

Neural Networks: Non-linear hypotheses

typically have more features than just Z

if $n=100$, 2nd order terms $\rightarrow \approx 5000$ features $O(n^2) \approx \frac{n^2}{2}$

subset

3rd order terms $O(n^3) \approx 170,000$ features

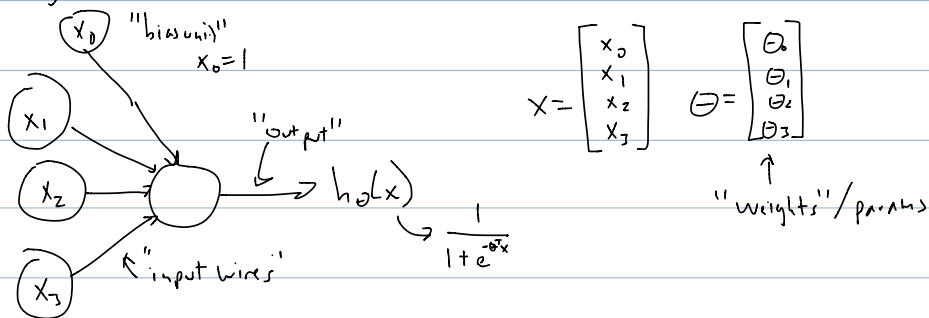
Neurons and the brain

algorithms to try to mimic the brain

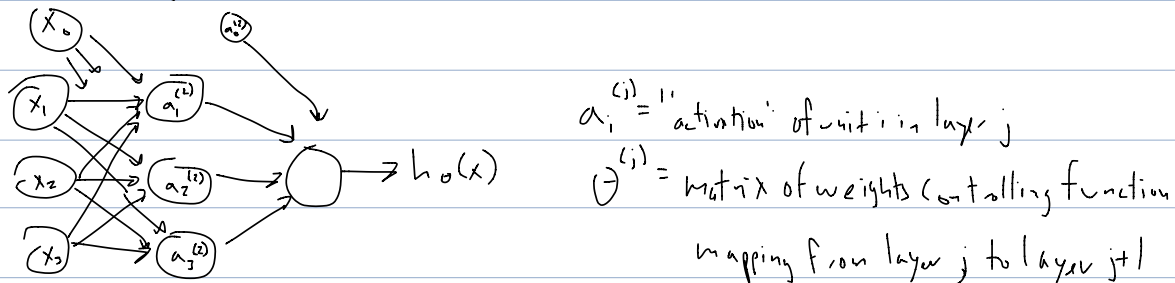
"one learning algorithm" hypothesis

Model Representation

Logistic unit



Sigmoid (logistic) activation function



Layer 1 "input" Layer 2 "hidden" Layer 3 "output"

$$a_1^{(2)} = \sigma(\Theta_{10}^{(1)} x_0 + \Theta_{11}^{(1)} x_1 + \Theta_{12}^{(1)} x_2 + \Theta_{13}^{(1)} x_3)$$

$$a_2^{(2)} = g(\theta_{20}^{(2)} x_0, \dots)$$

$$h_\theta(x) = a_1^{(3)} = g(\theta_{10}^{(2)} a_0^{(2)} + \theta_{11}^{(2)} a_1^{(2)} + \theta_{12}^{(2)} a_2^{(2)} + \theta_{13}^{(2)} a_3^{(2)})$$

If network has s_j units in layer j , s_{j+1} units in layer $j+1$, then $\theta^{(j)}$ will be of dimension $s_{j+1} \times (s_j + 1)$

Model Representation II

$$X = \begin{bmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \end{bmatrix} \quad Z^{(2)} = \begin{bmatrix} z_1^{(2)} \\ z_2^{(2)} \\ z_3^{(2)} \end{bmatrix}$$

$$a_1^{(2)} = g(z_1^{(2)})$$

$$Z^{(2)} = \Theta^{(1)} X$$

$$a_2^{(2)} = g(z_2^{(2)})$$

$$a^{(2)} = g(Z^{(2)})$$

$$a_3^{(2)} = g(z_3^{(2)})$$

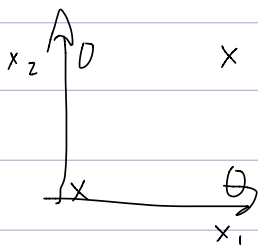
$$\mathbb{R}^3 \quad \text{Add } a_0^{(2)} = 1 \rightarrow a^{(2)} \in \mathbb{R}^4$$

$$Z^{(3)} = \Theta^{(2)} a^{(2)}$$

$$h_\theta(x) = a^{(3)} = g(Z^{(3)})$$

Examples and Intuition

x_1, x_2 are binary (0 or 1)

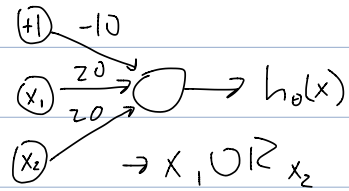
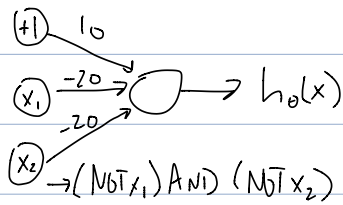
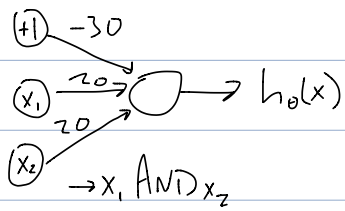


$$y = x_1 \text{ XOR } x_2$$

$$\rightarrow x_1 \text{ XOR } x_2$$

$$\rightarrow \text{NOT}(x_1 \text{ XOR } x_2)$$

Examples and Intuition II



Multi-class Classification

One vs. all