Congratulations! You passed!

Grade received 80%

1.

2.

Correct!

Latest Submission Grade 80% To pass 80% or higher

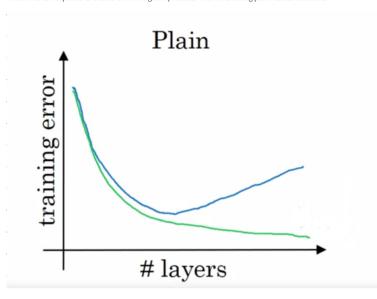
Retake the assignment in **7h 25m**

Go to next item

When building a ConvNet, typically you start with some POOL layers followed by some CONV layers. True/False?	1/1 point
○ True	
False	
∠ [™] Expand	
○ Correct Correct. It is typical for ConvNets to use a POOL layer after some Conv layers; sometimes even one POOL layer after each CONV layer; but is not common to start with POOL layers.	
In order to be able to build very deep networks, we usually only use pooling layers to downsize the height/width of the activation volumes while convolutions are used with "valid" padding. Otherwise, we would downsize the input of the model too quickly.	1/1 point
○ True	
False	
∠ [™] Expand	
○ Correct	

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.

1/1 point



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 blue one depicts the results in theory, and also in practice.	
∠ ⁷ Expand	
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.	
The following equation captures the computation in a ResNet block. What goes into the two blanks above?	1/1 po
$a^{[l+2]} = g(W^{[l+2]}g(W^{[l+1]}a^{[l]} + b^{[l+1]}) + b^{l+2} + \dots) + \dots$	
$lacksquare$ $a^{[l]}$ and 0, respectively	
0 and $z^{[l+1]}$, respectively	
0 and $a^{[l]}$, respectively $z^{[l]} \text{ and } a^{[l]}, \text{ respectively}$	
∠ ⁿ Expand	
Contect	
Which ones of the following statements on Residual Networks are true? (Check all that apply.)	1 / 1 pc
Using a skip-connection helps the gradient to backpropagate and thus helps you to train deeper networks	
✓ Correct	
This is true.	
The skip-connection makes it easy for the network to learn an identity mapping between the input and the output within the ResNet block.	
✓ Correct	
This is true.	
The skip-connections compute a complex non-linear function of the input to pass to a deeper layer in the network.	
$\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ $	
√ ² Expand	
1 imes 1 convolutions are the same as multiplying by a single number. True/False?	1/1
1 imes 1 convolutions are the same as multiplying by a single number. True/False?	1/1 po
$1 imes 1$ convolutions are the same as multiplying by a single number. True/False? $\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \$	1/1 pc
	1/1 po
○ True	1/1 pc

7.	Which of the following are true about bottleneck layers? (Check all that apply)	0 / 1 point
	☐ The use of bottlenecks doesn't seem to hurt the performance of the network.	
	By adding these layers we can reduce the computational cost in the inception modules.	
	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.	
	The bottleneck layer has a more powerful regularization effect than Dropout layers.	
	! This should not be selected No, bottleneck layers are not intended as a regularization strategy.	
	Bottleneck layers help to compress the 1x1, 3x3, 5x5 convolutional layers in the inception network.	
	∠ ^ス Expand	
8.	Models trained for one computer vision task can't be used directly in another task. In most cases, we must change the softmax layer, or the last layers of the model and re-train for the new task. True/False?	1/1 point
	○ False	
	True	
	∠ [™] Expand	
	Correct Yes, this is a good way to take advantage of open-source models trained more or less for the task you want to do. This may also help you save a great number of computational resources and data.	
9.	In Depthwise Separable Convolution you:	0/1 point
	For the "Depthwise" computations each filter convolves with all of the color channels of the input image.	, ,
	The final output is of the dimension $n_{out} \times n_{out} \times n'_{c}$ (where n'_{c} is the number of filters used in the pointwise convolution step).	
	Perform one step of convolution.	
	For the "Depthwise" computations each filter convolves with only one corresponding color channel of the input image.	
	✓ Correct	
	Perform two steps of convolution.	
	✓ Correct	
	The final output is of the dimension $n_{out} \times n_{out} \times n_c$ (where n_c is the number of color channels of the input image).	
	You convolve the input image with n_c number of $n_f \times n_f$ filters (n_c is the number of color channels of the input image).	
	You convolve the input image with a filter of $n_f \times n_f \times n_c$ where n_c acts as the depth of the filter (n_c is the number of color channels of the input image).	

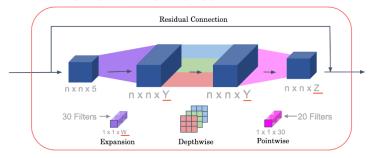
∠ Expand

Ves, a 1×1 layer doesn't act as a single number because it makes a sum over the depth of the volume.

 $\textbf{10.} \ \ \text{Fill in the missing dimensions shown in the image below (marked W, Y, Z).}$

1/1 point

MobileNet v2 Bottleneck



- W = 30, Y = 20, Z = 20
- W = 5, Y = 30, Z = 20
- W = 5, Y = 20, Z = 5
- W = 30, Y = 30, Z = 5



⊘ Correct