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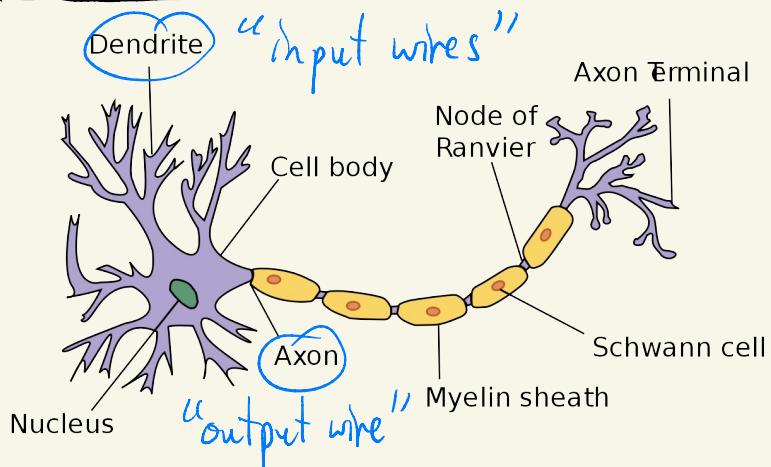
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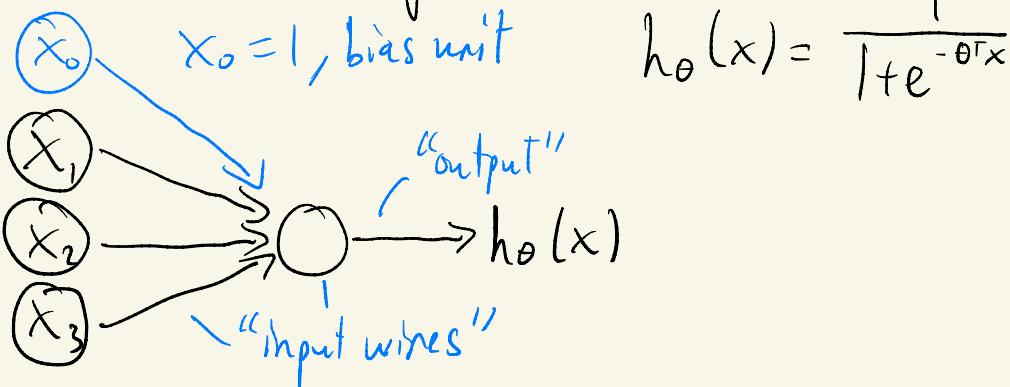
# Nerve Cell (Neuron);

May 18, 2021

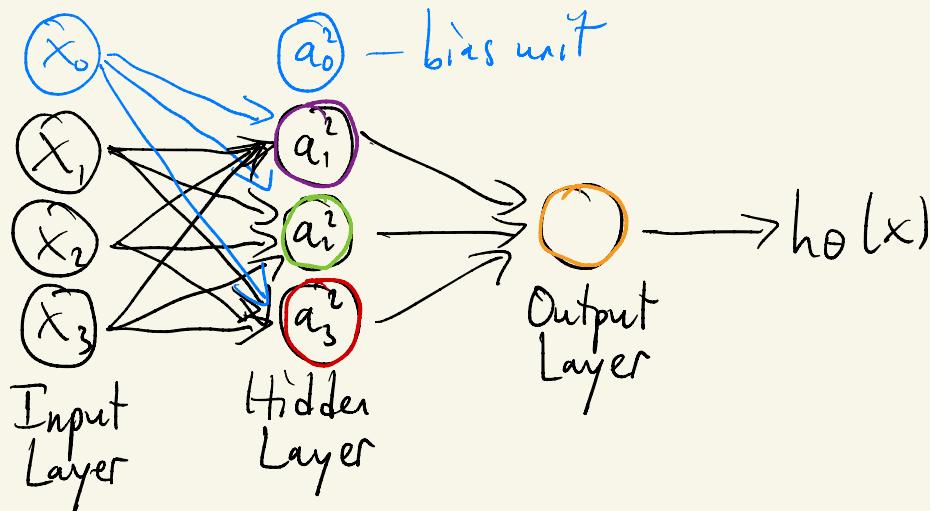


Neurons send signals via electrical signals from their output wires to another neuron's input wires,

# Neuron Model (Logistic Unit);



# Neural Network:



$a_i^{(j)}$  - "activation" of unit  $i$  in layer  $j$

$\Theta^{(j)}$  - matrix of weights controlling function mapping from layer  $j$  to  $j+1$

$$\begin{aligned} \text{i.e.,} \\ a_1^{(2)} &= g(\underline{\Theta_{10}^{(1)} x_0 + \Theta_{11}^{(1)} x_1 + \Theta_{12}^{(1)} x_2 + \Theta_{13}^{(1)} x_3}) \quad \underbrace{\hspace{10em}}_{z_1^{(2)}} \\ a_2^{(2)} &= g(\underline{\Theta_{20}^{(1)} x_0 + \Theta_{21}^{(1)} x_1 + \Theta_{22}^{(1)} x_2 + \Theta_{23}^{(1)} x_3}) \quad \underbrace{\hspace{10em}}_{z_2^{(2)}} \\ a_3^{(2)} &= g(\underline{\Theta_{30}^{(1)} x_0 + \Theta_{31}^{(1)} x_1 + \Theta_{32}^{(1)} x_2 + \Theta_{33}^{(1)} x_3}) \quad \underbrace{\hspace{10em}}_{z_3^{(2)}} \\ h_\theta(x) = a_1^{(3)} &= g(\underline{\Theta_{10}^{(2)} a_0^{(2)} + \Theta_{11}^{(2)} a_1^{(2)} + \Theta_{12}^{(2)} a_2^{(2)} + \Theta_{13}^{(2)} a_3^{(2)}}) \end{aligned}$$

Shape  $\Theta^j = (\# \text{hidden layer neurons}) \times (\# \text{inputs} + 1)$

i.e,

$$\Theta^{(1)} = \begin{bmatrix} \theta_{10} & \theta_{11} & \theta_{12} & \theta_{13} \\ \theta_{20} & \theta_{21} & \theta_{22} & \theta_{23} \\ \theta_{30} & \theta_{31} & \theta_{32} & \theta_{33} \end{bmatrix}$$

shape =  $S_{j+1} \times (S_j + 1)$

where  $S$  = # of units

$3 \times 4$

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

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

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

Add  $a_0^{(2)} = 1$ ,

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

$$h_\Theta(X) = a^{(3)} = g(Z^{(3)})$$

\* This is known as  
forward propagation

\* Compute activations from input, hidden, and output layers.

## Notice the Similarity!

$$h_{\theta}(x) = g(\theta_{0,0}^{(2)} a_0 + \theta_{1,0}^{(2)} a_1 + \theta_{2,0}^{(2)} a_2 + \theta_{3,0}^{(2)} a_3)$$

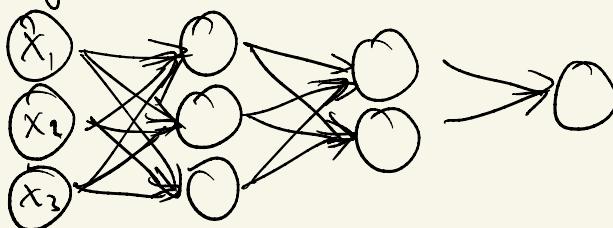
↳ like logistic regression, but uses  $a_0 \dots a_3$  instead of  $x_0 \dots x_3$

\* Neural networks are not constrained to feeding  $x_0 \dots x_3$  to logistic regression, but can feed its own learned features  $a_0 \dots a_3$  into logistic regression.

↳ Very flexible, can learn some very complex features

The neural network's **architecture** refers to how the neurons connect to each other

e.g.



Different architecture from before.