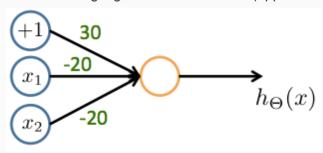
You submitted this quiz on **Sun 13 Apr 2014 2:45 PM IST**. You got a score of **5.00** out of **5.00**.

#### **Question 1**

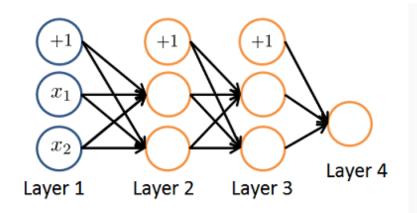
Consider the following neural network which takes two binary-valued inputs  $x_1, x_2 \in \{0, 1\}$  and outputs  $h_{\Theta}(x)$ . Which of the following logical functions does it (approximately) compute?



Your Answer	Score	e Explanation
OXOR (exclusive OR)		
<ul><li>NAND (meaning "NOT AND")</li></ul>	<b>✓</b> 1.00	This network outputs approximately 1 as long as one of the two inputs is 0.
OR		
AND		
Total	1.00 / 1.00	

### **Question 2**

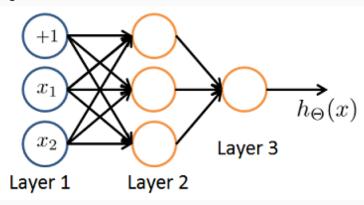
Consider the neural network given below. Which of the following equations correctly computes the activation  $a_1^{(3)}$ ? Note: g(z) is the sigmoid activation function.



Your Answer	Score	Explanation
$a_1^{(3)} = g(\Theta_{1,0}^{(2)}a_0^{(2)} + \Theta_{1,1}^{(2)}a_1^{(2)} + \Theta_{1,2}^{(2)}a_2^{(2)})$	<b>✓</b> 1.00	This correctly uses the first row of $\Theta^{(2)}$ and includes the "+1" term of $a_0^{(2)}$ .
$a_1^{(3)} = g(\Theta_{1,0}^{(1)}a_0^{(1)} + \Theta_{1,1}^{(1)}a_1^{(1)} + \Theta_{1,2}^{(1)}a_2^{(1)})$		
$a_1^{(3)} = g(\Theta_{1,0}^{(2)}a_0^{(1)} + \Theta_{1,1}^{(2)}a_1^{(1)} + \Theta_{1,2}^{(2)}a_2^{(1)})$		
$\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ $		
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.

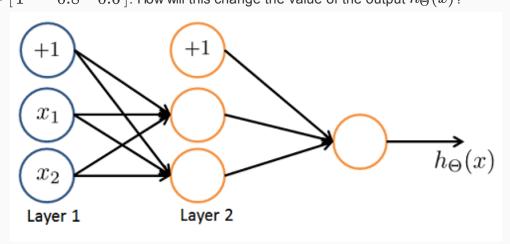
Your Answer	S	Score	Explanation
a2 = sigmoid (Thet a1 * x);	<b>✓</b> 0	).25	In the lecture's notation, $a^{(2)}=g(\Theta^{(1)}x)$ , so this version computes it directly, as the sigmoid function will act element-wise.
z = sigmoid(x); a2 = sigmoid (Theta1 * z);	✔ 0	).25	You do not need to apply the sigmoid function to the inputs.
a2 = sigmoid (Thet a2 * x);	✔ 0	).25	$\boldsymbol{\Theta}^{(2)}$ specifies the parameters from the second to third layers, not first to second.
a2 = sigmoid (x * T heta1);	✔ 0	).25	The order of the multiplication is important, this will not work as x is a vector of size $3\times 1$ while Theta1 is a matrix of size $3x3$ .
Total		.00 /	

#### **Question 4**

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

$$\Theta^{(1)}=egin{bmatrix}1&-1.5&3.7\\1&5.1&2.3\end{bmatrix}$$
 (used to compute  $a^{(2)}$ ) and  $\Theta^{(2)}=[\,1&0.6&-0.8\,]$  (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
● 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 decrease		
<ul> <li>Insufficient information to tell: it may increase or decrease.</li> </ul>		
○ It will increase.		
Total	1.00 / 1.00	

# **Question 5**

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

Your Answer		Score	Explanation
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
✓ The activation values of	~	0.25	

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 $g(z)=rac{1}{1+exp(-z)}$ has a range of (0, 1).
Suppose you have a multi-class classification problem with three classes, trained with a 3 layer network. Let $a_1^{(3)} = (h_\Theta(x))_1 \text{ be the activation of the first output unit, and similarly } a_2^{(3)} = (h_\Theta(x))_2 \text{ and } a_3^{(3)} = (h_\Theta(x))_3 \cdot \text{Then for any input } x, \text{ it must be the case that } a_1^{(3)} + a_2^{(3)} + a_3^{(3)} = 1 \cdot$	✔ 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.	<b>✓</b> 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	