

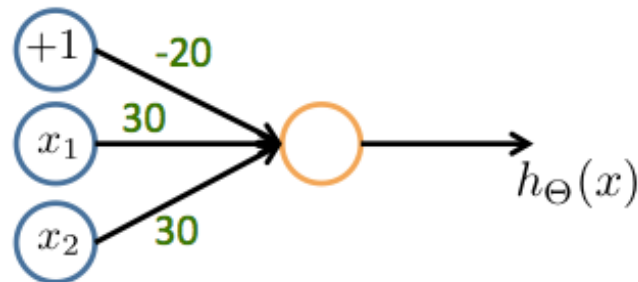
Feedback — VIII. Neural Networks: Representation

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You submitted this quiz on **Sun 13 Apr 2014 10:55 PM PDT**. 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

☐ AND

☐ NAND (meaning "NOT AND")

☒ OR



1.00

This network will output approximately 1 when either input is 1.

☐ XOR (exclusive OR)

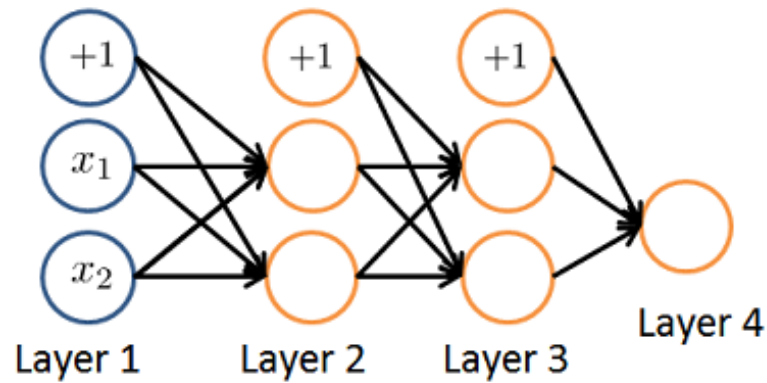
Total

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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**☐

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

☒

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

✓

1.00

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

☐

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

☐

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

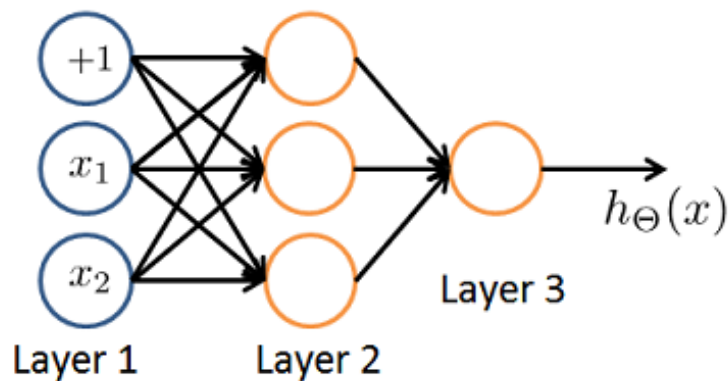
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Question 3

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 type="checkbox"/> $a2 = \text{sigmoid}(\text{Theta2} * x);$	✓ 0.25	$\Theta^{(2)}$ specifies the parameters from the second to third layers, not first to second.
<input checked="" type="checkbox"/> $a2 = \text{sigmoid}(\text{Theta1} * x);$	✓ 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"/> $z = \text{sigmoid}(x);$ $a2 = \text{Theta1} * z;$;	✓ 0.25	You should apply the sigmoid function after multiplying with $\Theta^{(1)}$, not before.

☐ 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 .
 $a_2 = \text{sigmoid}(x * \Theta_1)$;

Total 1.00 / 1.00

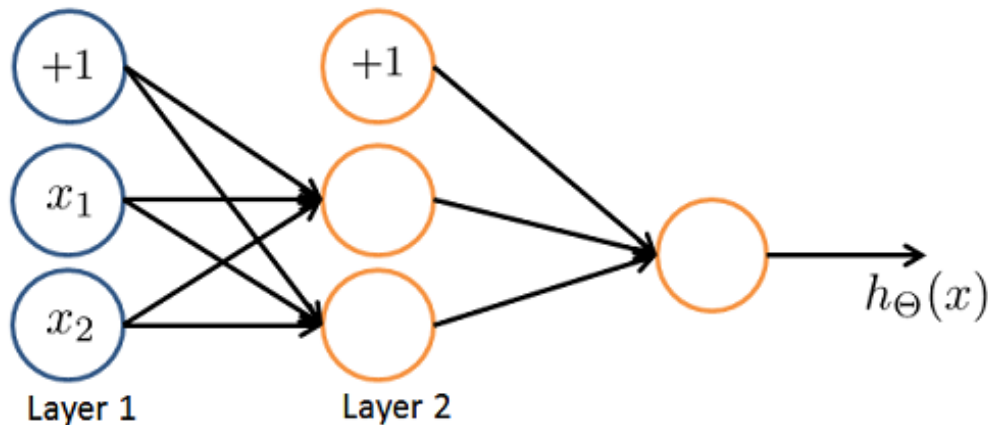
Question 4

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 $\Theta^{(2)} = [1 \quad 4.5 \quad 3.1]$ (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

$\Theta^{(2)} = [1 \quad 3.1 \quad 4.5]$. How will this change the value of the output $h_{\Theta}(x)$?



Your Answer **Score** **Explanation**

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

☒ 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.

☐ It will increase.

☐ It will decrease

Total 1.00 /
1.00

Question 5

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

Your Answer	Score	Explanation
<input checked="" type="checkbox"/> If a neural network is overfitting the data, one solution would be to increase the regularization parameter λ .	<input checked="" type="checkbox"/> 0.25	A larger value of λ will shrink the magnitude of the parameters Θ , thereby reducing the chance of overfitting the data.
<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"/> The activation values of the hidden units in a neural network, with the sigmoid activation function applied at every layer, are always in the range (0, 1).	<input checked="" type="checkbox"/> 0.25	The activation function $g(z) = \frac{1}{1+\exp(-z)}$ has a range of (0, 1).
Total	1.00 / 1.00	

