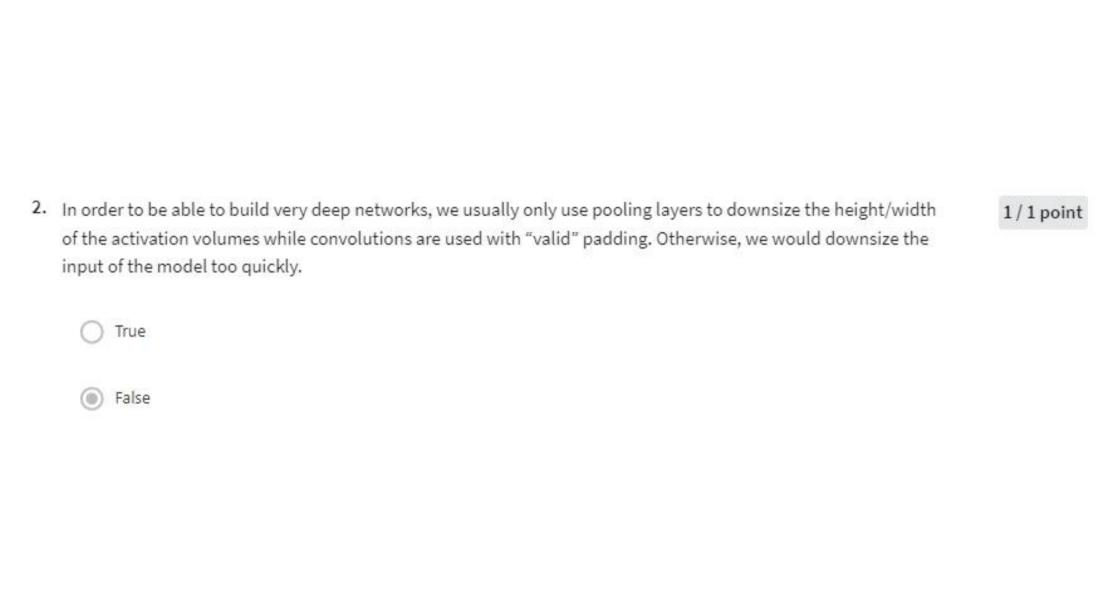
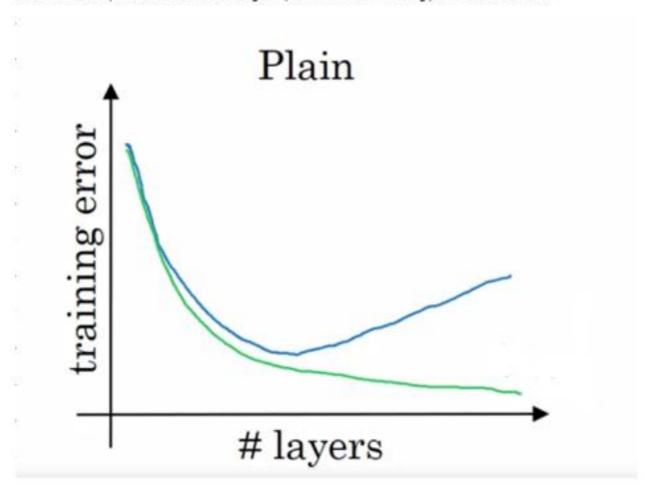
1/1 point



3. Based on the lectures, in the following picture, which curve corresponds to the expected behavior in theory, and which one corresponds to the behavior we get in practice? This when using plain neural networks.



- The blue one depicts the theory, and the green one the reality.
- The green one depicts the results in theory, and also in practice.
- The green one depicts the results in theory, and the blue one the reality.
- The blue one depicts the results in theory, and also in practice.



(Correct

Yes, in theory, we expect that as we increase the number of layers the training error decreases; but in practice after a certain number of layers the error increases.

$$a^{[l+2]} = g \left(W^{[l+2]} g \left(W^{[l+1]} a^{[l]} + b^{[l+1]} \right) + b^{[l+2]} + a^{[l]} \right)$$

Which part corresponds to the skip connection?

- The equation of ResNet.
- The term in the orange box, marked as B.
- The term in the blue box, marked as A.
- The term in the red box, marked as C.



(Correct

Yes, this term is the result of the skip connection or shortcut.

5. In the best scenario when adding a ResNet block it will learn to approximate the identity function after a lot of training, helping improve the overall performance of the network. True/False?

False

() True



(V) Correct

Correct. When adding a ResNet block it can easily learn to approximate the identity function, thus in a worst-case scenario, it will not affect the performance of the network at all.

6. For a volume of $125 \times 125 \times 64$ which of the following can be used to reduce this to a $125 \times 125 \times 32$ volume?

- Use a POOL layer of size 2 × 2 but with a stride of 1.
- Use a 1 x 1 convolutional layer with a stride of 1, and 32 filters.
- Use a 1 × 1 convolutional layer with a stride of 2, and 32 filters.
- Use a POOL layer of size 2 x 2 with a stride of 2.



(Correct

Yes, since using 1×1 convolutions is a great way to reduce the depth dimension without affecting the other dimensions.

By adding these layers we can reduce the computational cost in the inception modules.

✓ Correct

Yes, by using the 1 × 1 convolutional layers we can reduce the depth of the volume and help reduce the computational cost of applying other convolutional layers with different filter sizes.

- Bottleneck layers help to compress the 1x1, 3x3, 5x5 convolutional layers in the inception network.
 - ! This should not be selected

 No, the bottleneck layer doesn't combine any of these different layers.
- The use of bottlenecks doesn't seem to hurt the performance of the network.

✓ Correct

Yes, although it reduces the computational cost significantly.

The bottleneck layer has a more powerful regularization effect than Dropout layers.

Z Expand

(X) Incorrect

You chose the extra incorrect answers.

1/1 point

(2) Correct

Yes, this is a strategy that can provide a good result with small data.

The pointwise convolution convolves the output volume with 1×1 filters. ✓ Correct Yes, the number of filters for the output of the depthwise-separable convolution is determined by the number of 1 x 1 filters used. The depthwise convolution convolves the input volume with 1×1 filters over the depth dimension. Depthwise-separable convolutions are composed of two different types of convolutions. ✓ Correct Yes, it is composed of a depthwise convolution followed by a pointwise convolution. The depthwise convolution convolves each channel in the input volume with a separate filter.



Yes, the output of this kind of convolution is the same as the input.



(Correct

Great, you got all the right answers.

10. Suppose that in a MobileNet v2 Bottleneck block we have an $n \times n \times 5$ input volume, we use 30 filters for the expansion, in the depthwise convolutions we use 3×3 filters, and 20 filters for the projection. How many parameters are used in the complete block, suppose we don't use bias?

(80

() 8250

() 1101

() 1020



(Correct

Yes, the expansion filters use $5 \times 30 = 150$ parameters, the depthwise convolutions need $3 \times 3 \times 30 = 270$ parameters, and the projection part $30 \times 20 = 600$ parameters.