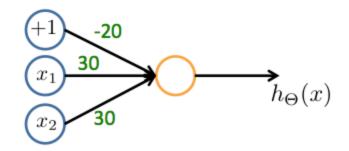
- 1. Which of the following statements are true? Check all that apply.
 - A two layer (one input layer, one output layer; no hidden layer) neural network can represent the XOR function.
 - The activation values of the hidden units in a neural network, with the sigmoid activation function applied at every layer, are always in the range (0, 1).
 - 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 $a_1^{(3)}+a_2^{(3)}+a_3^{(3)}=1$.
 - Any logical function over binary-valued (0 or 1) inputs x_1 and x_2 can be (approximately) represented using some neural network.

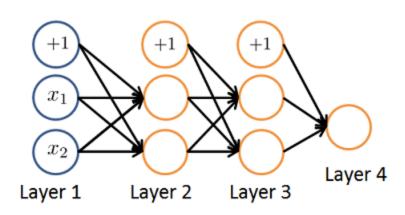
Consider the following neural network which takes two binary-valued inputs $x_1,x_2\in\{0,1\}$ and outputs $h_\Theta(x)$. Which of the following logical functions does it (approximately) compute?





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

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.



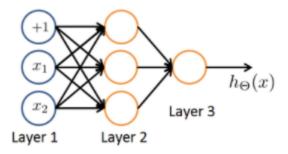
$$a_1^{(3)} = g(\Theta_{1,0}^{(2)}a_0^{(2)} + \Theta_{1,1}^{(2)}a_1^{(2)} + \Theta_{1,2}^{(2)}a_2^{(2)})$$

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

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

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





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:

```
% Thetal is Theta with superscript "(1)" from lecture
% ie, the matrix of parameters for the mapping from layer 1 (input) to layer 2
% Thetal 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) * Thetal(i, j);
   end
   a2(i) = sigmoid (a2(i));
end
```

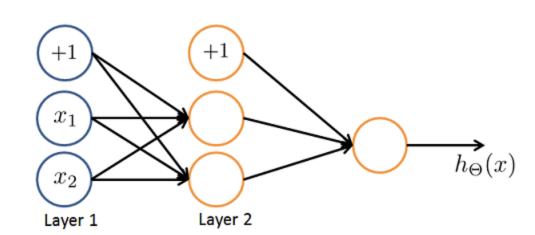
You want to have a vectorized implementation of this (i.e., one that does not use for loops). Which of the following implementations correctly compute $a^{(2)}$? Check all that apply.

```
a2 = sigmoid (Theta1 * x);
```

5. You are using the neural network pictured below and have learned the parameters $\Theta^{(1)} = \begin{bmatrix} 1 & 0.5 & 1.9 \\ 1 & 1.2 & 2.7 \end{bmatrix} \text{ (used to compute } a^{(2)} \text{) and } \Theta^{(2)} = \begin{bmatrix} 1 & -0.2 & -1.7 \end{bmatrix} \text{ (used to compute } a^{(2)} \text{)}$

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 $\Theta^{(1)}=\begin{bmatrix}1&1.2&2.7\\1&0.5&1.9\end{bmatrix}$ and also swap the output

layer so $\Theta^{(2)}=[1 \quad -1.7 \quad -0.2].$ 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.