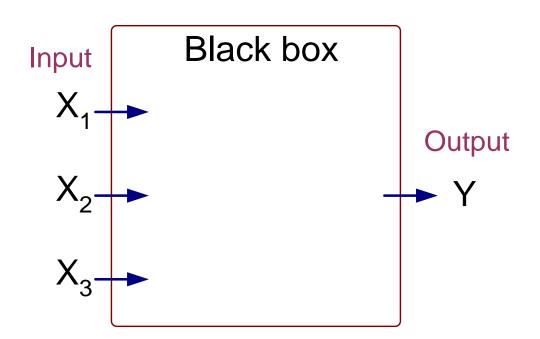
## **Artificial Neural Networks (ANN)**

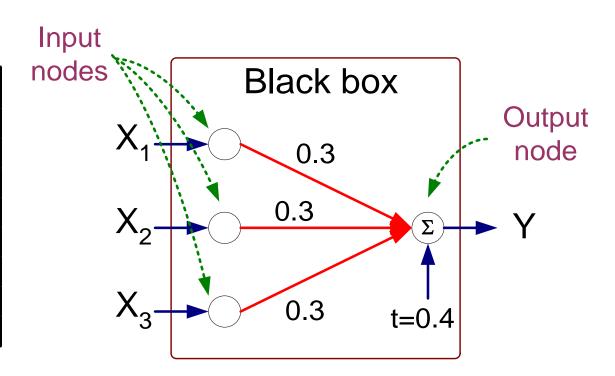
X <sub>1</sub>	$X_2$	$X_3$	Y
1	0	0	0
1	0	1	1
1	1	0	1
1	1	1	1
0	0	1	0
0	1	0	0
0	1	1	1
0	0	0	0



Output Y is 1 if at least two of the three inputs are equal to 1.

### **Artificial Neural Networks (ANN)**

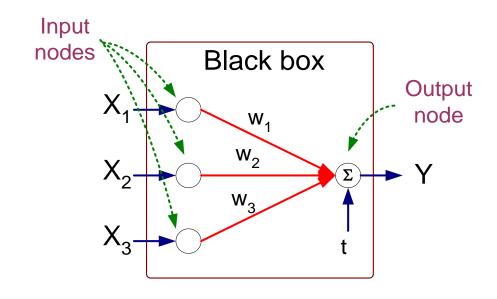
X <sub>1</sub>	$X_2$	X <sub>3</sub>	Υ
1	0	0	0
1	0	1	1
1	1	0	1
1	1	1	1
0	0	1	0
0	1	0	0
0	1	1	1
0	0	0	0



$$Y = I(0.3X_1 + 0.3X_2 + 0.3X_3 - 0.4 > 0)$$
where  $I(z) = \begin{cases} 1 & \text{if } z \text{ is true} \\ 0 & \text{otherwise} \end{cases}$ 

# **Artificial Neural Networks (ANN)**

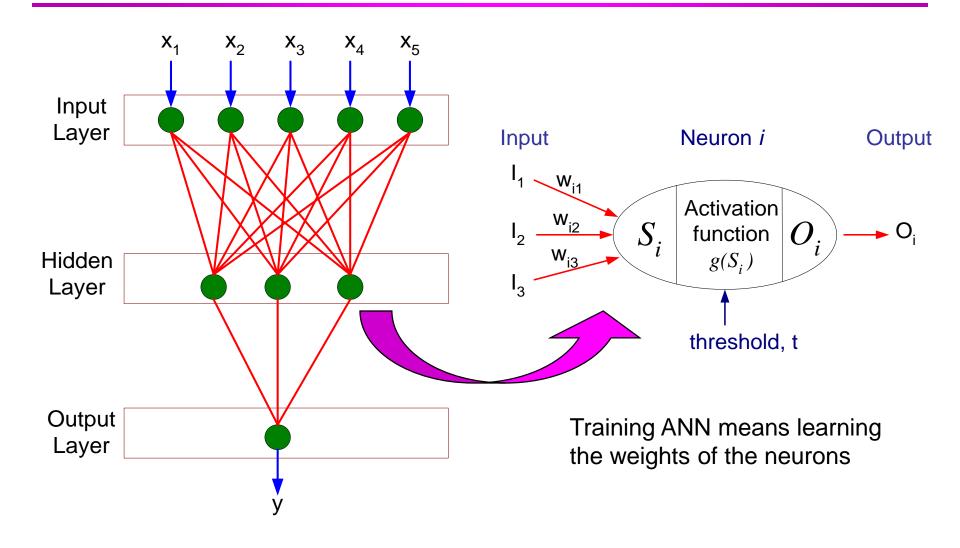
- Model is an assembly of inter-connected nodes and weighted links
- Output node sums up each of its input value according to the weights of its links
- Compare output node against some threshold t



#### **Perceptron Model**

$$Y = I(\sum_{i} w_{i}X_{i} - t)$$
 or 
$$Y = sign(\sum_{i} w_{i}X_{i} - t)$$

### **General Structure of ANN**



# **Algorithm for learning ANN**

- Initialize the weights (w<sub>0</sub>, w<sub>1</sub>, ..., w<sub>k</sub>)
- Adjust the weights in such a way that the output of ANN is consistent with class labels of training examples
  - Objective function:  $E = \sum_{i} [Y_i f(w_i, X_i)]^2$
  - Find the weights w<sub>i</sub>'s that minimize the above objective function
    - e.g., backpropagation algorithm (see lecture notes)