

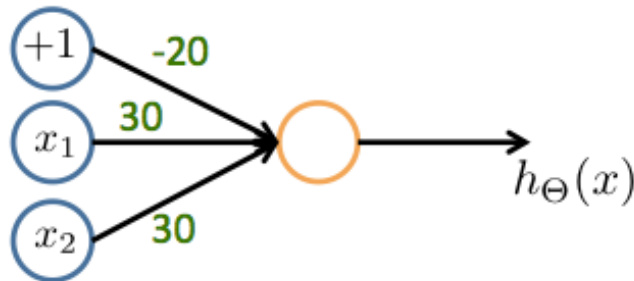
Feedback — VIII. Neural Networks: Representation

[Help](#)

You submitted this quiz on **Mon 11 Nov 2013 3:16 PM PST**. You got a score of **4.00** out of **5.00**. You can [attempt again](#) in 10 minutes.

Question 1

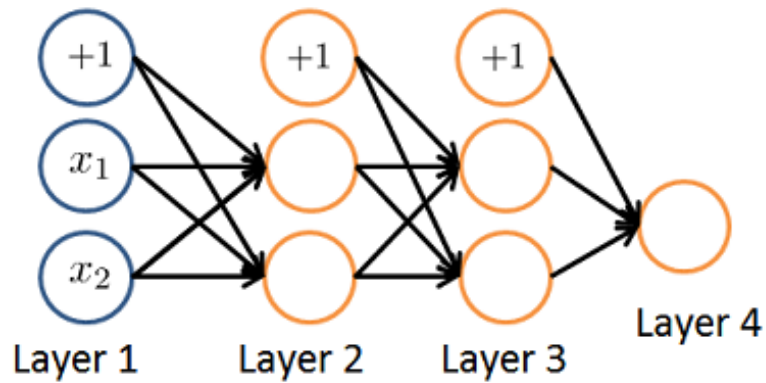
Consider the following neural network which takes two binary-valued inputs $x_1, x_2 \in \{0, 1\}$ and outputs $h_{\Theta}(x)$. Which of the following logical functions does it (approximately) compute?



Your Answer	Score	Explanation
<input checked="" type="radio"/> OR	✓ 1.00	This network will output approximately 1 when either input is 1.
<input type="radio"/> NAND (meaning "NOT AND")		
<input type="radio"/> AND		
<input type="radio"/> XOR (exclusive OR)		
Total	1.00 / 1.00	

Question 2

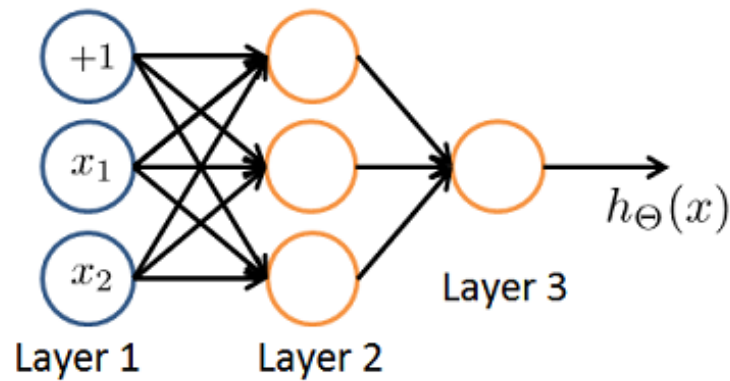
Consider the neural network given below. Which of the following equations correctly computes the activation $a_1^{(3)}$? Note: $g(z)$ is the sigmoid activation function.



Your Answer	Score	Explanation
<input checked="" type="radio"/> $a_1^{(3)} = g(\Theta_{1,0}^{(2)}a_0^{(2)} + \Theta_{1,1}^{(2)}a_1^{(2)} + \Theta_{1,2}^{(2)}a_2^{(2)})$	<input checked="" type="checkbox"/> 1.00	This correctly uses the first row of $\Theta^{(2)}$ and includes the "+1" term of $a_0^{(2)}$.
<input type="radio"/> $a_1^{(3)} = g(\Theta_{1,0}^{(1)}a_0^{(2)} + \Theta_{1,1}^{(1)}a_1^{(2)} + \Theta_{1,2}^{(1)}a_2^{(2)})$		
<input type="radio"/> The activation $a_1^{(3)}$ is not present in this network.		
<input type="radio"/> $a_1^{(3)} = g(\Theta_{1,0}^{(2)}a_0^{(1)} + \Theta_{1,1}^{(2)}a_1^{(1)} + \Theta_{1,2}^{(2)}a_2^{(1)})$		
Total	1.00 / 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"/> <code>a2 = sigmoid (Theta1 * x);</code>	<input checked="" type="checkbox"/> 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"/> <code>z = sigmoid(x); a2 = Theta1 * z;</code>	<input checked="" type="checkbox"/> 0.25	You should apply the sigmoid function after multiplying with $\Theta^{(1)}$, not before.
<input type="checkbox"/> <code>z = sigmoid(x); a2 = sigmoid (Theta1 * z);</code>	<input checked="" type="checkbox"/> 0.25	You do not need to apply the sigmoid function to the inputs.



$a2 = \text{sigmoid}(x * T$
 $\text{heta1});$



0.25

The order of the multiplication is important, this will not work as x is a vector of size 3×1 while Θ_1 is a matrix of size 3×3 .

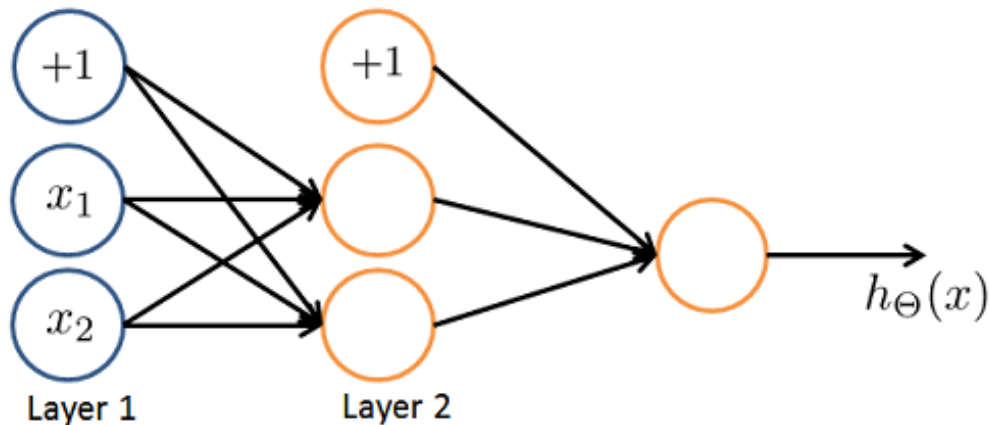
Total

1.00 /
1.00

Question 4

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

$\Theta^{(1)} = \begin{bmatrix} 1 & 2.1 & 1.3 \\ 1 & 0.6 & -1.2 \end{bmatrix}$ (used to compute $a^{(2)}$) and $\Theta^{(2)} = \begin{bmatrix} 1 & 4.5 & 3.1 \end{bmatrix}$ (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 $\Theta^{(1)} = \begin{bmatrix} 1 & 0.6 & -1.2 \\ 1 & 2.1 & 1.3 \end{bmatrix}$ and also swap the output layer so $\Theta^{(2)} = \begin{bmatrix} 1 & 3.1 & 4.5 \end{bmatrix}$. How will this change the value of the output $h_{\Theta}(x)$?



Your Answer

Score

Explanation

☐ It will stay the same.

☐ It will decrease

☐ It will increase.

☒ Insufficient information to tell: it may increase or decrease.



0.00

Swapping $\Theta^{(1)}$ swaps the hidden layers output $a^{(2)}$. But the swap of $\Theta^{(2)}$ cancels out the change, so we can be certain the output will remain unchanged.

Total	0.00 /
	1.00

Question 5

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

Your Answer	Score	Explanation
<input checked="" type="checkbox"/> Any logical function over binary-valued (0 or 1) inputs x_1 and x_2 can be (approximately) represented using some neural network.	<input checked="" type="checkbox"/> 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.
<input type="checkbox"/> 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 $a_1^{(3)} + a_2^{(3)} + a_3^{(3)} = 1$.	<input checked="" type="checkbox"/> 0.25	The outputs of a neural network are not probabilities, so their sum need not be 1.
<input type="checkbox"/> If a neural network is overfitting the data, one solution would be to decrease the regularization parameter λ .	<input checked="" type="checkbox"/> 0.25	A smaller value of λ allows the model to more closely fit the training data, thereby increasing the chances of overfitting.
<input checked="" type="checkbox"/> In a neural network with many layers, we think of each successive layer as being able to use the	<input checked="" type="checkbox"/> 0.25	Each layer computes a non-linear function of its input, so successive layers see more and more complex transformations of the original input.

earlier layers as features,
so as to be able to
compute increasingly
complex functions.

Total	1.00 /
	1.00