

Introduction to Artificial Neural Network Models

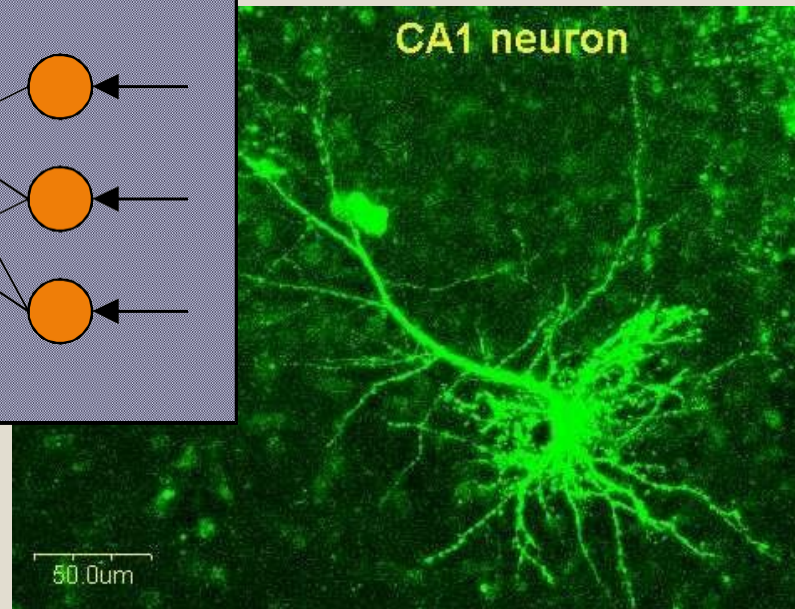
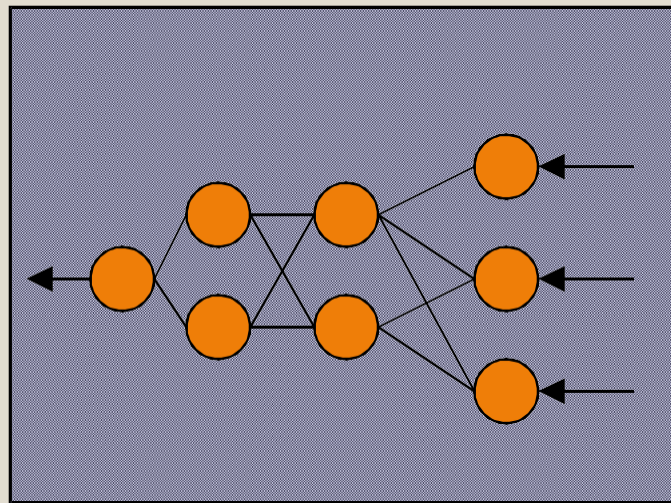


Image Source: www.physiol.ucl.ac.uk/fedwards/ca1%20neuron.jpg

Definition

Neural Network

A broad class of models that mimic functioning inside the human brain

There are various classes of NN models.

They are different from each other depending on

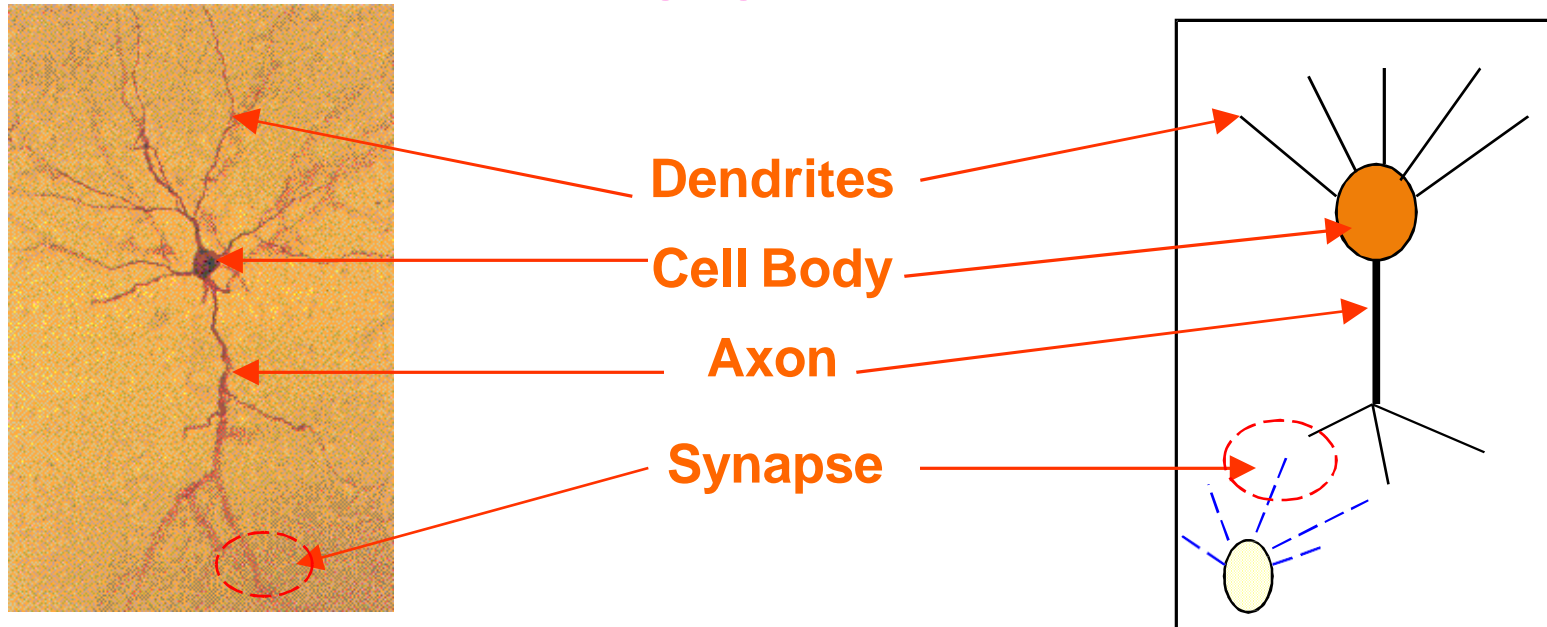
- ❑ Problem types
Prediction, Classification , Clustering
- ❑ Structure of the model
- ❑ Model building algorithm

For this discussion we are going to focus on

Feed-forward Back-propagation Neural Network
(used for Prediction and Classification problems)

A bit of biology . . .

Most important functional unit in human brain – a class of cells called –
NEURON



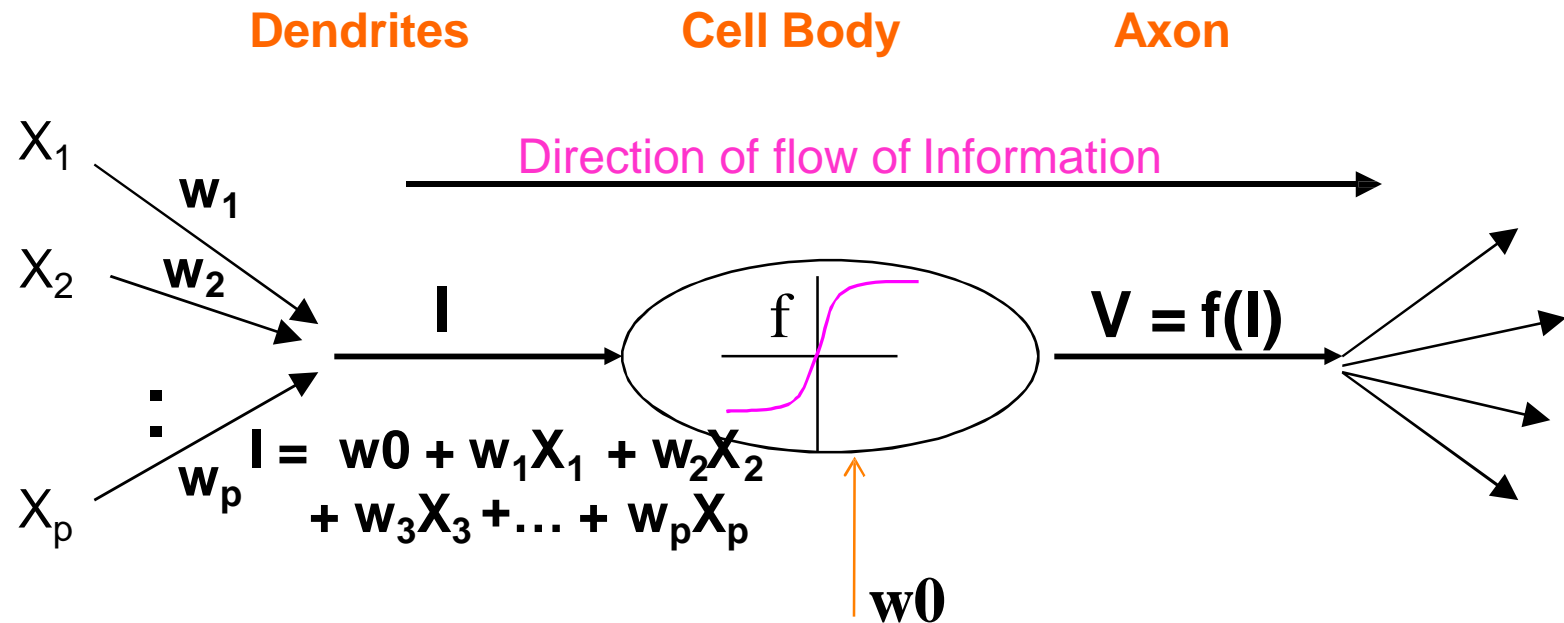
Hippocampal Neurons

Source: heart.cbl.utoronto.ca/~berj/projects.html

Schematic

- **Dendrites** – Receive information
- **Cell Body** – Process information
- **Axon** – Carries processed information to other neurons
- **Synapse** – Junction between Axon end and Dendrites of other Neurons

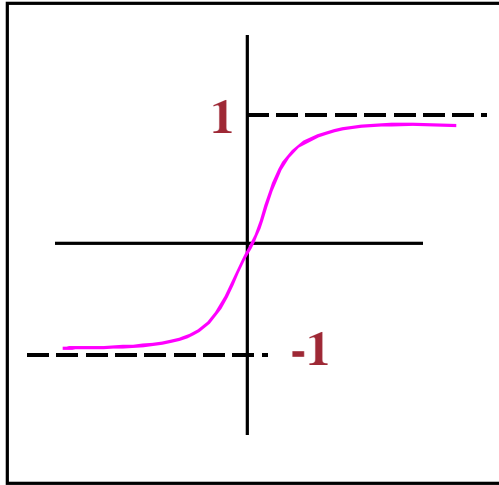
An Artificial Neuron



- Receives Inputs $X_1 X_2 \dots X_p$ from other neurons or environment
- Inputs fed-in through connections with 'weights'
- Total Input = Weighted sum of inputs from all sources
- Transfer function (Activation function) converts the input to output
- Output goes to other neurons or environment

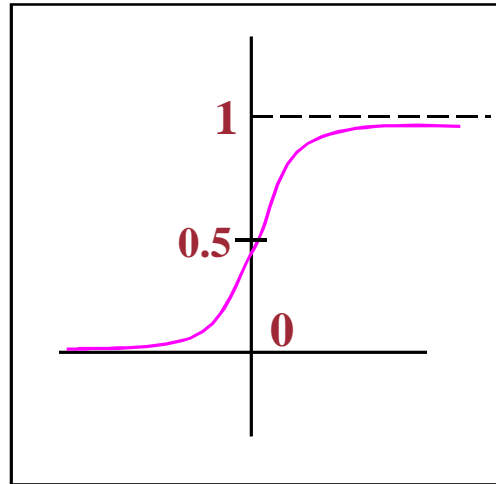
Transfer Functions

There are various choices for Transfer / Activation functions



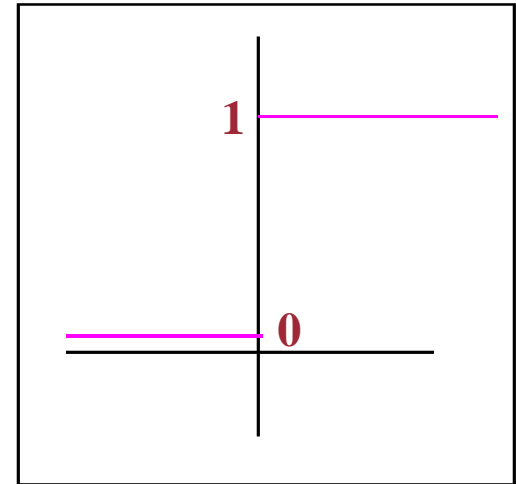
Tanh

$$f(x) = \frac{e^x - e^{-x}}{e^x + e^{-x}}$$



Logistic

$$f(x) = e^x / (1 + e^x)$$



Threshold

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

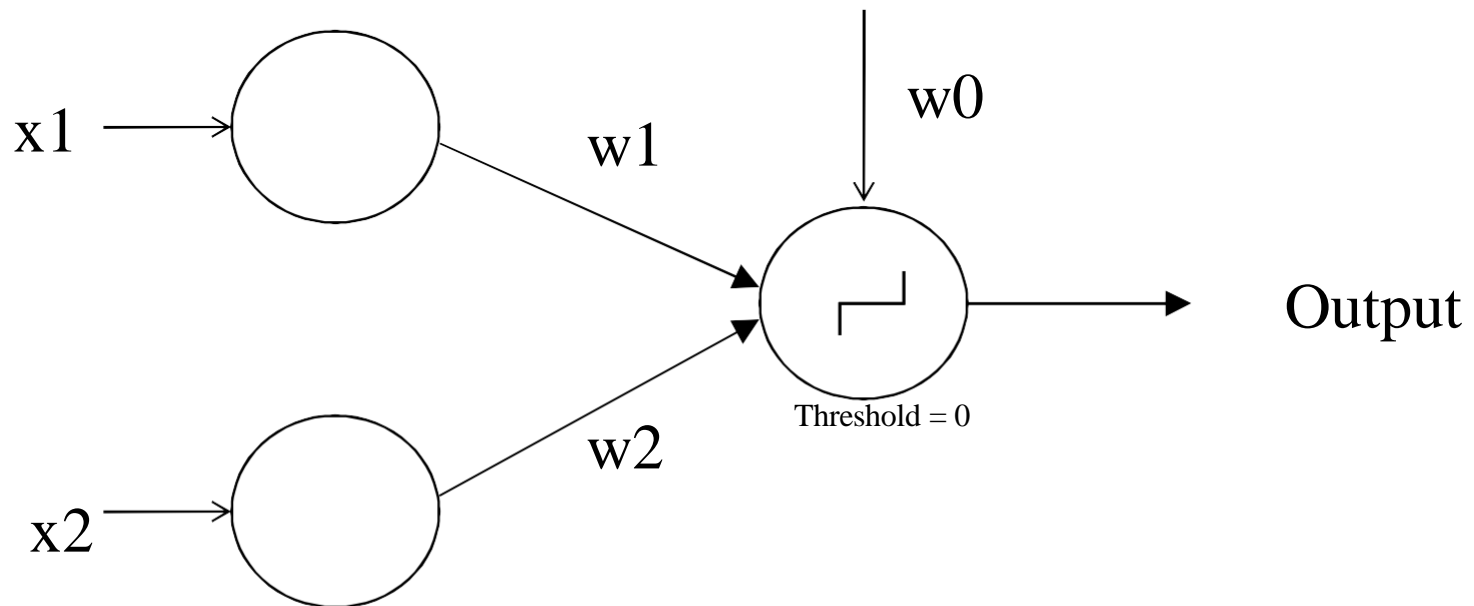
(Perceptron)

Linearly and non-linearly separable

Linearly separable – OR / AND

Linearly inseparable - XOR

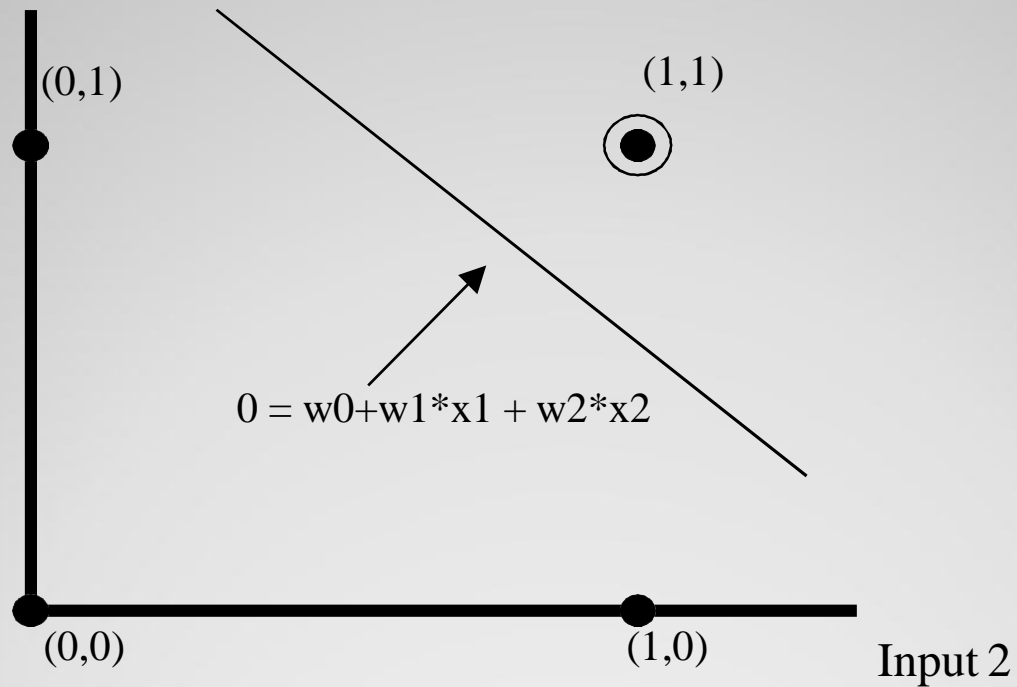
Single layer perceptron



□ Output space for AND gate

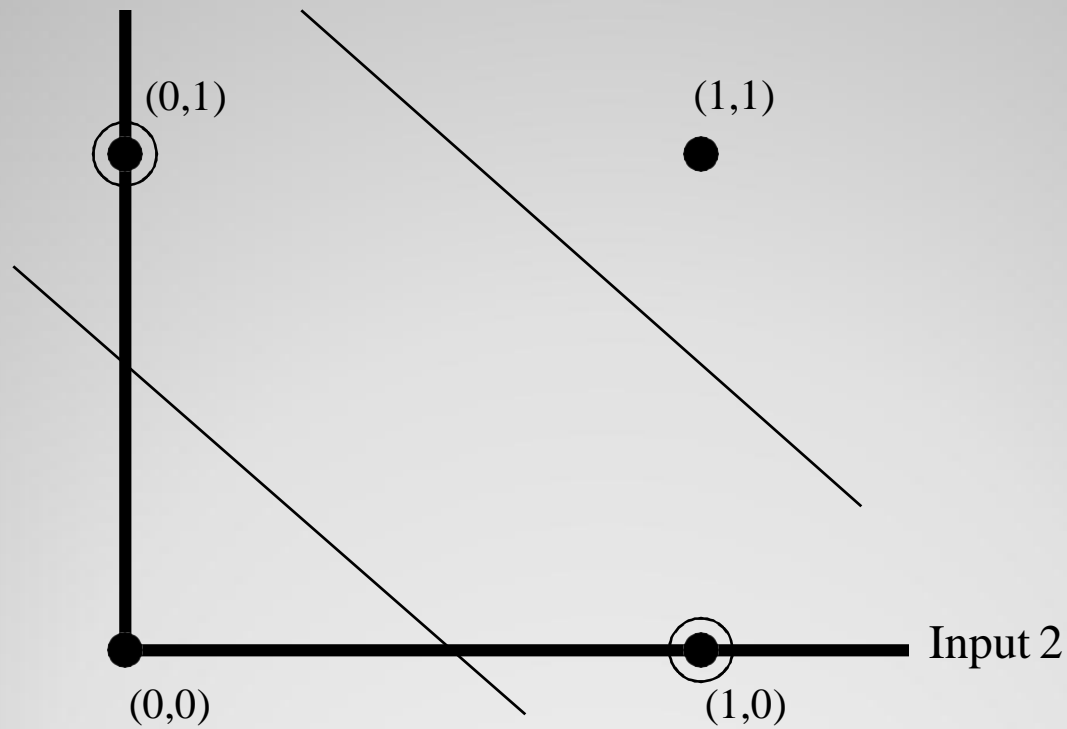
AND

Input 1



- Output space for XOR gate
- Demonstrates need for hidden layer

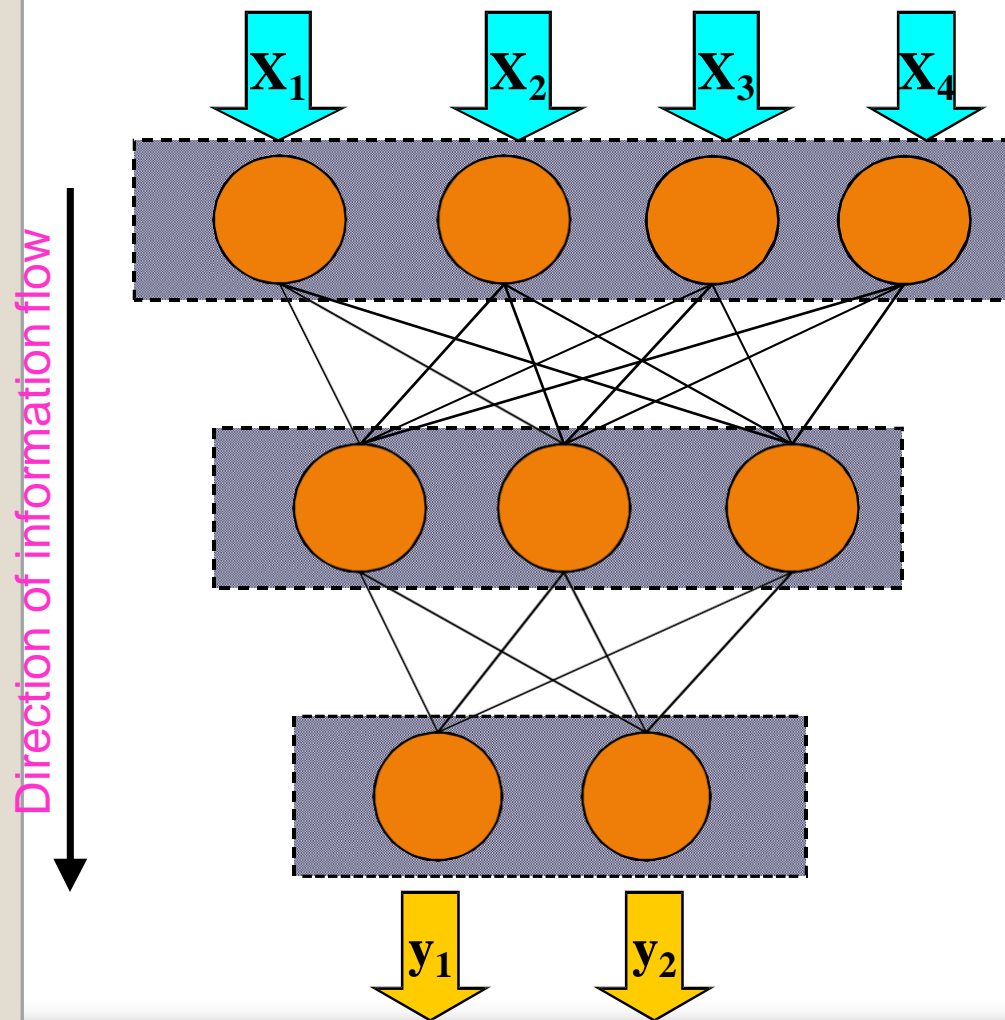
Input 1



XOR

ANN – Feed-forward Network

A collection of neurons form a 'Layer'



Input Layer

- Each neuron gets ONLY one input, directly from outside

Hidden Layer

- Connects Input and Output layers

Output Layer

- Output of each neuron directly goes to outside

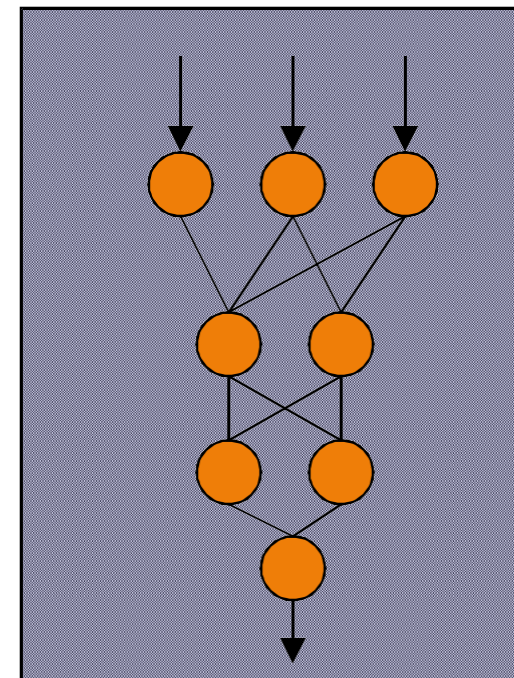
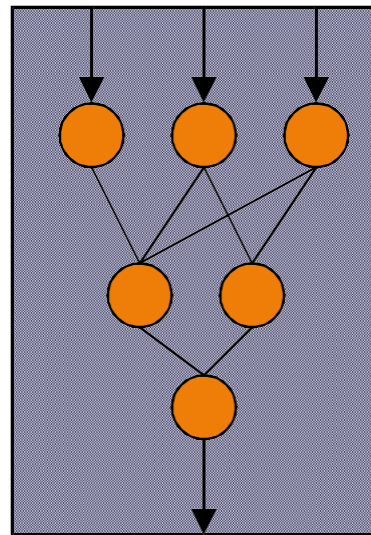
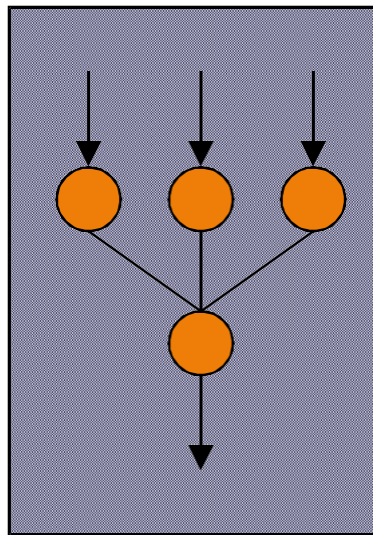
ANN – Feed-forward Network

Number of hidden layers can be

None

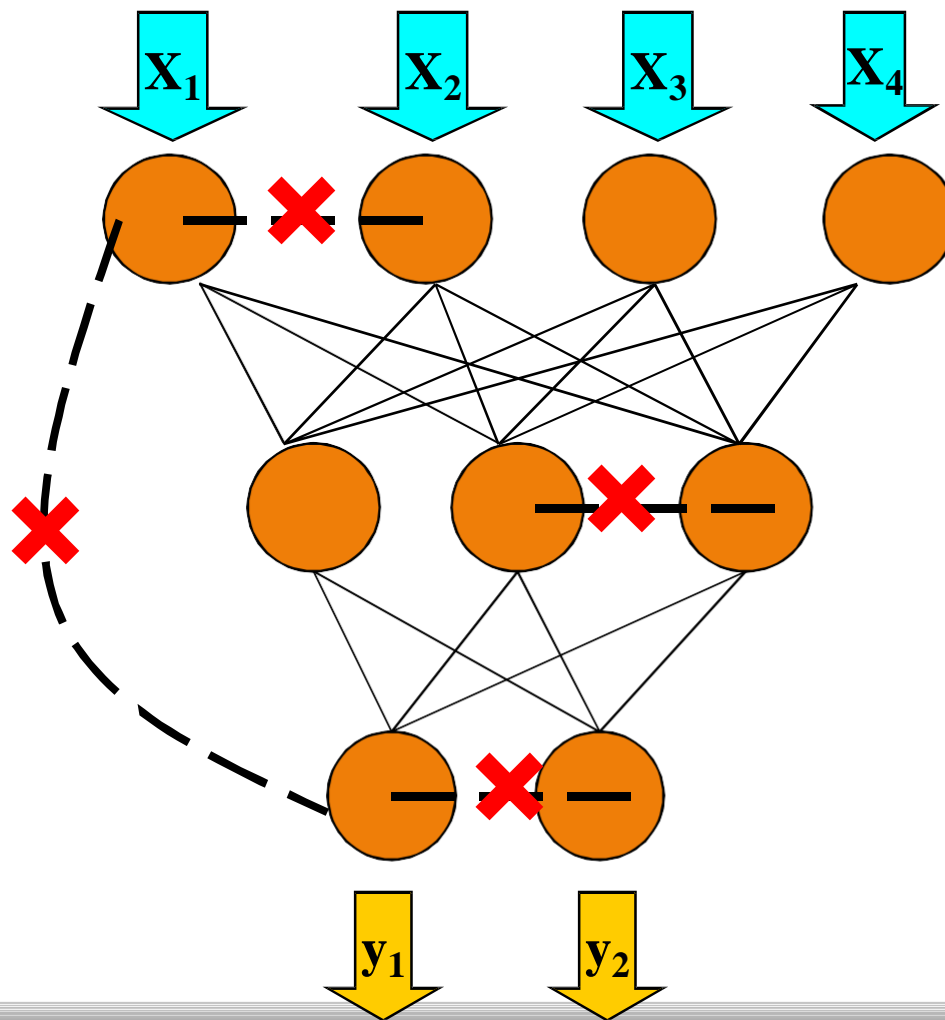
One

More



ANN – Feed-forward Network

Couple of things to note



- Within a layer neurons are NOT connected to each other.

- Neuron in one layer is connected to neurons ONLY in the NEXT layer. (Feed-forward)

- Jumping of layer is NOT allowed

One particular ANN model

What do we mean by 'A particular Model' ?

Input: $X_1 X_2 X_3$ Output: Y Model: $Y = f(X_1 X_2 X_3)$

For an ANN : Algebraic form of $f(.)$ is too complicated to write down.

However it is characterized by

- # Input Neurons
- # Hidden Layers
- # Neurons in each Hidden Layer
- # Output Neurons
- WEIGHTS for all the connections

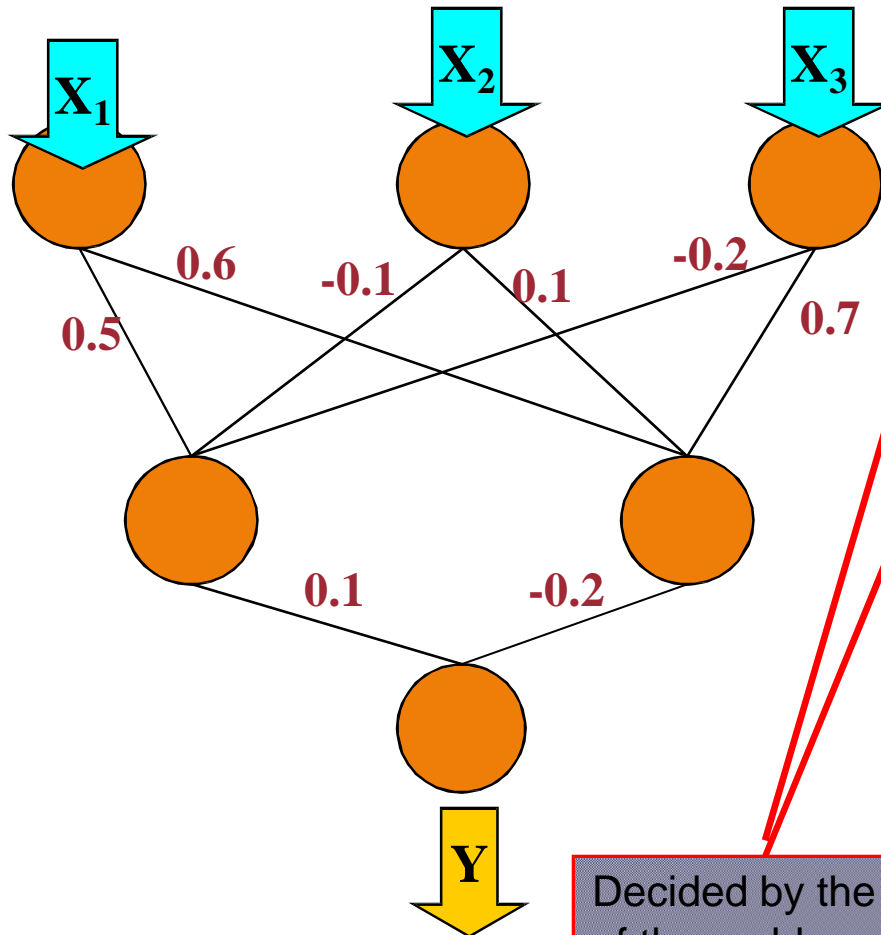
'Fitting' an ANN model = Specifying values for all those parameters

One particular Model – an Example

Input: $X_1 X_2 X_3$

Output: Y

Model: $Y = f(X_1 X_2 X_3)$



Parameters	Example
# Input Neurons	3
# Hidden Layers	1
# Hidden Layer Size	3
# Output Neurons	3
Weights	Specified

Decided by the structure
of the problem
Input Nrns = # of X 's
Output Nrns = # of Y 's

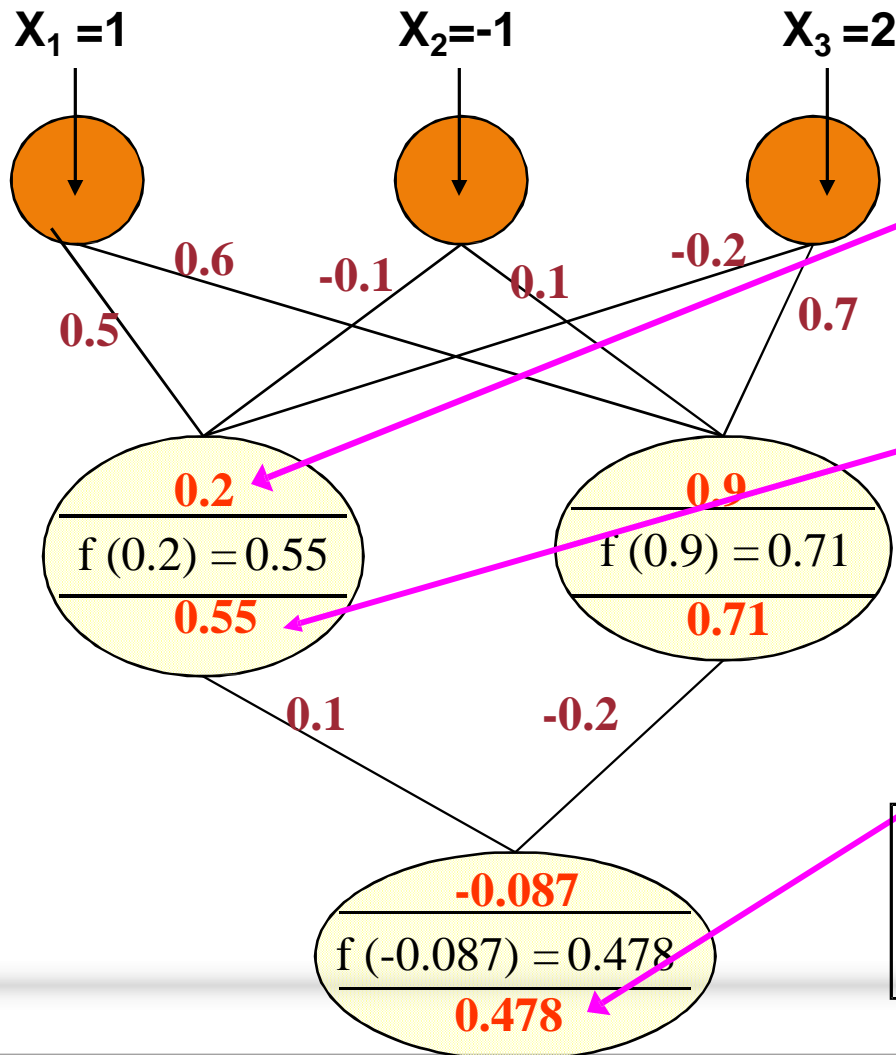
Free parameters

Prediction using a particular ANN Model

Input: $X_1 X_2 X_3$

Output: Y

Model: $Y = f(X_1 X_2 X_3)$



$$0.2 = 0.5 * 1 - 0.1 * (-1) - 0.2 * 2$$

$$f(x) = e^x / (1 + e^x)$$
$$f(0.2) = e^{0.2} / (1 + e^{0.2}) = 0.55$$

Predicted $Y = 0.478$

Suppose Actual $Y = 2$
Then
Prediction Error = $(2 - 0.478) = 1.522$