Key concepts on Deep Neural Networks

Quiz, 10 questions

1 point			
1. What is the "cache" used for in our implementation of forward propagation and backward propagation?			
	It is used to keep track of the hyperparameters that we are searching over, to speed up computation.		
	It is used to cache the intermediate values of the cost function during training.		
	We use it to pass variables computed during forward propagation to the corresponding backward propagation step. It contains useful values for backward propagation to compute derivatives.		
	We use it to pass variables computed during backward propagation to the corresponding forward propagation step. It contains useful values for forward propagation to compute activations.		
1 point2.Among the following, which ones are "hyperparameters"? (Check all that apply.)			
	number of layers L in the neural network		
	number of iterations		
	activation values $a^{[l]}$		
	size of the hidden layers $n^{[l]}$		
	weight matrices $oldsymbol{W}^{[l]}$		
	learning rate $lpha$		
	bias vectors $b^{[l]}$		

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3.

Which of the following statements is true?

The deeper layers of a neural network are typically computing more complex features
of the input than the earlier layers.

The earlier layers of a neural network are typically computing more complex features o
the input than the deeper layers.

1 point

4.

Vectorization allows you to compute forward propagation in an L-layer neural network without an explicit for-loop (or any other explicit iterative loop) over the layers l=1, 2, ...,L. True/False?



True



1 point

5.

Assume we store the values for $n^{[l]}$ in an array called layers, as follows: layer_dims = $[n_x, 4,3,2,1]$. So layer 1 has four hidden units, layer 2 has 3 hidden units and so on. Which of the following forloops will allow you to initialize the parameters for the model?

```
1 for(i in range(1, len(layer_dims)/2)):
2  parameter['W' + str(i)] = np.random.randn(layers[i], layers[i-1])) * 0.01
3  parameter['b' + str(i)] = np.random.randn(layers[i], 1) * 0.01
```

```
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```

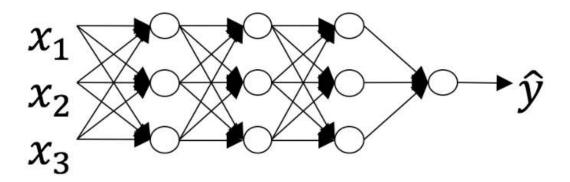
```
1 for(i in range(1, len(layer_dims))):
2  parameter['W' + str(i)] = np.random.randn(layers[i-1], layers[i])) * 0.01
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```

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6.

Consider the following neural network.



How many layers does this network have?

	The number of layers ${\cal L}$ is 4. The number of hidden layers is 3.
	The number of layers $\it L$ is 3. The number of hidden layers is 3.
\bigcirc	The number of layers $\it L$ is 4. The number of hidden layers is 4.
	The number of layers $\it L$ is 5. The number of hidden layers is 4.

1 point

7.

During forward propagation, in the forward function for a layer l you need to know what is the activation function in a layer (Sigmoid, tanh, ReLU, etc.). During backpropagation, the corresponding backward function also needs to know what is the activation function for layer l, since the gradient depends on it. True/False?



True



False

1 point

There are certain functions with the following properties:

(i) To compute the function using a shallow network circuit, you will need a large network (where we measure size by the number of logic gates in the network), but (ii) To compute it using a deep network circuit, you need only an exponentially smaller network. True/False?



True



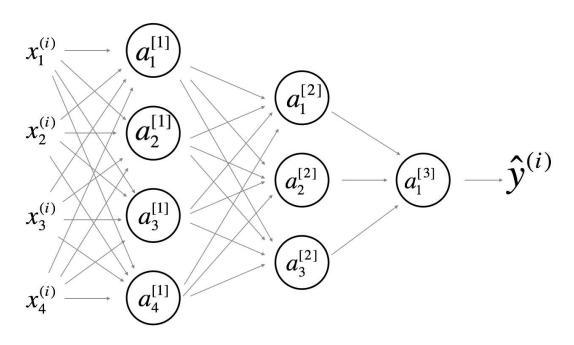
Key concepts on Deep Neural Networks

Quiz, 10 questions

1 point

9.

Consider the following 2 hidden layer neural network:



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

- $W^{[1]}$ will have shape (4, 4)
- $b^{[1]}$ will have shape (4, 1)
- $W^{[1]}$ will have shape (3, 4)
- $b^{[1]}$ will have shape (3, 1)
- $W^{[2]}$ will have shape (3, 4)
- $b^{[2]}$ will have shape (1, 1)
- $W^{[2]}$ will have shape (3, 1)
- $b^{[2]}$ will have shape (3, 1)
- $oxed{W}^{[3]}$ will have shape (3, 1)
- $b^{[3]}$ will have shape (1, 1)
- $W^{[3]}$ will have shape (1, 3)
- $b^{[3]}$ will have shape (3, 1)