

✓ Congratulations! You passed!

TO PASS 80% or higher

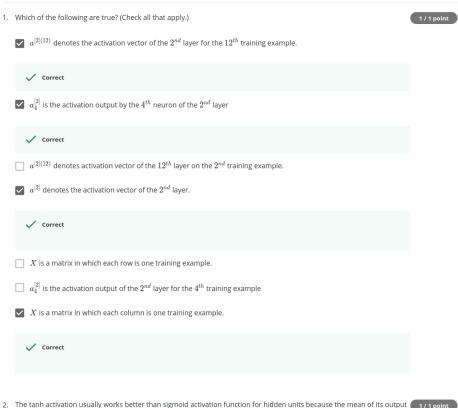


GRADE 100%

Shallow Neural Networks

LATEST SUBMISSION GRADE

100%



2. The tanh activation usually works better than sigmoid activation function for hidden units because the mean of its output is closer to zero, and so it centers the data better for the next layer. True/False?

- True
- False

✓ Correct

Yes. As seen in lecture the output of the tanh is between -1 and 1, it thus centers the data which makes the learning simpler for the next layer.

3. Which of these is a correct vectorized implementation of forward propagation for layer l, where $1 \le l \le L$?

1/1 point

- $\bigcirc \ \, \bullet \ \, Z^{[l]} = W^{[l-1]}A^{[l]} + b^{[l-1]}$
 - $A^{[l]} = g^{[l]}(Z^{[l]})$
- $Z^{[l]} = W^{[l]}A^{[l-1]} + b^{[l]}$
 - $\bullet \ \, A^{[l]} = g^{[l]} \big(Z^{[l]} \big)$
- $\bigcirc \ \, \bullet \ \, Z^{[l]} = W^{[l]} A^{[l]} + b^{[l]}$
 - $A^{[l+1]} = g^{[l+1]}(Z^{[l]})$
- $\bigcirc \ \, \bullet \ \, Z^{[l]} = W^{[l]} A^{[l]} + b^{[l]}$
 - $A^{[l+1]} = g^{[l]}(Z^{[l]})$

✓ Correct

4. You are building a binary classifier for recognizing cucumbers (y=1) vs. watermelons (y=0). Which one of these activation functions would you recommend using for the output layer?

ReLU

	Leaky ReLU	
	sigmoid	
	○ tanh	
	✓ Correct	
	Yes. Sigmoid outputs a value between 0 and 1 which makes it a very good choice for binary classification. You	
	can classify as 0 if the output is less than 0.5 and classify as 1 if the output is more than 0.5. It can be done with tanh as well but it is less convenient as the output is between -1 and 1.	
	ann as wen out to sites convenient as the output's between 1 and 1.	
5.	Consider the following code:	1 / 1 point
	1 A = np.random.randn(4,3)	
	2 B = np.sum(A, axis = 1, keepdims = True)	
	What will be B.shape? (If you're not sure, feel free to run this in python to find out).	
	(4, 1)	
	(4,)	
	(1, 3)	
	(,3)	
	✓ Correct Ver we use (keep dims = True) to make sure that A shape is (A.1) and not (A.). It makes our code more rigorous	
	Yes, we use (keepdims = True) to make sure that A.shape is (4,1) and not (4,). It makes our code more rigorous.	
6.	Suppose you have built a neural network. You decide to initialize the weights and biases to be zero. Which of the following	1 / 1 point
	statements is true?	
	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.	
	F	
	✓ Correct	
	Correct	
7.	Logistic regression's weights w should be initialized randomly rather than to all zeros, because if you initialize to all zeros,	1 / 1 point
	then logistic regression will fail to learn a useful decision boundary because it will fail to "break symmetry", True/False?	
	○ True	
	False	
	✓ Correct	
	Yes, 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.	
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.rando()*1000. What will happen?	1 / 1 point
	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.	
	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 be close to zero. The	
	optimization algorithm will thus become slow.	
	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 $lpha$ to be very small to prevent divergence; this will slow down learning.	
	✓ Correct Voc. tanh horomor flat for large values, this leads its gradient to be close to zero. This closes down the	
	Yes. tanh becomes flat for large values, this leads its gradient to be close to zero. This slows down the optimization algorithm.	

✓ Correct