

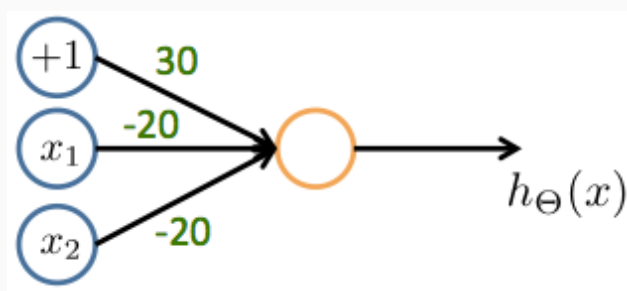
Feedback — VIII. Neural Networks: Representation

[Help](#)

You submitted this quiz on **Sun 13 Apr 2014 2:45 PM IST**. You got a score of **5.00** out of **5.00**.

Question 1

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?



Your Answer	Score	Explanation
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<input type="radio"/> XOR (exclusive OR)		
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<input checked="" type="radio"/> NAND (meaning "NOT AND")	✓ 1.00	This network outputs approximately 1 as long as one of the two inputs is 0.
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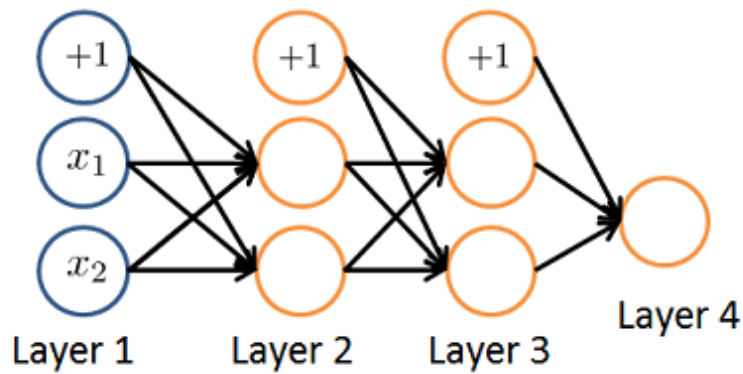
<input type="radio"/> OR		
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<input type="radio"/> AND		
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Total	1.00 / 1.00	
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Question 2

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.



Your Answer

Score

Explanation

☒

$$a_1^{(3)} = g(\Theta_{1,0}^{(2)} a_0^{(2)} + \Theta_{1,1}^{(2)} a_1^{(2)} + \Theta_{1,2}^{(2)} a_2^{(2)})$$

✓

1.00

This correctly uses the first row of $\Theta^{(2)}$ and includes the "+1" term of $a_0^{(2)}$.

☐

$$a_1^{(3)} = g(\Theta_{1,0}^{(1)} a_0^{(1)} + \Theta_{1,1}^{(1)} a_1^{(1)} + \Theta_{1,2}^{(1)} a_2^{(1)})$$

☐

$$a_1^{(3)} = g(\Theta_{1,0}^{(2)} a_0^{(1)} + \Theta_{1,1}^{(2)} a_1^{(1)} + \Theta_{1,2}^{(2)} a_2^{(1)})$$

☐ The activation $a_1^{(3)}$ is not present in this network.

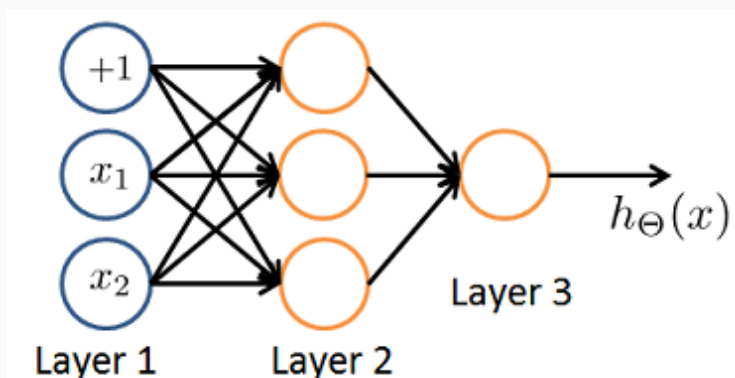
Total

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1.00

Question 3

You have the following neural network:



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

following Octave code:

```
% 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.

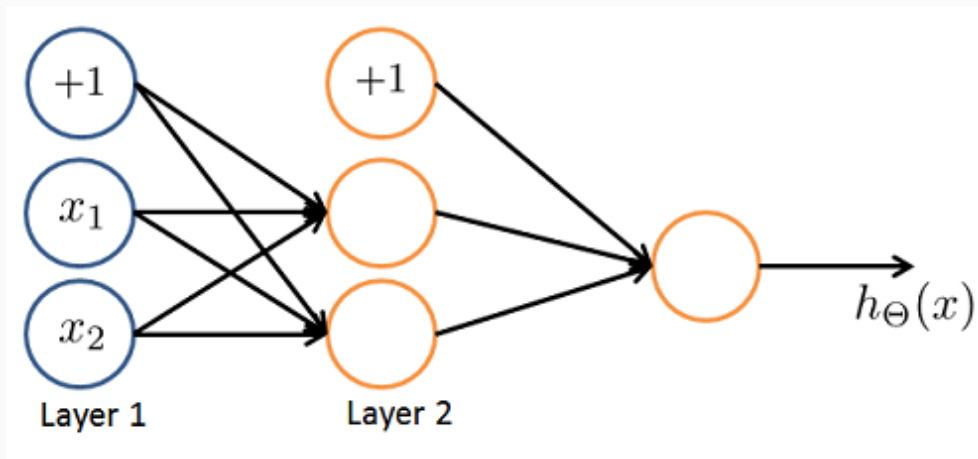
Your Answer	Score	Explanation
<input checked="" type="checkbox"/> a2 = sigmoid (Theta1 * x);	✓ 0.25	In the lecture's notation, $a^{(2)} = g(\Theta^{(1)}x)$, so this version computes it directly, as the sigmoid function will act element-wise.
<input type="checkbox"/> z = sigmoid(x); a2 = sigmoid (Theta1 * z);	✓ 0.25	You do not need to apply the sigmoid function to the inputs.
<input type="checkbox"/> a2 = sigmoid (Theta1 * x);	✓ 0.25	$\Theta^{(2)}$ specifies the parameters from the second to third layers, not first to second.
<input type="checkbox"/> a2 = sigmoid (x * Theta1);	✓ 0.25	The order of the multiplication is important, this will not work as x is a vector of size 3×1 while Theta1 is a matrix of size 3×3 .
Total	1.00 / 1.00	

Question 4

You are using the neural network pictured below and have learned the parameters

$\Theta^{(1)} = \begin{bmatrix} 1 & -1.5 & 3.7 \\ 1 & 5.1 & 2.3 \end{bmatrix}$ (used to compute $a^{(2)}$) and $\Theta^{(2)} = [1 \quad 0.6 \quad -0.8]$ (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 $\Theta^{(1)} = \begin{bmatrix} 1 & 5.1 & 2.3 \\ 1 & -1.5 & 3.7 \end{bmatrix}$ and also swap the output layer so $\Theta^{(2)} = \begin{bmatrix} 1 & -0.8 & 0.6 \end{bmatrix}$. How will this change the value of the output $h_{\Theta}(x)$?



Your Answer	Score	Explanation
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<input checked="" type="radio"/> It will stay the same.	✓ 1.00	Swapping $\Theta^{(1)}$ swaps the hidden layers output $a^{(2)}$. But the swap of $\Theta^{(2)}$ cancels out the change, so the output will remain unchanged.
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<input type="radio"/> It will decrease		
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<input type="radio"/> Insufficient information to tell: it may increase or decrease.		
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<input type="radio"/> It will increase.		
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Total	1.00 / 1.00	
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Question 5

Which of the following statements are true? Check all that apply.

Your Answer	Score	Explanation
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<input type="checkbox"/> A two layer (one input layer, one output layer; no hidden layer) neural network can represent the XOR function.	✓ 0.25	We must compose multiple logical operations by using a hidden layer to represent the XOR function.
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<input checked="" type="checkbox"/> The activation values of	✓ 0.25	
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the hidden units in a neural network, with the sigmoid activation function applied at every layer, are always in the range (0, 1).

The activation function $g(z) = \frac{1}{1+\exp(-z)}$ has a range of (0, 1).

☐ 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 $a_1^{(3)} + a_2^{(3)} + a_3^{(3)} = 1$.

✓ 0.25

The outputs of a neural network are not probabilities, so their sum need not be 1.

☒ Any logical function over binary-valued (0 or 1) inputs x_1 and x_2 can be (approximately) represented using some neural network.

✓ 0.25

Since we can build the basic AND, OR, and NOT functions with a two layer network, we can (approximately) represent any logical function by composing these basic functions over multiple layers.

Total

1.00 /
1.00