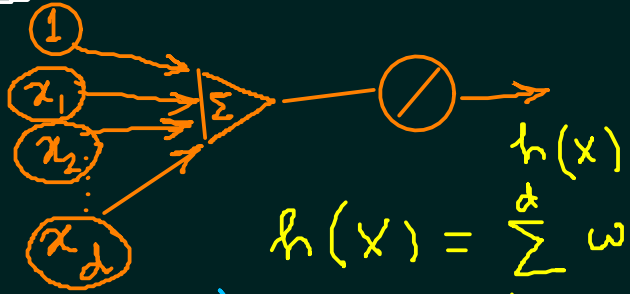
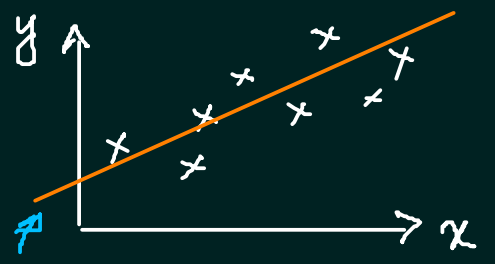


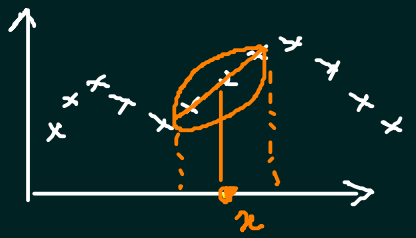
# Linear Regression:

## SUMMARY

(Extn.) Locally weighted Regression



$$h(x) = \sum_{i=0}^d w_i x_i$$



$$h(x) = \sum_{i=0}^d \alpha_i w_i x_i$$

where  $\alpha_i = e^{-\frac{(x_i - x)^2}{2\tau^2}}$

$$E_{in}(w) = \frac{1}{N} \sum_{n=1}^N (y_n - \sum_{i=0}^d w_i x_{ni})^2$$

Minimize  $E_{in}(w) \Rightarrow w = (X^T X)^{-1} X^T y$

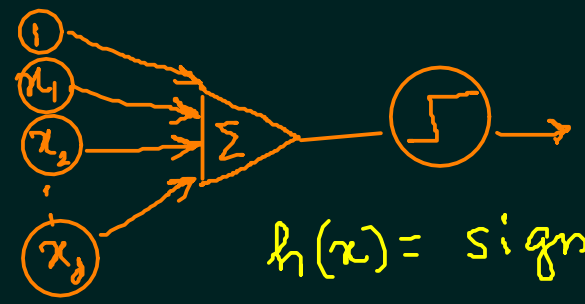
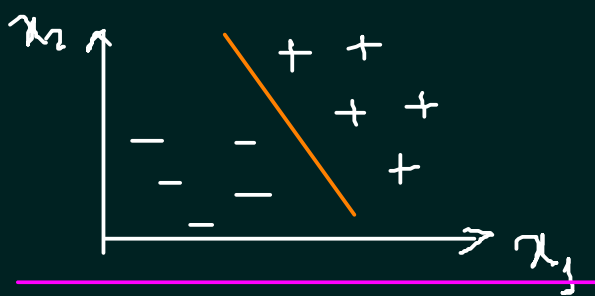
One step learning  
 Optimization Problem (Sum Sq. Error)

# Linear Classification:

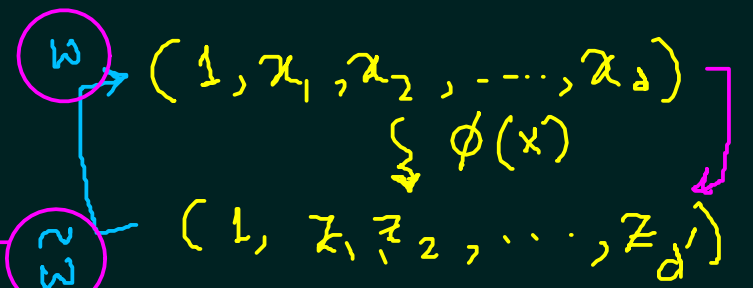
$$f: X \rightarrow Y = \{-1, +1\}$$

(Extn.)

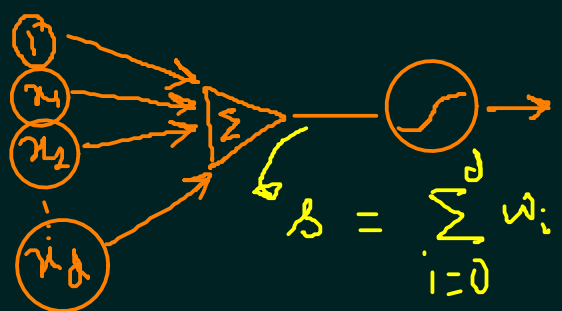
Non-linear Transform



$$h(x) = \text{sign}\left(\sum_{i=0}^d w_i x_i\right)$$



# Logistic Regression:



$$h(x) = \Theta(s) = \frac{1}{1 + e^{-s}}$$

[Sigmoid]

$$E_{in}(w) = \frac{1}{N} \sum_{n=1}^N \ln(1 + e^{-y_n w^T x_n})$$

[Cross-entropy error]  $e(h(x_n), y_n)$

Gradient Descent  
 Batch / One step  
 Stochastic / Incremental

- Draws the analogy from Biology

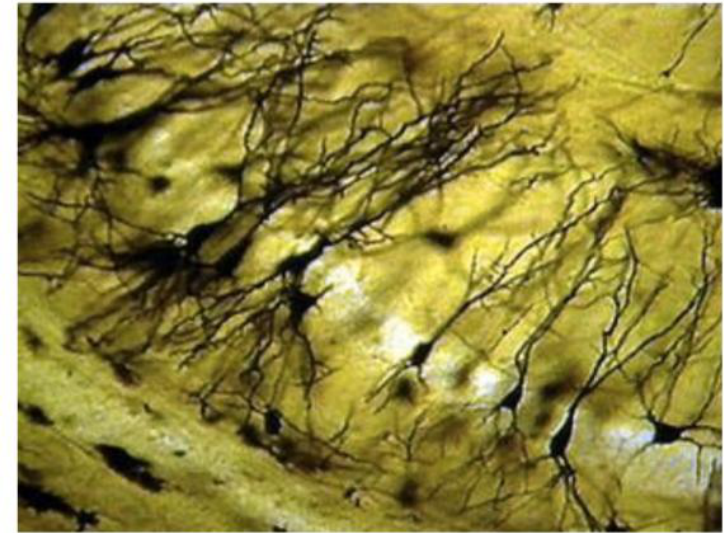
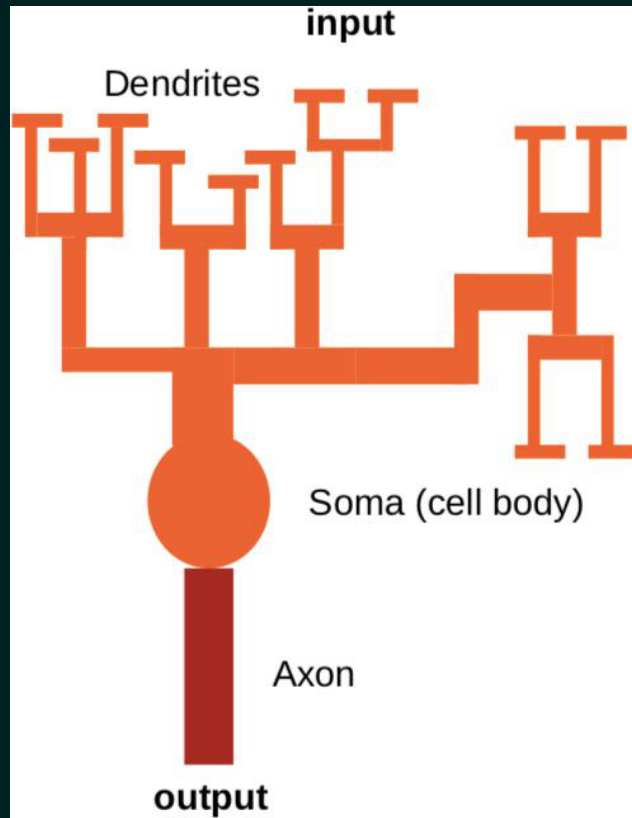
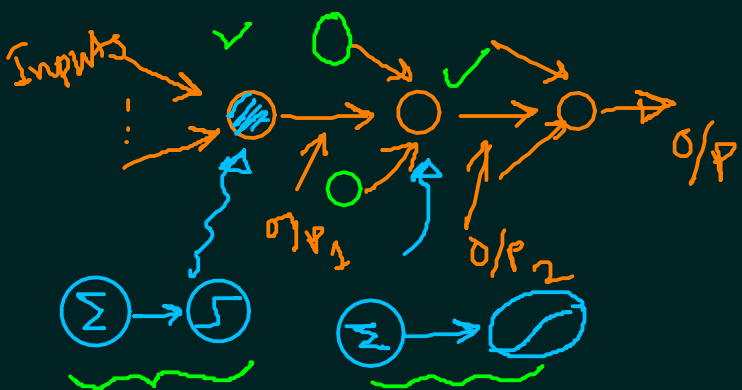
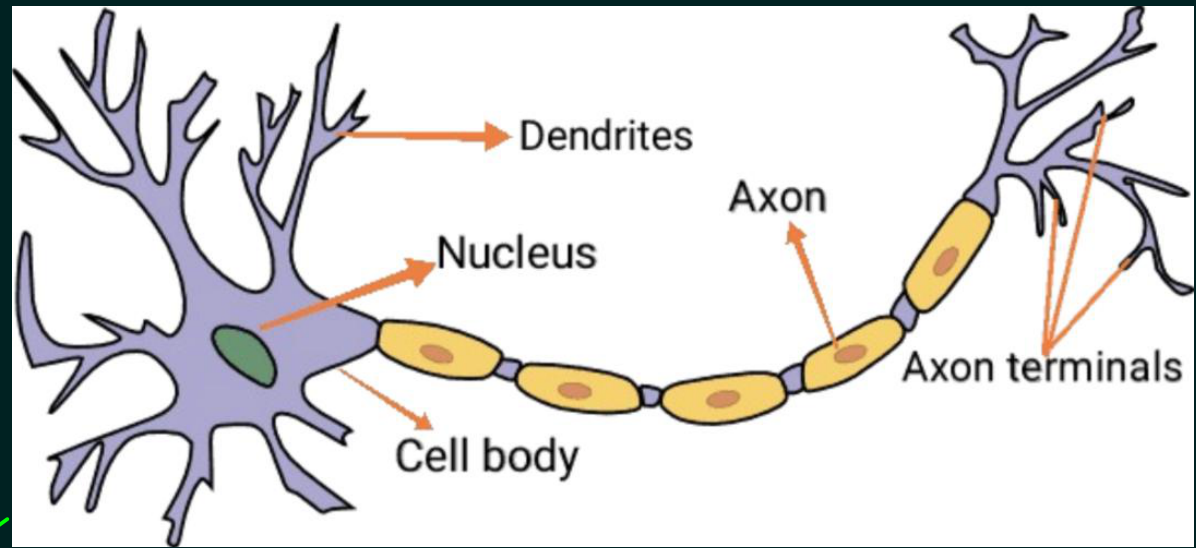
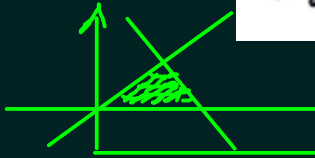
$10^{11}$  neurons

→ each neuron →  $10^4$  others (Connected)

▷ Recognize  $10^{-1}$  sec

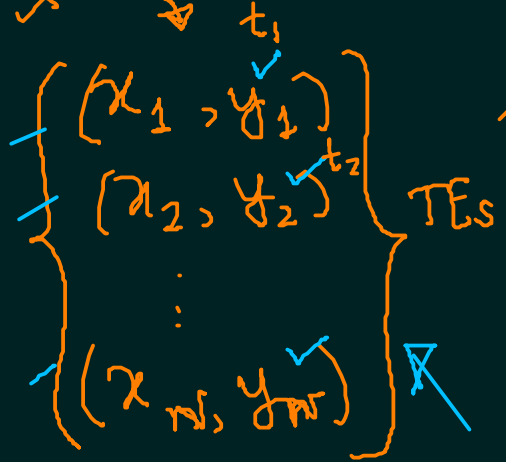
→  $10^{-3}$  switching time

▷  $\geq 100$  neurons (used)



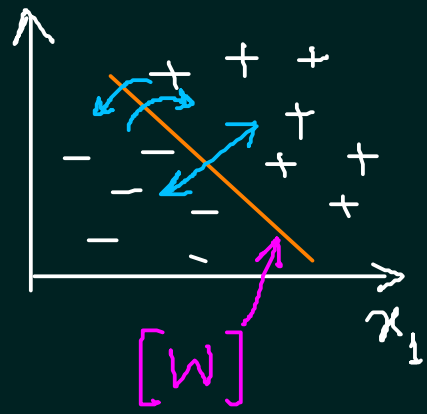
▷ Linear classification: Perceptron (linear discriminator)

$$f: X \rightarrow Y \begin{cases} \rightarrow +1 \\ \rightarrow -1 \end{cases}$$



$$s = \sum_{i=0}^d w_i x_i \quad (x_0 = 1)$$

$$h(x) = \text{sign} \left( \sum_{i=0}^d w_i x_i \right)$$



Goal:

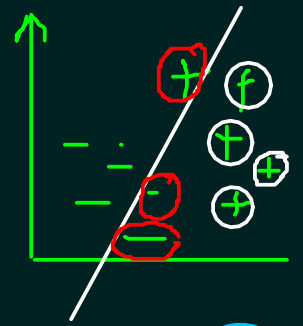
$$x^{\text{new}} = [x_1^{\text{new}} \dots x_d^{\text{new}}]$$

$$y^{\text{new}} = + / -$$

← classify

$(x_n, y_n)$  → Min

$$E_{in}(w) = \frac{1}{2} \sum_{i=1}^n \left[ y_n - \text{sign} \left( \sum_{i=0}^d w_i x_{ni} \right) \right]^2$$



$$0 = \nabla E_{in}(w) = \left[ \frac{\partial E_{in}}{\partial w_0}, \frac{\partial E_{in}}{\partial w_1}, \dots, \frac{\partial E_{in}}{\partial w_d} \right]$$

$$\frac{\partial E_{in}}{\partial w_k} = \sum_{n=1}^N (y_n - \text{sign} \left( \sum_{i=0}^d w_i x_{ni} \right)) (-\text{sign} \left( x_{nk} \right))$$

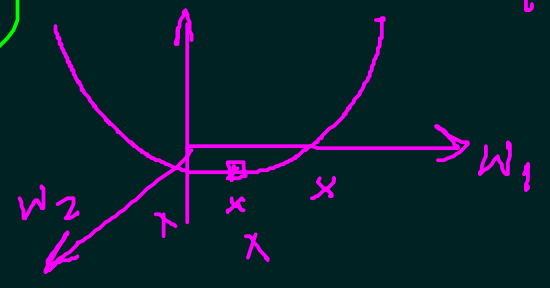
$$w_i \leftarrow w_i + \Delta w_i$$

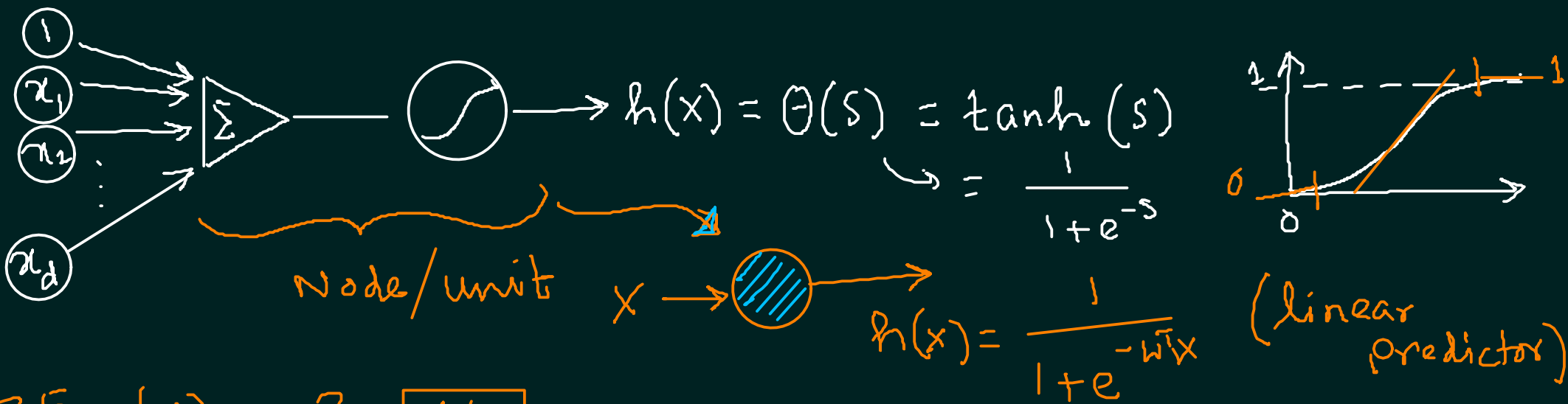
$$\Delta w_i = -\eta \frac{\partial E_{in}}{\partial w_i}$$

(Batch) G.D.

$$w_k \leftarrow w_k + \eta \sum_{n=1}^N (t_n - o_n) (x_{nk})$$

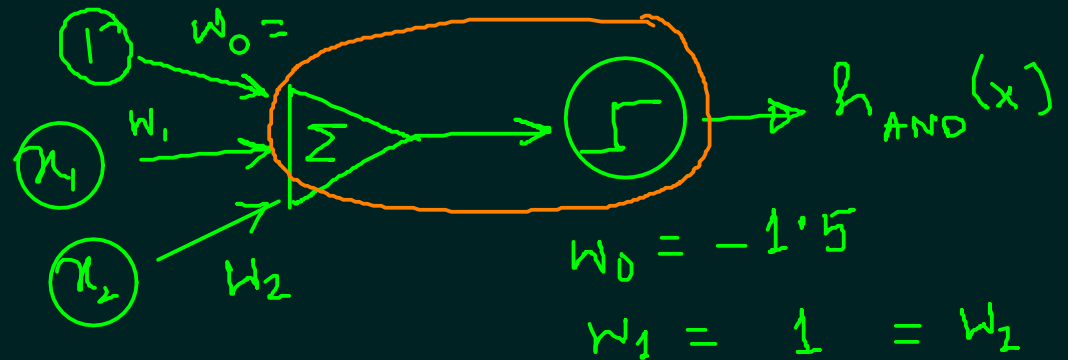
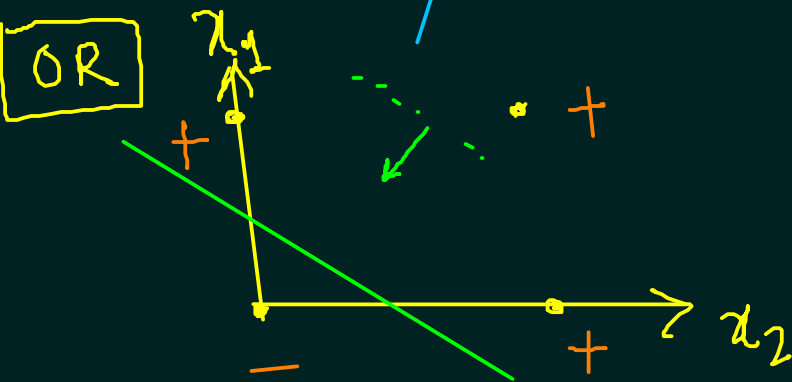
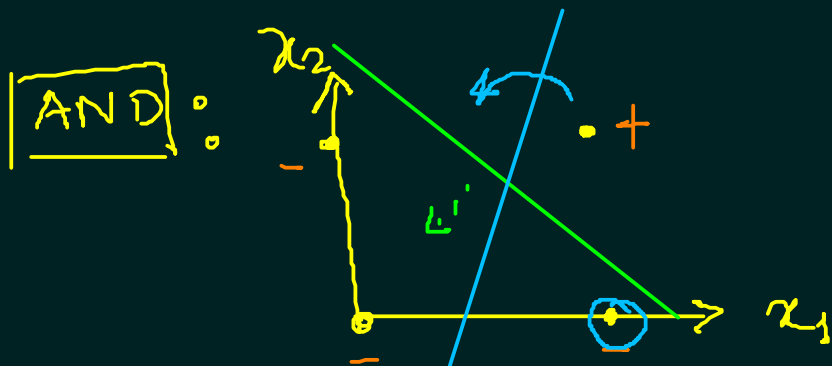
(Delta)





$\nabla E_{in}(w) = ?$  H/w

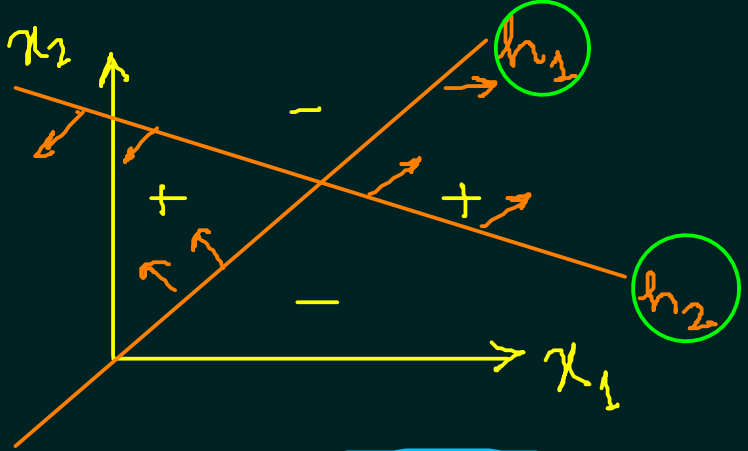
(Artificial Neural Node)



$h_1 = (w_0 + w_1 x_1 + w_2 x_2) \rightarrow \begin{matrix} + \\ - \end{matrix}$

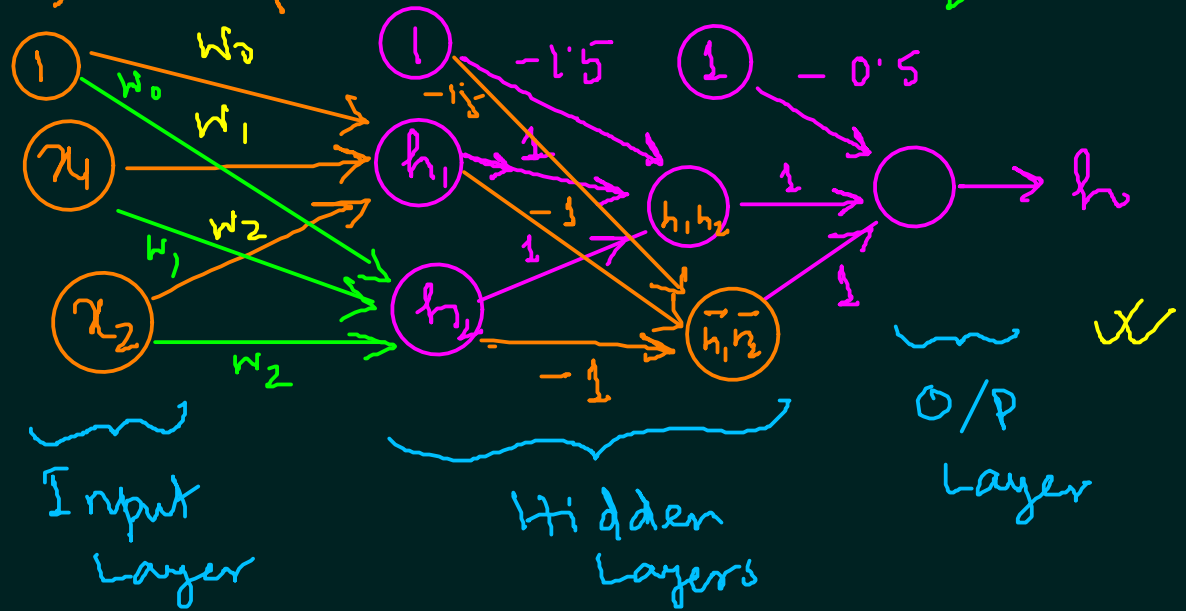
$h_2 = (w_0 + w_1 x_1 + w_2 x_2)$   $w_0 = -0.5$   
 $w_1 = 1 = w_2$

$w_i \leftarrow w_i - \eta \frac{\partial E_{in}}{\partial w_i}$



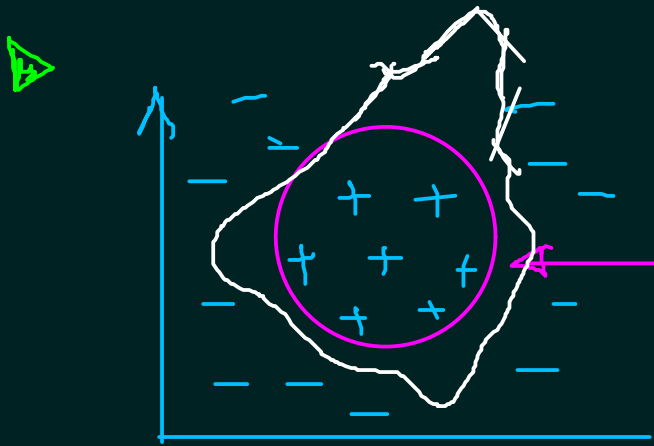
$$h = \overbrace{h_1 \cdot h_2}^{+ \text{ sign}} + \overbrace{h_1 \cdot h_2}^{- \text{ sign}}$$

OR



MULTI-LAYER PERCEPTRON

Neural Network (feed-forward NN)



Question: How to categorize +'s from -'s??  
(Think!)

NN →

