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.
- 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.
- 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.
- It is used to cache the intermediate values of the cost function during training.

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1 point

2.
Among the following, which ones are "hyperparameters"? (Check all that apply.)

- [ ] size of the hidden layers $n^{[l]}$
- [ ] learning rate α
- [ ] activation values $a^{[l]}$
- [ ] number of layers L in the neural network
- [ ] weight matrices $W^{[l]}$
- [ ] number of iterations
- [ ] bias vectors $b^{[l]}$

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1 point

3.

Which of the following statements is true?

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- 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 of the input than the deeper layers.

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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
- False

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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 for-loops will allow you to initialize the parameters for the model?

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

```
for(i in range(1, len(layer_dims)/2)):
  parameter['W' + str(i)] = np.random.randn(layers[i], layers[i
      -1])) * 0.01
  parameter['b' + str(i)] = np.random.randn(layers[i-1], 1) * 0
      .01
```

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

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

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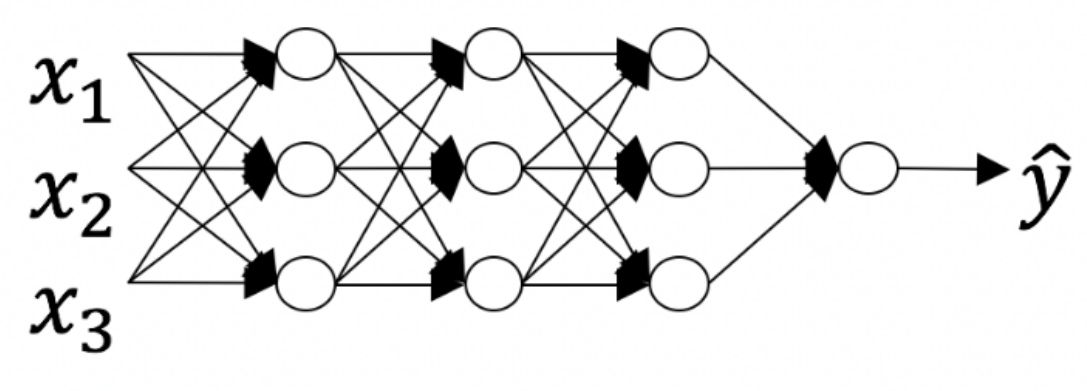
1 point

6.
Consider the following neural network.



How many layers does this network have?

- The number of layers L is 4. The number of hidden layers is 3.
- The number of layers L is 3. The number of hidden layers is 3.
- The number of layers L is 4. The number of hidden layers is 4.
- The number of layers 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

8.
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

False

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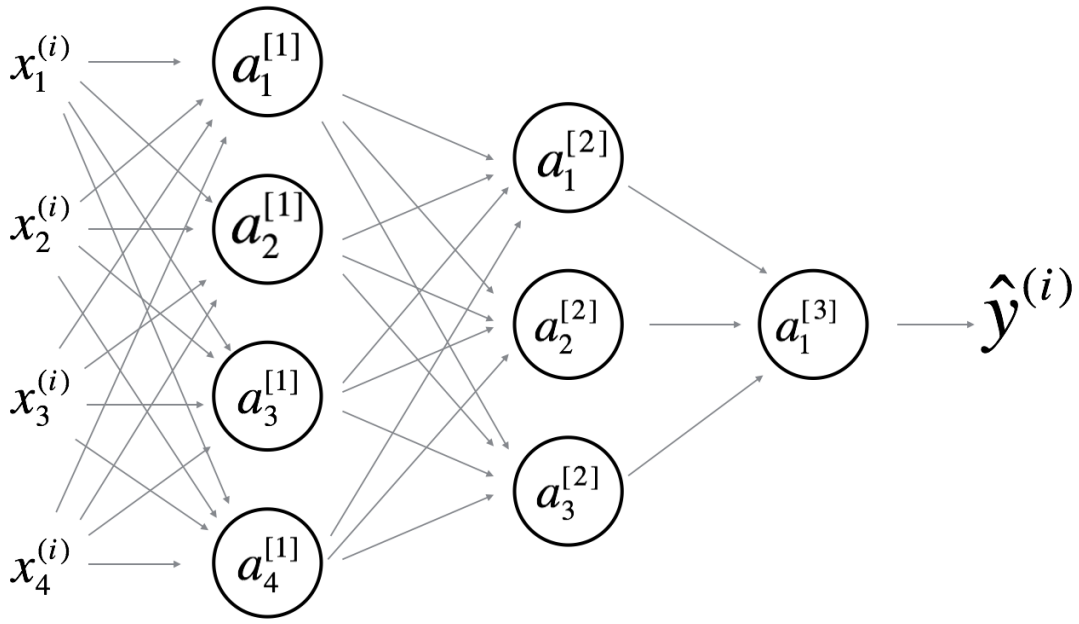
1 point

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