← Deep convolutional models Quiz, 10 questions

1 point	
1. Which of the	he following do you typically see as you move to deeper layers in a ConvNet?
\bigcap n_H	$_{H}$ and n_{W} decreases, while n_{C} also decreases
\bigcap n_H	$_{H}$ and n_{W} decrease, while n_{C} increases
\bigcap n_H	$_{H}$ and n_{W} increases, while n_{C} decreases
\bigcap n_H	$_{H}$ and n_{W} increases, while n_{C} also increases
1 point 2.	
	he following do you typically see in a ConvNet? (Check all that apply.)
Mu	ultiple CONV layers followed by a POOL layer
Mu	ultiple POOL layers followed by a CONV layer
FC	layers in the last few layers
FC	layers in the first few layers
and thus a	deeper network (for example, adding additional layers to the network) allows the network to fit more complex functions almost always results in lower training error. For this question, assume we're referring to "plain" networks. ue

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5.

goes into the two blanks above?

The fol	llowing equation captures the computation in a ResNet block. What
$a^{[l+2]}$:	$= g(W^{[l+2]}g(W^{[l+1]}a^{[l]} + b^{[l+1]}) + b^{l+2} + \underline{\hspace{1cm}}) + \underline{\hspace{1cm}}$
	$z^{[l]}$ and $a^{[l]}$, respectively
	0 and $z^{\left[l+1 ight]}$, respectively
	$oldsymbol{a}^{[l]}$ and 0, respectively
	0 and $a^{[l]}$, respectively

point

Which ones of the following statements on Residual Networks are true? (Check all that apply.)

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.
A ResNet with L layers would have on the order of L^2 skip connections in total.

Using a skip-connection helps the gradient to backpropagate and thus helps you to train deeper networks

1 point

Suppose you have an input volume of dimension 64x64x16. How many parameters would a single 1x1 convolutional filter have (including the bias)?

4097

point

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.)

		You can use a pooling layer to reduce n_H , n_W , but not n_C .
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You can use a 1x1 convolutional layer to reduce n_H , n_W , and n_C .

You can use a 1x1 convolutional layer to reduce n_C but not n_H , n_W .

1 oint	
ch o	nes of the following statements on Inception Networks are true? (Check all that apply.)
	nception blocks usually use 1x1 convolutions to reduce the input data volume's size before applying 3x3 and 5x5
	convolutions.
,	A single inception block allows the network to use a combination of 1x1, 3x3, 5x5 convolutions and pooling.
	nception 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.
	Making an inception network deeper (by stacking more inception blocks together) should not hurt training set performance.
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oint	f the following are common reasons for using open-source implementations of ConvNets (both the model and/or weights)?
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ch ock al	l that apply.
ch o ck al	t is a convenient way to get working an implementation of a complex ConvNet architecture. A model trained for one computer vision task can usually be used to perform data augmentation even for a different
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ch o ock al	t is a convenient way to get working an implementation of a complex ConvNet architecture. A model trained for one computer vision task can usually be used to perform data augmentation even for a different computer vision task. 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
ch o ock al	It is a convenient way to get working an implementation of a complex ConvNet architecture. A model trained for one computer vision task can usually be used to perform data augmentation even for a different computer vision task. 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.