**Due** Jan 17, 2:59 AM EST

## Congratulations! You passed!

**Grade received** 100% **To pass** 80% or higher

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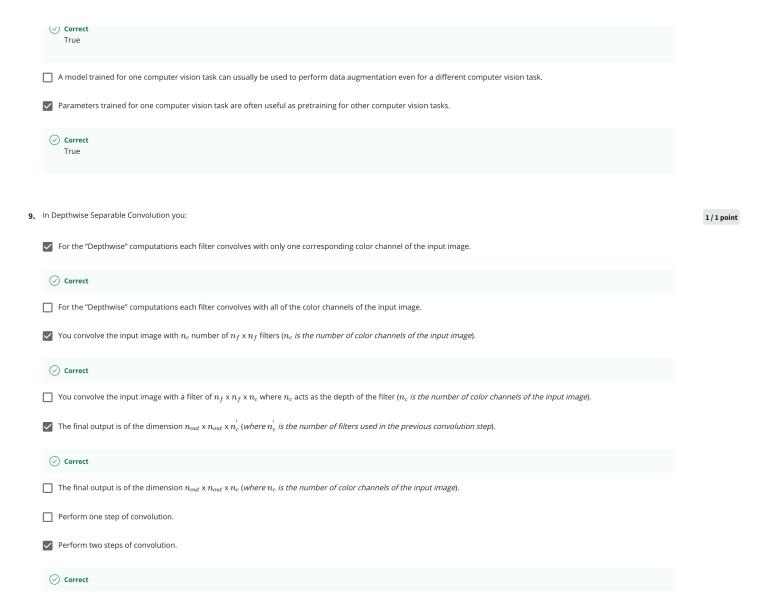
## Deep Convolutional Models

Latest Submission Grade 100%

1.	Which of the following do you typically see in a ConvNet? (Check all that apply.)	1/1 point
	Multiple POOL layers followed by a CONV layer	
	FC layers in the first few layers	
	Multiple CONV layers followed by a POOL layer	
	Correct True, as seen in the case studies.	
	FC layers in the last few layers	
	<ul> <li>Correct</li> <li>True, fully-connected layers are often used after flattening a volume to output a set of classes in classification.</li> </ul>	
2.	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
	<ul><li>True</li><li>False</li></ul>	
3.	Training a deeper network (for example, adding additional layers to the network) allows the network to fit more complex functions and thus almost always results in lower training error. For this question, assume we're referring to "plain" networks.	1/1 point
	<ul><li>False</li><li>True</li></ul>	
	<ul> <li>Correct</li> <li>Correct, Resnets are here to help us train very deep neural networks.</li> </ul>	
4.	The following equation captures the computation in a ResNet block. What goes into the two blanks above? $a^{[l+2]} = g(W^{[l+2]}g(W^{[l+1]}a^{[l]} + b^{[l+1]}) + b^{l+2} + \underline{\hspace{1cm}}) + \underline{\hspace{1cm}}$	1/1 point
	$igcirc$ 0 and $a^{[l]}$ , respectively	
	$left{igorall} a^{[l]}$ and 0, respectively	
	$igcirc$ $0$ and $z^{[l+1]}$ , respectively	
	$igcup z^{[l]}$ and $a^{[l]}$ , respectively	

5.	Which ones of the following statements on Residual Networks are true? (Check all that apply.)	1 / 1 point
	Using a skip-connection helps the gradient to backpropagate and thus helps you to train deeper networks	
	$oxed{igsquare}$ A ResNet with L layers would have on the order of $L^2$ skip connections in total.	
	The skip-connections compute a complex non-linear function of the input to pass to a deeper layer in the network.	
	The skip-connection makes it easy for the network to learn an identity mapping between the input and the output within the ResNet block.	
5.	Suppose you have an input volume of dimension $n_H \times n_W \times n_C$ . Which of the following statements you agree with? (Assume that "1x1 convolutional layer" below always uses a stride of 1 and no padding.)	1 / 1 point
	$\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ $	
	$igwedge$ You can use a 2D pooling layer to reduce $n_H, n_W$ , but not $n_C$ .	
	$\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ $	
	$igwedge$ You can use a 1x1 convolutional layer to reduce $n_C$ but not $n_H$ , $n_W$ .	
	igodots Correct Yes, a 1x1 convolutional layer with a small number of filters is going to reduce $n_C$ but will keep the dimensions $n_H$ and $n_W$	
7.	Which ones of the following statements on Inception Networks are true? (Check all that apply.)	1/1 point
	Making an inception network deeper (by stacking more inception blocks together) <i>might</i> not hurt training set performance.	
	Inception blocks usually use 1x1 convolutions to reduce the input data volume's size before applying 3x3 and 5x5 convolutions.	
	Inception networks incorporate a variety of network architectures (similar to dropout, which randomly chooses a network architecture on each step) and thus has a similar regularizing effect as dropout.	
	A single inception block allows the network to use a combination of 1x1, 3x3, 5x5 convolutions and pooling.	
3.	Which of the following are common reasons for using open-source implementations of ConvNets (both the model and/or weights)? Check all that apply.	1/1 point
	The same techniques for winning computer vision competitions, such as using multiple crops at test time, are widely used in practical deployments (or production system deployments) of ConvNets.	
	It is a convenient way to get working with an implementation of a complex ConvNet architecture.	

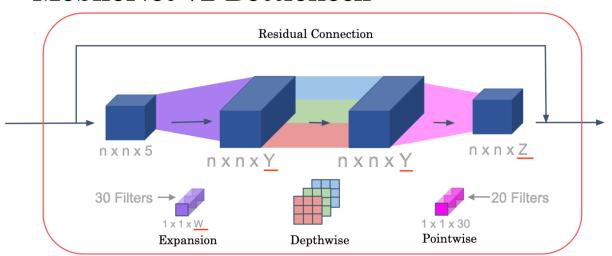
Correct



10. 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 = 20, ∠ = 5

● W = 5, Y = 30, Z = 20

W = 30, Y = 30, Z = 5

**⊘** Correct