Correct

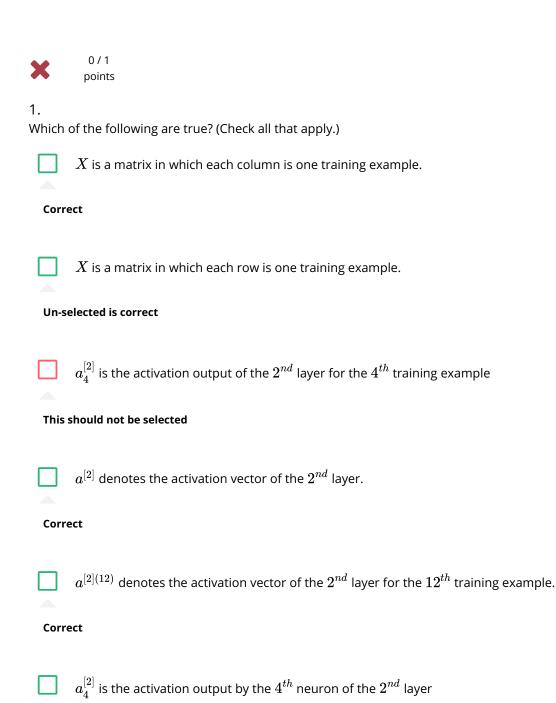
★ Try again once you are ready.

Required to pass: 80% or higher

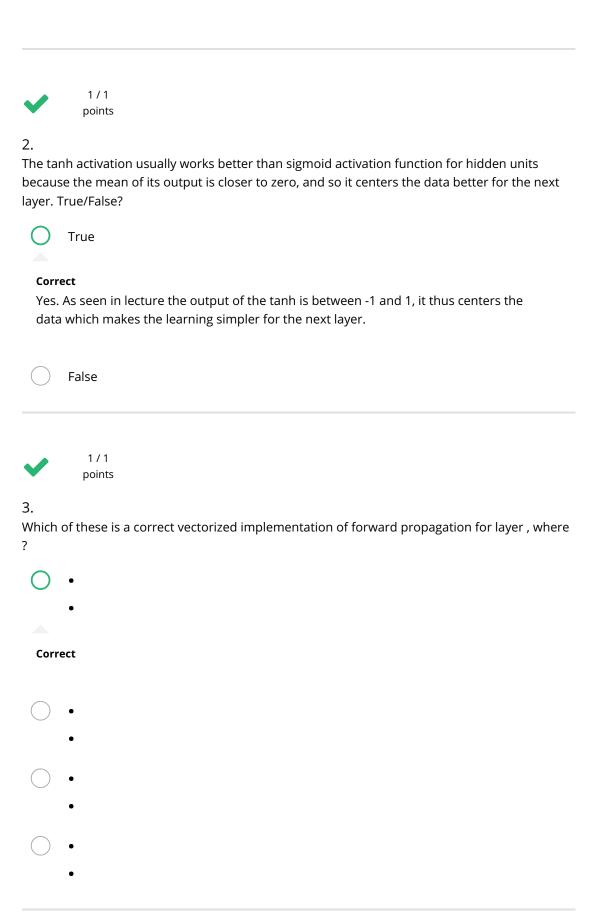
You can retake this quiz up to 3 times every 8 hours.

Back to Week 3

Retake



Un-selected is correct



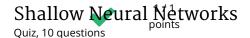
4.

You are building a binary classifier for recognizing cucumbers (y=1) vs. which one of these activation functions would you recommend using for	-
ReLU	
Leaky ReLU	
sigmoid	
Yes. Sigmoid outputs a value between 0 and 1 which makes it a very binary classification. You can classify as 0 if the output is less than 0.5 the output is more than 0.5. It can be done with tanh as well but it is the output is between -1 and 1.	5 and classify as 1 if
tanh	
1/1 points	
5.	
Consider the following code:	
<pre>1 A = np.random.randn(4,3) 2 B = np.sum(A, axis = 1, keepdims = True)</pre>	
What will be B.shape? (If you're not sure, feel free to run this in python	to find out).
(1, 3)	
(4,)	

Correct

(, 3)

Yes, we use (keepdims = True) to make sure that A.shape is (4,1) and not (4,). It makes our code more rigorous.



Suppose you have built a neural network. You decide to initialize the weights and biases to be zero. Which of the following statements is true?

Corre	Each neuron in the first hidden layer will perform the same computation. So even after multiple iterations of gradient descent each neuron in the layer will be computing the same thing as other neurons.
	Each neuron in the first hidden layer will perform the same computation in the first iteration. But after one iteration of gradient descent they will learn to compute different things because we have "broken symmetry".
	Each neuron in the first hidden layer will compute the same thing, but neurons in different layers will compute different things, thus we have accomplished "symmetry breaking" as described in lecture.
	The first hidden layer's neurons will perform different computations from each other even in the first iteration; their parameters will thus keep evolving in their own way.
×	0/1

points

7.

Logistic regression's weights w should be initialized randomly rather than to all zeros, because if you initialize to all zeros, then logistic regression will fail to learn a useful decision boundary because it will fail to "break symmetry", True/False?



True

This should not be selected

No, Logistic Regression doesn't have a hidden layer. If you initialize the weights to zeros, the first example x fed in the logistic regression will output zero but the derivatives of the Logistic Regression depend on the input x (because there's no hidden layer) which is not zero. So at the second iteration, the weights values follow x's distribution and are different from each other if x is not a constant vector.

()	False

8.

You have built a network using the tanh activation for all the hidden units. You initialize the weights to relative large values, using np.random.randn(.,...)*1000. What will happen?

It doesn't matter. So long as you initialize the weights randomly gradient descent is not affected by whether the weights are large or small.

This will cause the inputs of the tanh to also be very large, thus causing gradients to also become large. You therefore have to set \$\$\alpha\$\$ to be very small to prevent divergence; this will slow down learning.

This should not be selected

No. It matters, remember that when the inputs of a tanh is far from zero, its gradient is very close to zero (because of flat edges of tanh).

This will cause the inputs of the tanh to also be very large, thus causing gradients to be close to zero. The optimization algorithm will thus become slow.

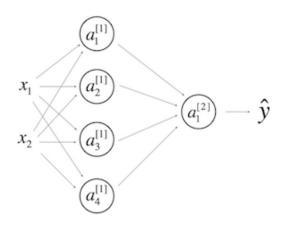
This will cause the inputs of the tanh to also be very large, causing the units to be "highly activated" and thus speed up learning compared to if the weights had to start from small values.



0/1 points

9.

Consider the following 1 hidden layer neural network:



Which of the following statements are True? (Check all that apply).

\$\$W^{[1]}\$\$ will have shape (2, 4)

_						
\mathbf{r}	n	r	r	Δ	•	•

	\$\$W^{[1]}\$\$ will have shape (4, 2)
Corre	ect
	\$\$b^{[1]}\$\$ will have shape (2, 1)
Un-se	elected is correct
	\$\$W^{[2]}\$\$ will have shape (1, 4)
This	should be selected
	\$\$b^{[2]}\$\$ will have shape (4, 1)
Un-se	elected is correct
	\$\$W^{[2]}\$\$ will have shape (4, 1)
This	should not be selected
	\$\$b^{[2]}\$\$ will have shape (1, 1)
Corre	ect
×	0 / 1 points
0. n the s s\$A^{[same network as the previous question, what are the dimensions of \$\$Z^{[1]}\$\$ and 1]}\$\$?
	\$\$Z^{[1]}\$\$ and \$\$A^{[1]}\$\$ are (4,m)
	\$\$Z^{[1]}\$\$ and \$\$A^{[1]}\$\$ are (4,2)

5/10 points (50%)

Remember that $T^{[1]}$ and $A^{[1]}$ are quantities computed over a batch of training examples, not only 1.

\$\$Z^{[1]}\$\$ and \$\$A^{[1]}\$\$ are (1,4)





