

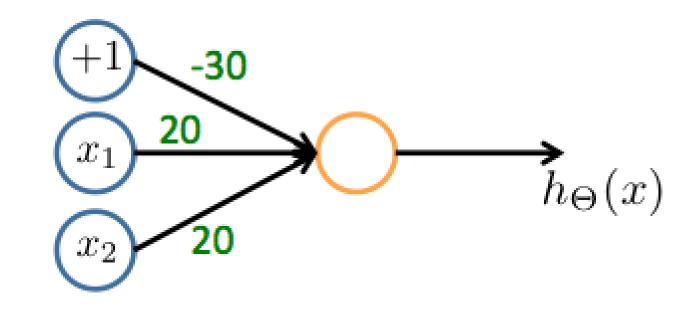
1 point

- 1. Which of the following statements are true? Check all that apply.
  - Any logical function over binary-valued (0 or 1) inputs  $x_1$  and  $x_2$  can be (approximately) represented using some neural network.
  - activation function applied at every layer, are always in the range (0, 1).

The activation values of the hidden units in a neural network, with the sigmoid

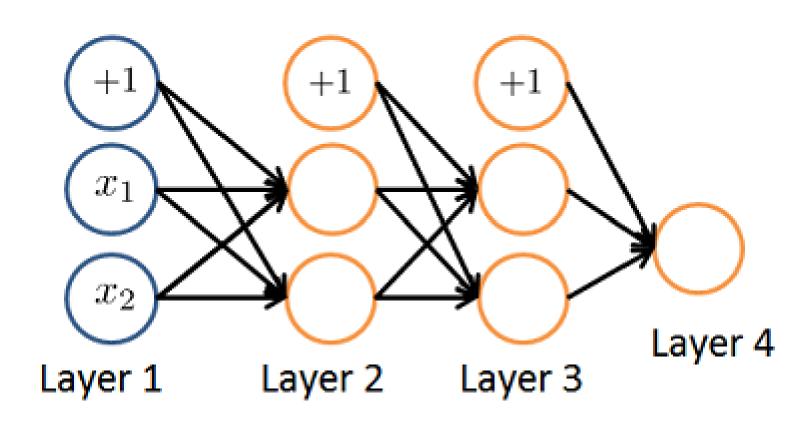
- A two layer (one input layer, one output layer; no hidden layer) neural network can represent the XOR function.
- Suppose you have a multi-class classification problem with three classes, trained with a 3 layer network. Let  $a_1^{(3)}=(h_\Theta(x))_1$  be the activation of the first output unit, and similarly  $a_2^{(3)}=(h_\Theta(x))_2$  and  $a_3^{(3)}=(h_\Theta(x))_3$ . Then for any input x, it must be the case that  $a_1^{(3)}+a_2^{(3)}+a_3^{(3)}=1$ .

1 point 2. Consider the following neural network which takes two binary-valued inputs  $x_1,x_2\in\{0,1\}$  and outputs  $h_\Theta(x)$ . Which of the following logical functions does it (approximately) compute?



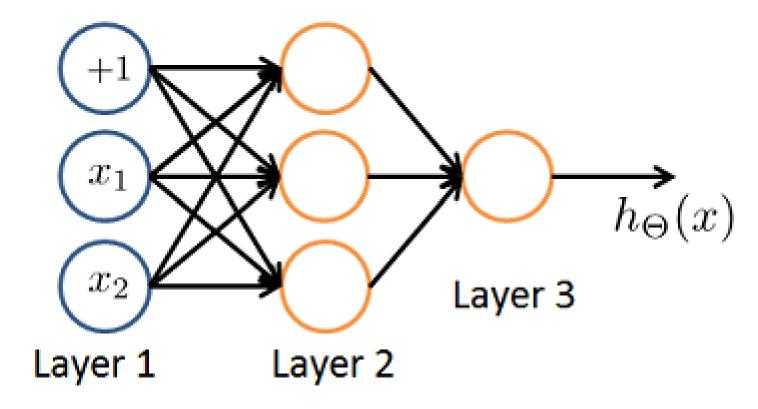
- AND
- NAND (meaning "NOT AND")
- OR
- XOR (exclusive OR)

1 point 3. Consider the neural network given below. Which of the following equations correctly computes the activation  $a_1^{(3)}$ ? Note: g(z) is the sigmoid activation function.



- $a_1^{(3)} = g(\Theta_{1,0}^{(1)}a_0^{(1)} + \Theta_{1,1}^{(1)}a_1^{(1)} + \Theta_{1,2}^{(1)}a_2^{(1)})$
- $a_1^{(3)} = g(\Theta_{1,0}^{(1)}a_0^{(2)} + \Theta_{1,1}^{(1)}a_1^{(2)} + \Theta_{1,2}^{(1)}a_2^{(2)})$
- The activation  $a_1^{(3)}$  is not present in this network.

1 point 4. You have the following neural network:



the following Octave code:

You'd like to compute the activations of the hidden layer  $a^{(2)} \in \mathbb{R}^3$ . One way to do so is

% Theta1 is Theta with superscript "(1)" from lecture
% ie, the matrix of parameters for the mapping from layer 1 (input) to layer 2
% Theta1 has size 3x3
% Assume 'sigmoid' is a built-in function to compute 1 / (1 + exp(-z))

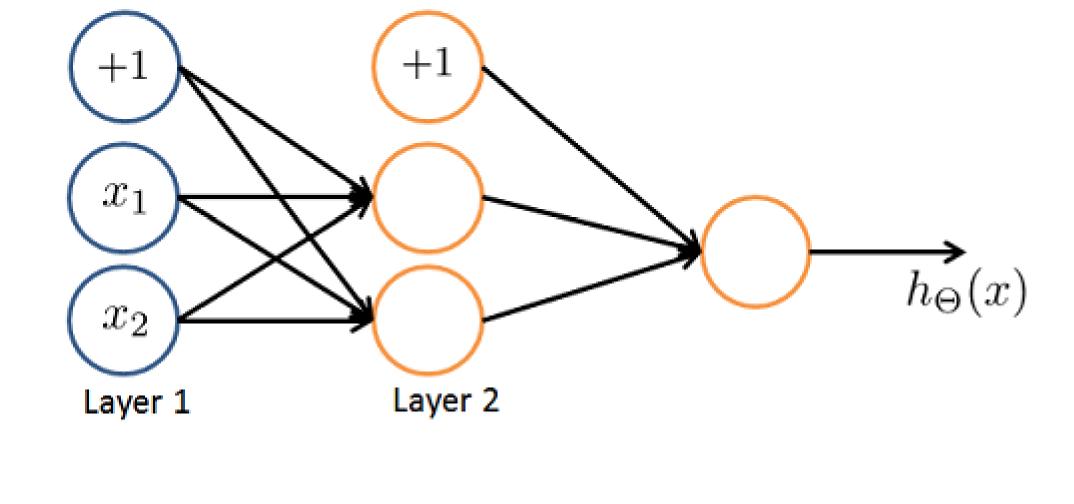
a2 = zeros (3, 1);
for i = 1:3
 for j = 1:3
 a2(i) = a2(i) + x(j) \* Theta1(i, j);
 end
 a2(i) = sigmoid (a2(i));
end

You want to have a vectorized implementation of this (i.e., one that does not use for loops). Which of the following implementations correctly compute  $a^{(2)}$ ? Check all that apply.

- a2 = sigmoid (Theta1 \* x);
- a2 = sigmoid (x \* Theta1);
- a2 = sigmoid (Theta2 \* x);

z = sigmoid(x); a2 = Theta1 \* z;

1 point You are using the neural network pictured below and have learned the parameters  $\Theta^{(1)}=\begin{bmatrix}1&1&2.4\\1&1.7&3.2\end{bmatrix}$  (used to compute  $a^{(2)}$ ) and  $\Theta^{(2)}=\begin{bmatrix}1&0.3&-1.2\end{bmatrix}$  (used to compute  $a^{(3)}$ ) as a function of  $a^{(2)}$ ). Suppose you swap the parameters for the first hidden layer between its two units so  $\Theta^{(1)}=\begin{bmatrix}1&1.7&3.2\\1&1&2.4\end{bmatrix}$  and also swap the output layer so  $\Theta^{(2)}=\begin{bmatrix}1&-1.2&0.3\end{bmatrix}$ . How will this change the value of the output  $h_{\Theta}(x)$ ?



- It will stay the same.
- It will increase.
- It will decrease
- Insufficient information to tell: it may increase or decrease.

I, **Anderson Hitoshi Uyekita**, understand that submitting work that isn't my own may result in permanent failure of this course or deactivation of my Coursera account.

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