Deep convolutional models

1.	Which of the following do you typically see as you move to deeper layers in a ConvNet?	1 point	
	$igcup n_H$ and n_W increases, while n_C decreases		
	$\bigcirc \ \ n_H$ 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 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		
	False		Activate Wi
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	Go to Sattings t
	True		
	False		
5.	The following equation captures the computation in a ResNet block. What goes into the two blanks above?	1 point	
	$a^{[l+2]} = g(W^{[l+2]}g(W^{[l+1]}a^{[l]} + b^{[l+1]}) + b^{l+2} + \underline{\hspace{1cm}}) + \underline{\hspace{1cm}}) + \underline{\hspace{1cm}}$		
	$\bigcirc \ a^{[l]}$ and 0, respectively		
	$\bigcirc \ 0$ and $z^{[l+1]}$, respectively		
	\bigcirc 0 and $a^{[l]}$, respectively		
	$\bigcup z^{[l]}$ and $a^{[l]}$, respectively		
6.	Which ones of the following statements on Residual Networks are true? (Check all that apply.)	1 point	
	☐ The skip-connection makes it easy for the network to learn an identity mapping between the input and the output within the ResNet block.		
	Using a skip-connection helps the gradient to backpropagate and thus helps you to train deeper networks		
	The skip-connections compute a complex non-linear function of the input to pass to a deeper layer in the network.		Activate Win
	$\ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ \ $		Go to Settings to

	have (including the bias)?	
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	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.)	
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	9. Which ones of the following statements on Inception Networks are true? (Check all that apply.)	
	Inception blocks usually use 1x1 convolutions to reduce the input data volume's size before applying 3x3 and 5x5 convolutions.	
	Activ	at
	9. Which ones of the following statements on Inception Networks are true? (Check all that apply.) 1 point	
	☐ Inception blocks usually use 1x1 convolutions to reduce the input data volume's size before applying 3x3 and 5x5	
	convolutions. 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.	
	A single inception block allows the network to use a combination of 1x1, 3x3, 5x5 convolutions and pooling.	
	Making an inception network deeper (by stacking more inception blocks together) should not hurt training set performance.	
	10. Which of the following are common reasons for using open-source implementations of ConvNets (both the model and/or weights)? Check all that apply.	
	lt is a convenient way to get working an implementation of a complex ConvNet architecture.	
	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.	
	A model trained for one computer vision task can usually be used to perform data augmentation even for a different computer vision task. Activate Upgrade to submit Gay to Settin	
1. 2.	$_{H}$ and n_{W} decrease, while n_{C} increases.	
	a. Multiple CONV layers followed by a POOL layer	
	b. FC layers in the last few years	
3.	ulse_	
4.	<u> </u>	
5.	$rac{l}{l}$ and 0	
5. 6.	and o	
υ.	a. Using a skip-connection helps the gradient to backpropagate and thus helps you to trai	<u>n</u>

deeper networks.

b. The skip-connections makes it easy for the network to learn an identity mapping between the input and the output within the ResNet block.

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8.

- a. You can use a 1*1 convolutional layer to reduce n_C , but not n_H , n_W .
- b. You can use a pooling layer to reduce n_H , n_W , but not n_C .

9.

- a. A single inception block allows the network to use a combination of 1*1, 3*3, 5*5 convolutions and pooling.
- b. inception blocks usually use 1*1 convolutions to reduce the input data volume's size before applying 3*3 and 5*5 convolutions.

10.

- a. Parameters trained for one computer vision task are often useful as pretraining for other computer vision tasks.
- b. <u>It is a convenient way to get working an implementation of a complex ConvNet architecture.</u>