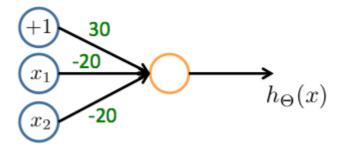
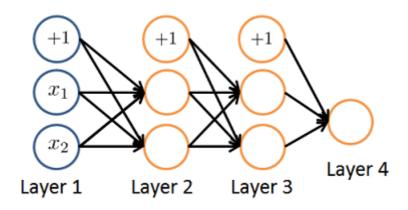
Neural Networks: Representation

- 1. Which of the following statements are true? Check all that apply.
 - 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).
 - \square Any logical function over binary-valued (0 or 1) inputs x_1 and x_2 can be (approximately) represented using some neural network.
 - 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 $a_1^{(3)}+a_2^{(3)}+a_3^{(3)}=1$
 - A two layer (one input layer, one output layer; no hidden layer) neural network can represent the XOR function.
- 2. 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?



NAND (meaning "NOT AND")

- AND
- OR
- 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.



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

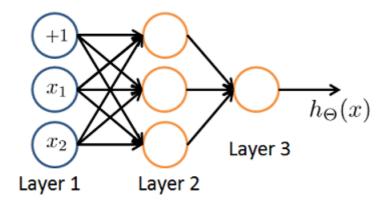
•
$$a_1^{(3)} = g(\Theta_{1,0}^{(1)}a_0^{(1)} + \Theta_{1,1}^{(1)}a_1^{(1)} + \Theta_{1,2}^{(1)}a_2^{(1)})$$

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

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

. .,. . .,. .

- The activation $a_1^{(3)}$ is not present in this network.
- 4. 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))

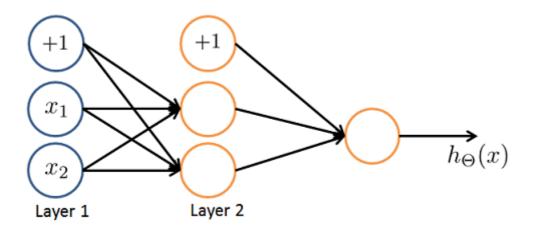
a2 = zeros (3, 1);
for i = 1:3
   for j = 1:3
    a2(i) = a2(i) + x(j) * Theta1(i, j);
end

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.

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

$$\Theta^{(1)} = \begin{bmatrix} 1 & 1 & 2.4 \\ 1 & 1.7 & 3.2 \end{bmatrix} \text{ (used to compute } a^{(2)} \text{) and } \Theta^{(2)} = \begin{bmatrix} 1 & 0.3 & -1.2 \end{bmatrix} \text{ (used to compute } a^{(3)} \text{ as a function of } a^{(2)} \text{). Suppose you swap the parameters for the first hidden layer between its two units so } \Theta^{(1)} = \begin{bmatrix} 1 & 1.7 & 3.2 \\ 1 & 1 & 2.4 \end{bmatrix} \text{ and also swap the output layer so } \Theta^{(2)} = \begin{bmatrix} 1 & -1.2 & 0.3 \end{bmatrix}. \text{ How will this change the value of the output } h_{\Theta}(x)$$
?



It will stay the same.

- It will increase.
- It will decrease
- Insufficient information to tell: it may increase or decrease.