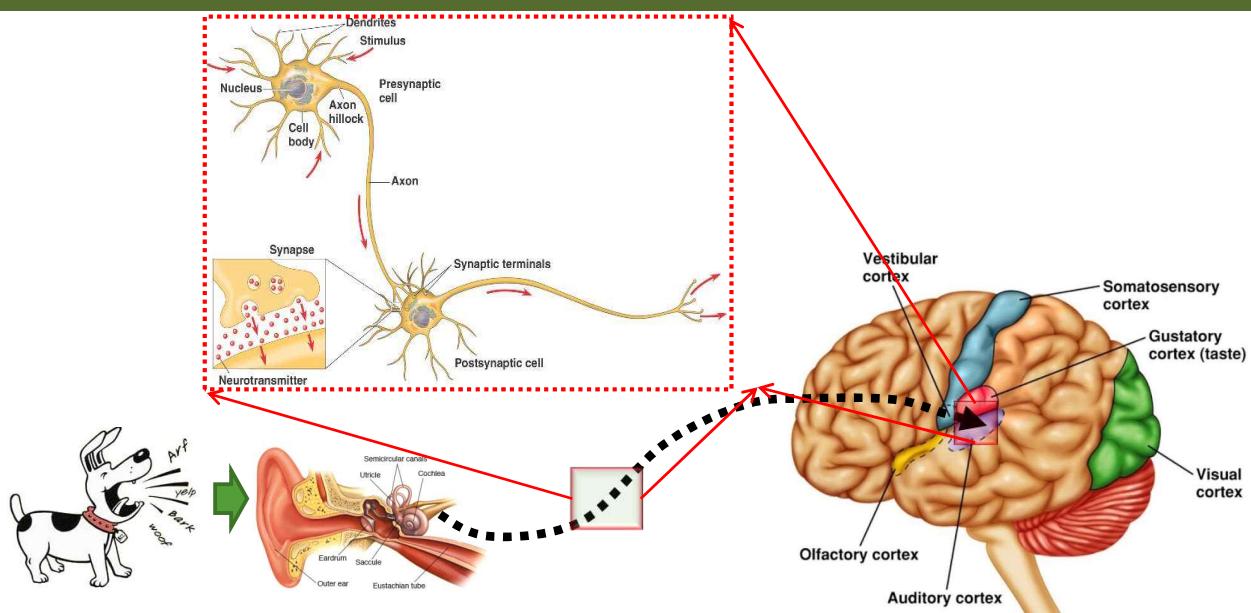


- The “one learning algorithm” hypothesis



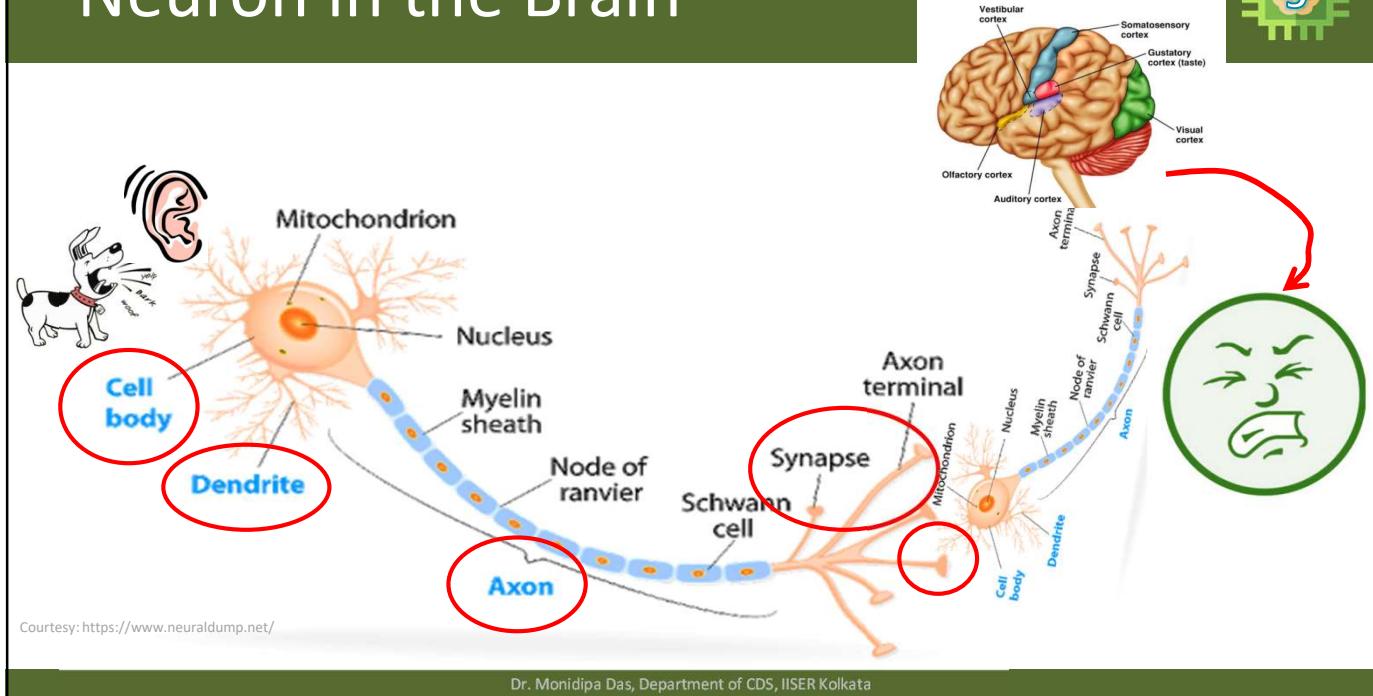
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Artificial Neural Networks (ANN_s) --- Origin



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Neuron in the Brain



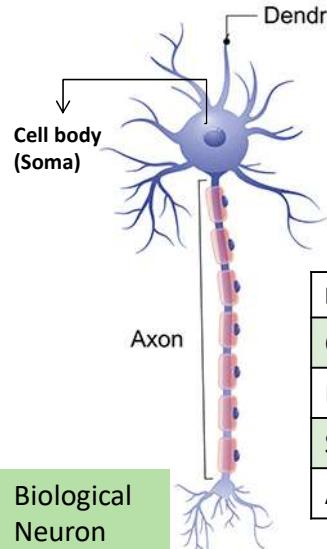
Artificial Neural Network (ANN)



- Artificial Neural Networks
 - Information processing model
 - Resembles biological neural network
 - Possesses large number of highly interconnected ***processing elements (neurons)***
 - **Objective:** to develop a computational device for modeling the brain

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ANN vs. Biological Neural Network



Artificial Neuron

$$y_{in} = \sum_{i=1}^n x_i w_i$$

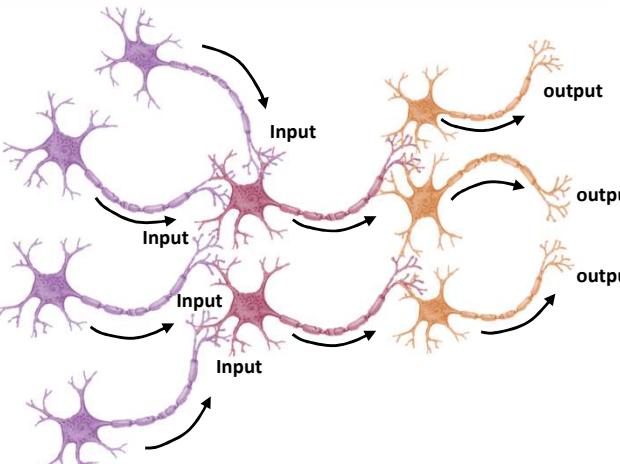
$$y = f(y_{in})$$

Processing Element

Biological Neuron	Artificial Neuron
Cell	Neuron
Dendrites	Interconnections/ Weighted-inputs
Soma (cell body)	Net input
Axon	Output

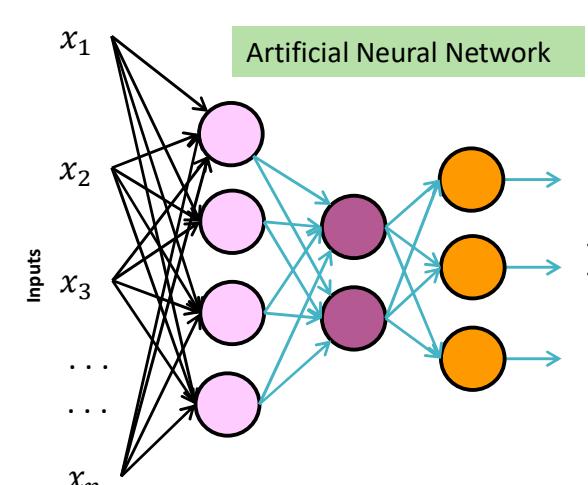
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ANN vs. Biological Neural Network



Biological Neural Network

Artificial Neural Network



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ANN vs. Biological Neural Network



Criteria	Brain	Computer
Speed	Execution time in ms	Execution time in ns
Processing	Can perform massive parallel computation; but slower	Can perform several parallel computation; but faster
Size and Complexity	10^{11} neurons and 10^{15} interconnections; Complex	Complexity depends on application and designer
Storage Capacity	Information is stored in synapse strength	Information is stored in memory May cause overload
Tolerance	Fault tolerant	Not fault tolerant
Control Mechanism	Complicated process, based on chemical action	Simpler process

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Basic Models of ANNs



- ANN models are primarily specified by the following entities:
 1. Model's *Synaptic Interconnections*
 2. *Training or learning rules* for adapting and adjusting the connection weights
 3. *Activation functions*

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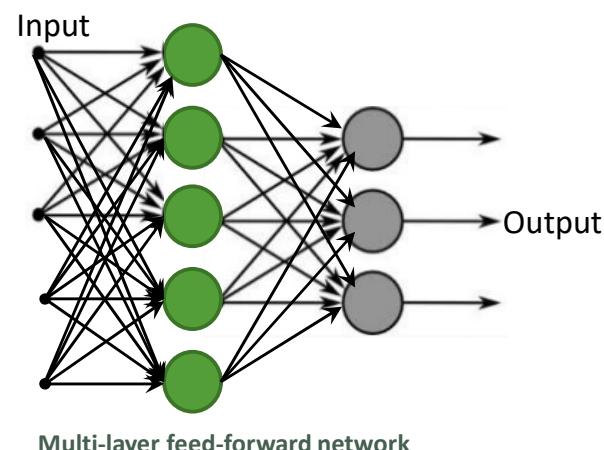
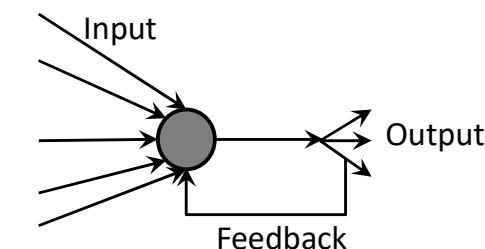
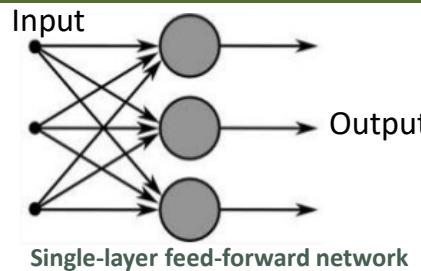
Model's Synaptic Interconnections



- Five basic types of neuron connection architectures
 1. Single-layer feed-forward network
 2. Multi-layer feed-forward network
 3. Single node with its own feedback
 4. Single-layer recurrent network
 5. Multilayer recurrent network

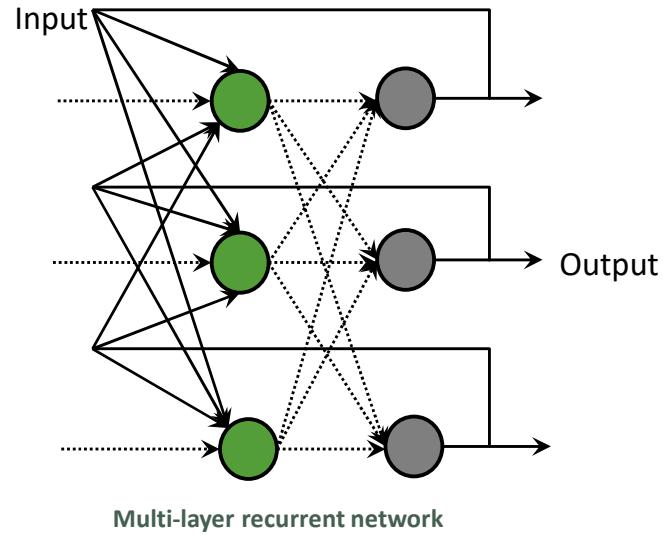
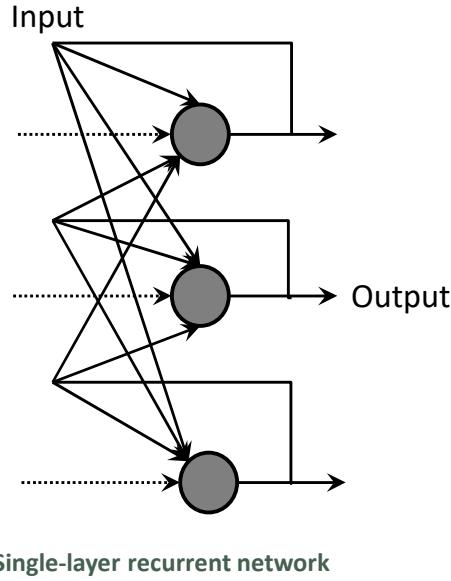
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Model's Synaptic Interconnections



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Model's Synaptic Interconnections



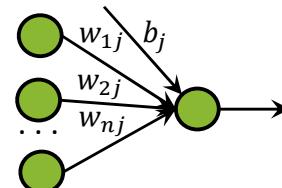
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ANN---Basic Terminologies



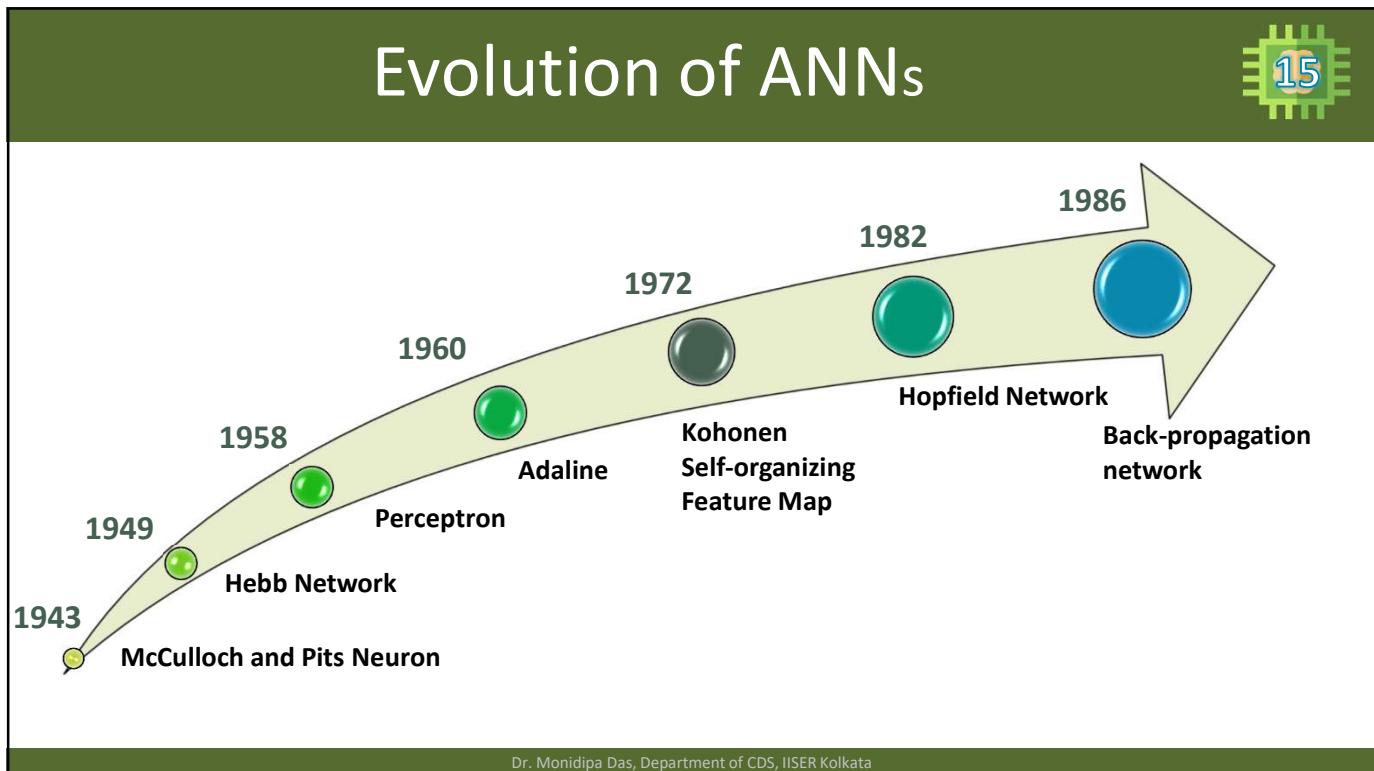
- Weights
- Bias
- Activation Function
- Learning rate
- Momentum factor

$$W = \begin{bmatrix} w_{11} & w_{12} & \dots & w_{1m} \\ w_{21} & w_{22} & \dots & w_{2m} \\ \dots & \dots & \dots & \dots \\ w_{n1} & w_{n2} & \dots & w_{nm} \end{bmatrix}$$



$$f(\text{net}) = \begin{cases} 1, & \text{if } \text{net} \geq \theta \\ -1, & \text{if } \text{net} < \theta \end{cases}$$

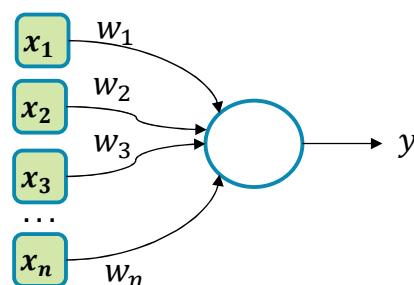
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Perceptron: Introduction

- Frank Rosenblatt (American psychologist) proposed the ***classical perceptron model*** (1958)
- Carefully analyzed by Minsky and Papert ---- ***perceptron model*** (1969)

$$y = \begin{cases} 1 & \text{if } \sum_{i=1}^n w_i x_i \geq \theta \\ 0 & \text{if } \sum_{i=1}^n w_i x_i < \theta \end{cases}$$



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Perceptron: Introduction

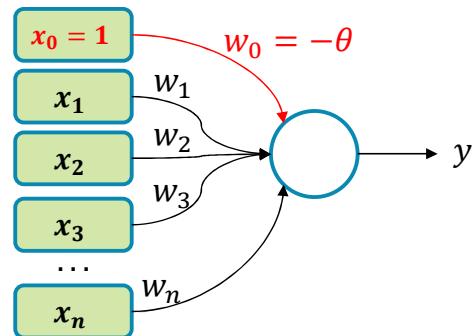


- Re-writing activation function:
- More accepted convention:

$$y = \begin{cases} 1 & \text{if } \sum_{i=0}^n w_i x_i \geq 0 \\ 0 & \text{if } \sum_{i=0}^n w_i x_i < 0 \end{cases}$$

$x_0 = 1, w_0 = -\theta$

$$y = \begin{cases} 1 & \text{if } \sum_{i=1}^n w_i x_i - \theta \geq 0 \\ 0 & \text{if } \sum_{i=1}^n w_i x_i - \theta < 0 \end{cases}$$



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Perceptron: Learning Algorithm



Algorithm: Perceptron Learning Algorithm

```

 $P \leftarrow$  inputs with label 1;
 $N \leftarrow$  inputs with label 0;
Initialize  $\mathbf{w}$  randomly;
while !convergence do
    Pick random  $\mathbf{x} \in P \cup N$  ;
    if  $\mathbf{x} \in P$  and  $\sum_{i=0}^n w_i * x_i < 0$  then
         $\mathbf{w} = \mathbf{w} + \mathbf{x}$  ;
    end
    if  $\mathbf{x} \in N$  and  $\sum_{i=0}^n w_i * x_i \geq 0$  then
         $\mathbf{w} = \mathbf{w} - \mathbf{x}$  ;
    end
end
//the algorithm converges when all the
inputs are classified correctly

```

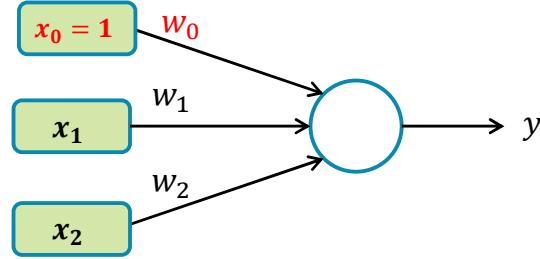
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Perceptron Learning Example



- Implement OR function with binary inputs and binary targets using perceptron training algorithm

x_1	x_2	t
0	0	0
0	1	1
1	0	1
1	1	1



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Perceptron: Learning

EPOCH-1 EPOCH-2 EPOCH-3

Input			Target	Net Input	Calculated output	Weights		
x_1	x_2	x_0	t	y_{in}	y	w_1	w_2	w_0
						0	0	0
0	0	1	0	0	1	0	0	-1
0	1	1	1	-1	0	0	1	0
1	0	1	1	0	1	0	1	0
1	1	1	1	1	1	0	1	0
0	0	1	0	0	1	0	1	-1
0	1	1	1	0	1	0	1	-1
1	0	1	1	-1	0	1	1	0
1	1	1	1	2	1	1	1	0
0	0	1	0	0	1	1	1	-1
0	1	1	1	0	1	1	1	-1
1	0	1	1	0	1	1	1	-1
1	1	1	1	1	1	1	1	-1

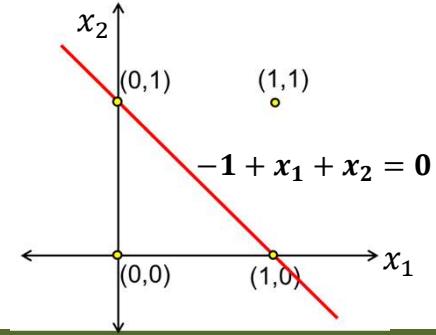
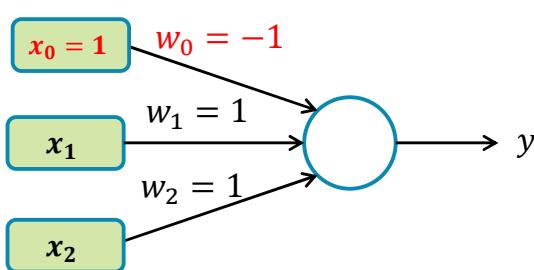
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Perceptron: Learning



EPOCH-4

Input			Target	Net Input	Calculated output	Weights		
x_1	x_2	x_0	t	y_{in}	y	w_1	w_2	w_0
						1	1	-1
0	0	1	0	-1	0	1	1	-1
0	1	1	1	0	1	1	1	-1
1	0	1	1	0	1	1	1	-1
1	1	1	1	1	1	1	1	-1



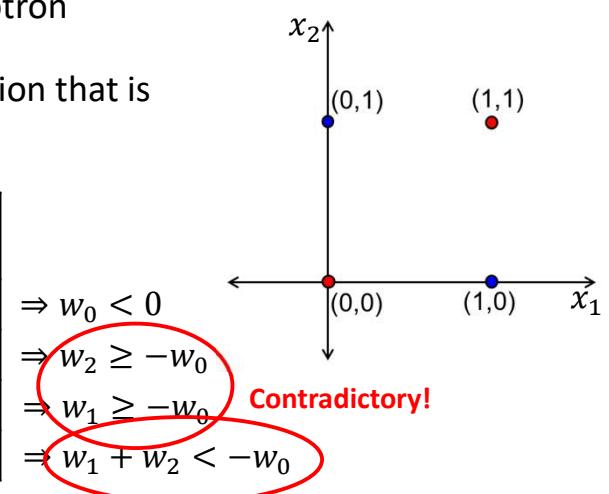
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Points to Remember



- Real valued inputs are allowed in perceptron
- A single perceptron cannot learn a function that is **not linearly separable**

x_1	x_2	Target (XOR)	Objective
0	0	0	$w_0 + w_1x_1 + w_2x_2 < 0$
0	1	1	$w_0 + w_1x_1 + w_2x_2 \geq 0$
1	0	1	$w_0 + w_1x_1 + w_2x_2 \geq 0$
1	1	0	$w_0 + w_1x_1 + w_2x_2 < 0$

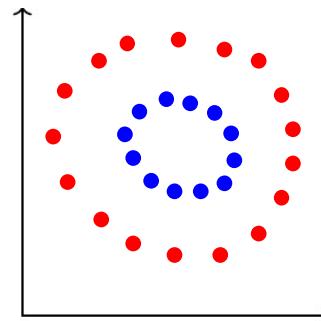
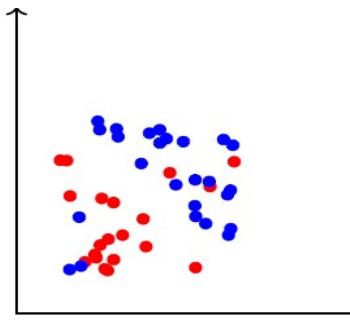


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Points to Remember



- Most real-world data is **not linearly separable** and will always contain some outliers



How do we implement functions that are not linearly separable ?

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Questions?

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