축하합니다! 통과하셨습니다!

받은 성적 100% 통과 점수: 80% 이상

다음 항목으로 이동

Deep Convolutional Models

최근 제출물 성적 100%

1. Which of the following do you typically see in a ConvNet? (Check all that apply.) 1 / 1점

- [x] Multiple CONV layers followed by a POOL layer

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  True, as seen in the case studies.

- [x] FC layers in the last few layers

  맞습니다
  True, fully-connected layers are often used after flattening a volume to output a set of classes in classification.

- [ ] Multiple POOL layers followed by a CONV layer
- [ ] FC layers in the first few layers

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 / 1점

- ( ) True
- (•) False

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  Correct!

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 / 1점

- (•) False
- ( ) True

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  Correct, Resnets are here to help us train very deep neural networks.

4. The following equation captures the computation in a ResNet block. What goes into the two blanks above? 1 / 1점

$$a^{[l+2]} = g(W^{[l+2]}g(W^{[l+1]}a^{[l]} + b^{[l+1]}) + b^{l+2} + \_\_\_\_\_) + \_\_\_\_\_$$

- ( ) 0 and $z^{[l+1]}$, respectively
- (•) $a^{[l]}$ and 0, respectively
- ( ) 0 and $a^{[l]}$, respectively
- ( ) $z^{[l]}$ and $a^{[l]}$, respectively

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  Correct

5. Which ones of the following statements on Residual Networks are true? (Check all that apply.) 1 / 1점

- [ ] The skip-connections compute a complex non-linear function of the input to pass to a deeper layer in the network.
- [x] The skip-connection 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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  This is true.

- [ ] A ResNet with L layers would have on the order of L^2 skip connections in total.
- [x] Using a skip-connection helps the gradient to backpropagate and thus helps you to train deeper networks

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  This is true.

6. Suppose you have an input volume of dimension n_H x n_W x 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 / 1점

- [x] You can use a 2D pooling layer to reduce n_H, n_W, but not n_C.

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This is correct.

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

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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

- [ ] You can use a 2D pooling layer to reduce n_H, n_W, and n_C.
- [ ] You can use a 1x1 convolutional layer to reduce n_H, n_W, and n_C.

7. Which ones of the following statements on Inception Networks are true? (Check all that apply.) **1 / 1점**

- [x] Inception blocks usually use 1x1 convolutions to reduce the input data volume's size before applying 3x3 and 5x5 convolutions.

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- [ ] 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.
- [ ] Making an inception network deeper (by stacking more inception blocks together) *might* not hurt training set performance.
- [x] A single inception block allows the network to use a combination of 1x1, 3x3, 5x5 convolutions and pooling.

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8. 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 / 1점**

- [x] Parameters trained for one computer vision task are often useful as pretraining for other computer vision tasks.

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True

- [ ] A model trained for one computer vision task can usually be used to perform data augmentation even for a different computer vision task.
- [x] It is a convenient way to get working with an implementation of a complex ConvNet architecture.

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True

- [ ] 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.

9. In Depthwise Separable Convolution you: **1 / 1점**

- [ ] For the "Depthwise" computations each filter convolves with all of the color channels of the input image.
- [x] The final output is of the dimension n_{out} x n_{out} x $n_c^{'}$ (*where $n_c^{'}$ is the number of filters used in the previous convolution step*).

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- [x] You convolve the input image with n_c number of n_f x n_f filters (*n_c is the number of color channels of the input image*).

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- [x] Perform two steps of convolution.

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- [ ] You convolve the input image with a filter of n_f x n_f x n_c where n_c acts as the depth of the filter (*n_c is the number of color channels of the input image*).
- [ ] The final output is of the dimension n_{out} x n_{out} x n_c (*where n_c is the number of color channels of the input image*).
- [ ] Perform one step of convolution.
- [x] For the "Depthwise" computations each filter convolves with only one corresponding color channel of the input image.

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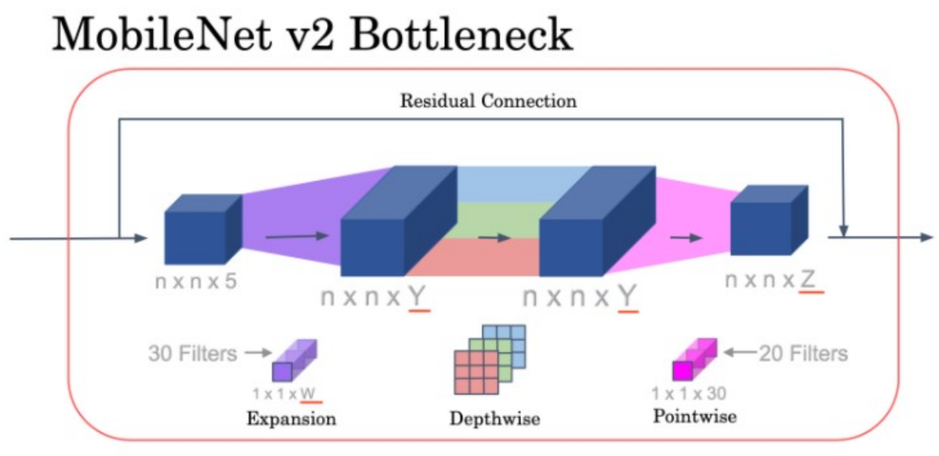
10. Fill in the missing dimensions shown in the image below (*marked W, Y, Z*). **1 / 1점**



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

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