Try again once you are ready

Grade received 70%

Latest Submission Grade 70%

To pass 80% or higher

Try again

1. What is stored in the 'cache' during forward propagation for latter use in backward propagation? **1 / 1 point**

- (•) $Z^{[l]}$
- ( ) $W^{[l]}$
- ( ) $A^{[l]}$
- ( ) $b^{[l]}$

Expand

Correct
Yes. This value is useful in the calculation of $dW^{[l]}$ in the backward propagation.

2. Among the following, which ones are "hyperparameters"? (Check all that apply.) **1 / 1 point**

- [x] size of the hidden layers $n^{[l]}$

Correct

- [x] number of layers L in the neural network

Correct

- [ ] activation values $a^{[l]}$
- [ ] bias vectors $b^{[l]}$
- [x] number of iterations

Correct

- [ ] weight matrices $W^{[l]}$
- [x] learning rate α

Correct

Expand

Correct
Great, you got all the right answers.

3. Which of the following statements is true? **1 / 1 point**

- (•) 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.

Expand

Correct

4. We can not use vectorization to calculate $da^{[l]}$ in backpropagation, we must use a for loop over all the examples. True/False? **0 / 1 point**

- (•) True
- ( ) False

Expand

Incorrect
Incorrect. We can use vectorization in backpropagation to calculate $dA^{[l]}$ for each layer. This computation is done over all the training examples.

5. Assume we store the values for $n^{[l]}$ in an array called layer_dims, 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? **0 / 1 point**

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

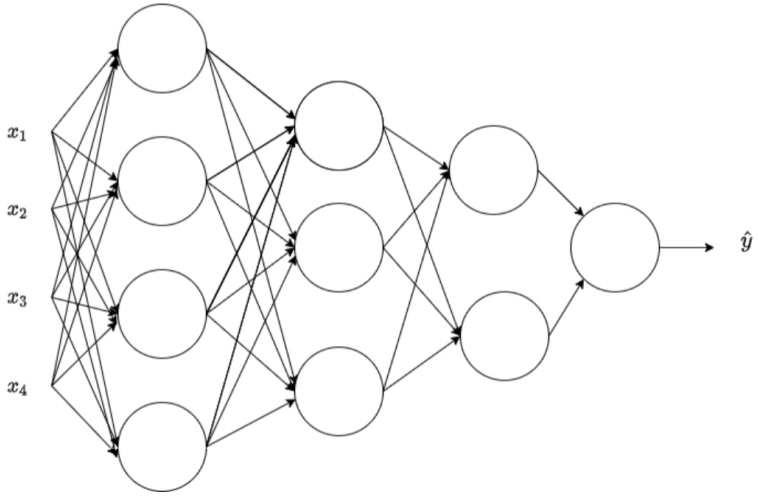
parameter['W' + str(i)] = np.random.randn(layer_dims[i], layer_dims[i-1]) * 0.01
parameter['b' + str(i)] = np.random.randn(layer_dims[i-1], 1) * 0.01

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

Expand

Incorrect

6. Consider the following neural network: **1 / 1 point**



What are all the values of $n^{[0]}, n^{[1]}, n^{[2]}, n^{[3]}$ and $n^{[4]}$?

- 4, 4, 3, 2, 1
- 4, 3, 2, 1
- 4, 3, 2
- 4, 4, 3, 2

Expand

Correct
Yes. The $n^{[l]}$ are the number of units in each layer, notice that $n^{[0]} = n_x$.

7. During forward propagation, in the forward function for a layer l you need to know what is the activation function **1 / 1 point**

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?

- False
- True

Expand

Correct
Yes, as you've seen in week 3 each activation has a different derivative. Thus, during backpropagation you need to know which activation was used in the forward propagation to be able to compute the correct derivative.

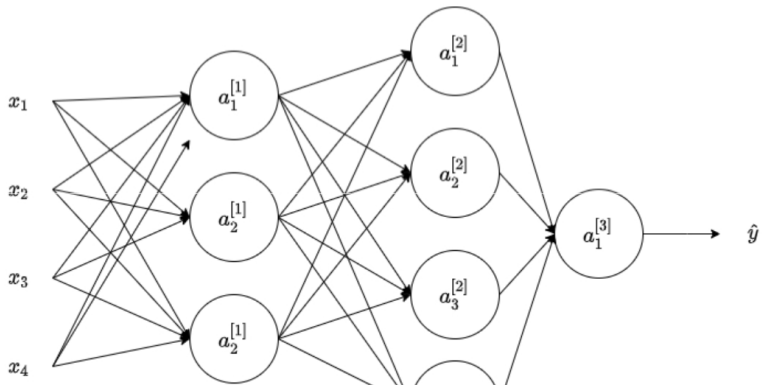
8. A shallow neural network with a single hidden layer and 6 hidden units can compute any function that a neural network with 2 hidden layers and 6 hidden units can compute. True/False? **0 / 1 point**

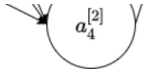
- False
- True

Expand

Incorrect
Incorrect. As seen during the lectures there are functions you can compute with a "small" L-layer deep neural network that shallower networks require exponentially more hidden units to compute.

9. Consider the following 2 hidden layers neural network: **1 / 1 point**





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

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

  ✓ **Correct**
  Yes. More generally, the shape of $W^{[l]}$ is $(n^{[l]}, n^{[l-1]})$.

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

  ✓ **Correct**
  Yes. More generally, the shape of $W^{[l]}$ is $(n^{[l]}, n^{[l-1]})$.

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

  ✓ **Correct**
  Yes. More generally, the shape of $b^{[l]}$ is $(n^{[l]}, 1)$.

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

Expand

✓ **Correct**
Great, you got all the right answers.

10. In the general case if we are training with m examples what is the shape of $A^{[l]}$? **1 / 1 point**

- ( ) $(m, n^{[l]})$
- ( ) $(n^{[l+1]}, m)$
- (•) $(n^{[l]}, m)$
- ( ) $(m, n^{[l+1]})$

Expand

Correct

Yes. The number of rows in $A^{[1]}$ corresponds to the number of units in the l-th layer.