## Deep convolutional models

TOTAL POINTS 10

1.	Which of the following do you typically see as you move to deeper layers in a ConvNet?	1 point
	$igcirc$ $n_H$ and $n_W$ increases, while $n_C$ also increases	
	$igcup n_H$ and $n_W$ decreases, while $n_C$ also decreases	
	$igcup n_H$ and $n_W$ increases, while $n_C$ decreases	
	$lacksquare$ $n_H$ and $n_W$ decrease, while $n_C$ increases	
2.	Which of the following do you typically see in a ConvNet? (Check all that apply.)	1 point
	Multiple CONV layers followed by a POOL layer	
	Multiple POOL layers followed by a CONV layer	
	FC layers in the last few layers	
	FC layers in the first few layers	
3.	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 point
	○ True	
	<ul><li>False</li></ul>	
4.	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 point
	○ True	
	False	
5.	The following equation captures the computation in a ResNet block. What goes into the two blanks above?	1 point
	$a^{[l+2]} = g(W^{[l+2]}g(W^{[l+1]}a^{[l]} + b^{[l+1]}) + b^{l+2} + \underline{\hspace{1cm}}) + \underline{\hspace{1cm}}$	
	$lacktriangle$ $a^{[l]}$ and 0, respectively	
	$\bigcirc$ 0 and $oldsymbol{a}^{[l]}$ , respectively	
	$igcirc$ $z^{[l]}$ and $a^{[l]}$ , respectively	
	$igcirc$ $0$ and $z^{[l+1]}$ , respectively	

6.	Which ones of the following statements on Residual Networks are true? (Check all that apply.)	1 point
	$\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ $	
	✓ Using a skip-connection helps the gradient to backpropagate and thus helps you to train deeper networks	
	The skip-connection makes it easy for the network to learn an identity mapping between the input and the output within the ResNet block.	
	The skip-connections compute a complex non-linear function of the input to pass to a deeper layer in the network.	
7.	Suppose you have an input volume of dimension 64x64x16. How many parameters would a single 1x1 convolutional filter have (including the bias)?	1 point
	O 2	
	17	
	O 1	
	O 4097	
8.	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 point
	$\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ $	
	$\checkmark$ You can use a 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 $n_H$ , $n_W$ .	

9.	Wh	ich ones of the following statements on Inception Networks are true? (Check all that apply.)	1 point
		Inception networks incorporates 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.	
	<b>~</b>	Inception blocks usually use $1x1$ convolutions to reduce the input data volume's size before applying $3x3$ and $5x5$ convolutions.	
		Making an inception network deeper (by stacking more inception blocks together) should not hurt training set performance.	
	<b>V</b>	A single inception block allows the network to use a combination of $1x1$ , $3x3$ , $5x5$ convolutions and pooling.	
10.		ich of the following are common reasons for using open-source implementations of ConvNets th the model and/or weights)? Check all that apply.	1 point
	<b>~</b>	Parameters trained for one computer vision task are often useful as pretraining for other computer vision tasks.	
		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.	