

FIT 3181/5215 Deep Learning

Quiz for: Advanced Convolutional Neural Networks

Teaching team

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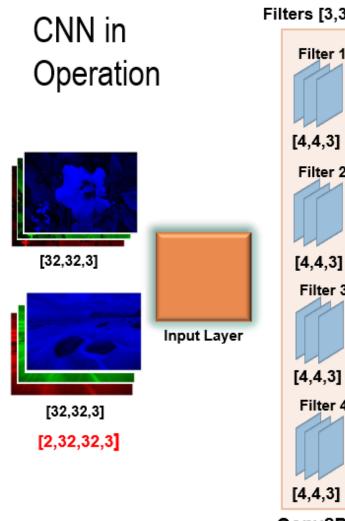
Which statements are correct? (MC)

- □ A. In traditional approach, the training signal from classifier can be used to improve feature extractor.
- □ B. In deep learning approach, the training signal from classifier can be used to improve feature extractor.
- C. In traditional approach, the training signal from classifier cannot be used to improve feature extractor.
- D. In deep learning approach, the training signal from classifier cannot be used to improve feature extractor.

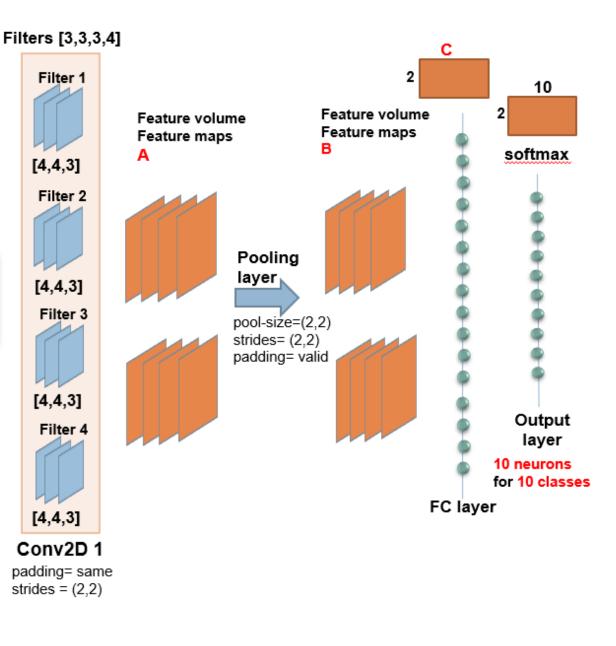
Which statements are correct? (MC)

- A. In traditional approach, the training signal from classifier can be used to improve feature extractor.
- □ B. In deep learning approach, the training signal from classifier can be used to improve feature extractor. [x]
- C. In traditional approach, the training signal from classifier cannot be used to improve feature extractor. [x]
- D. In deep learning approach, the training signal from classifier cannot be used to improve feature extractor.

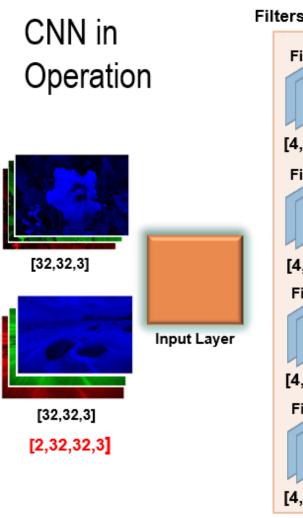
What are the shapes of tensors in A, B and the value of the width in C?

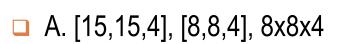


- □ A. [15,15,4], [8,8,4], 8x8x4
- □ B. [2,15,15,4], [2,8,8,4], 2x8x8x4
- □ C. [2,16,16,4], [2,8,8,4], 2x8x8x4
- □ D. [2,16,16,4], [2,8,8,4], 8x8x4

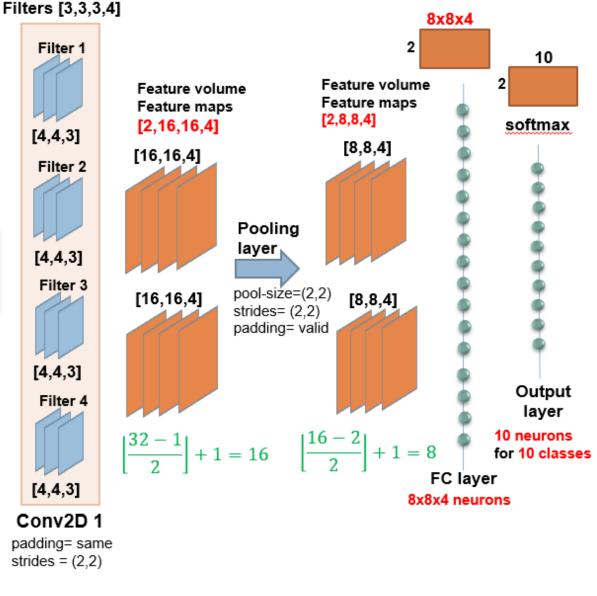


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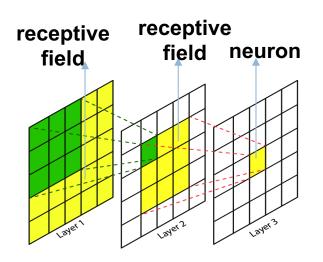
- □ B. [2,15,15,4], [2,8,8,4], 2x8x8x4
- □ C. [2,16,16,4], [2,8,8,4], 2x8x8x4
- **D**. [2,16,16,4], [2,8,8,4], 8x8x4 **[x]**

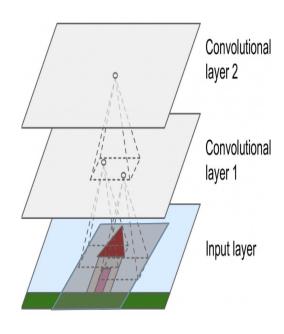


What are correct statements about the receptive field? (MC)

- □ A. Receptive field of neurons on higher layers become smaller.
- □ B. The value of a neuron is not computationally relevant to its receptive field.
- C. Receptive field of neurons on higher layers become larger.
- □ D. The value of a neuron is computationally relevant to its receptive field.

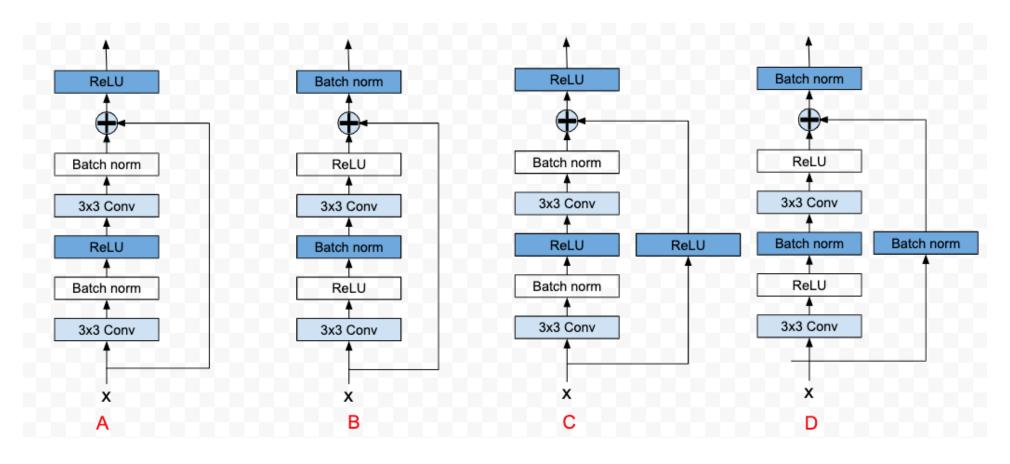
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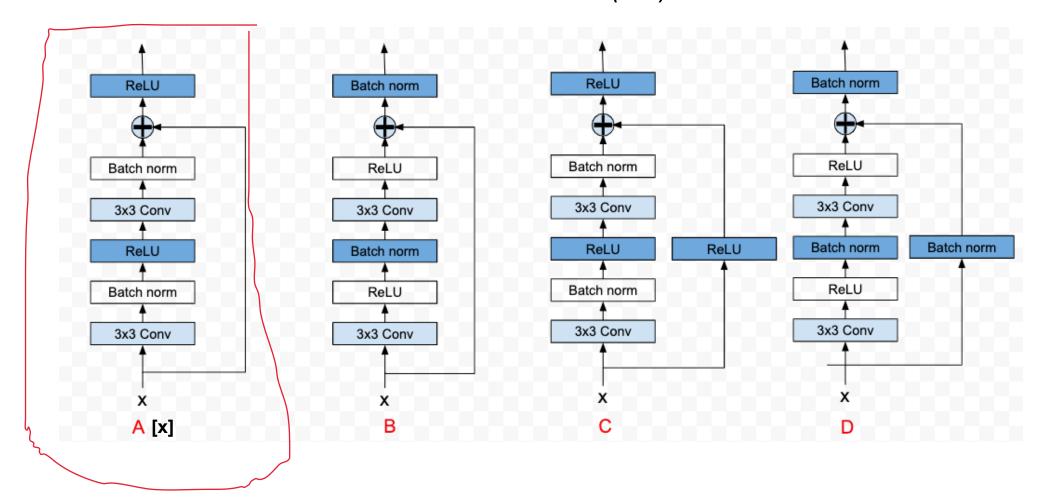


- □ A. Receptive field of neurons on higher layers become smaller.
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- C. Receptive field of neurons on higher layers become larger. [x]
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Which illustration is correct for the residual block? (SC).



Which illustration is correct for the residual block? (SC).



- A. [10,32,32,3]
- B. [10,16,16,3]
- c. [3,32,32,3]
- D. Raise an error.

```
class Residual(tf.keras.Model):
    def init (self, num channels, use 1x1conv=False, strides=1):
        super().__init__()
        self.conv1 = tf.keras.layers.Conv2D(
            num channels, padding='same', kernel size=3, strides=strides)
        self.conv2 = tf.keras.layers.Conv2D(num_channels, kernel_size=3, padding='same')
        self.conv3 = None
        if use 1x1conv:
            self.conv3 = tf.keras.layers.Conv2D(
                num_channels, kernel_size=1, strides=strides)
        self.bn1 = tf.keras.layers.BatchNormalization()
        self.bn2 = tf.keras.layers.BatchNormalization()
    def call(self, X):
        Y = tf.keras.activations.relu(self.bn1(self.conv1(X)))
       Y = self_bn2(self_conv2(Y))
       if self.conv3 is not None:
            X = self.conv3(X)
       Y += X
        return tf.keras.activations.relu(Y)
blk = Residual(num_channels=3)
X = tf.random.uniform((10, 32, 32, 3))
Y = blk(X)
print(Y.shape)
```

- A. [10,32,32,3] **[x]**
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Which statements are correct for ResNet architecture? (MC).

- □ A. In ResNet architecture, ReLU activation function is followed by Batch Normalization layer.
- B. It is possible to replace ReLU by Sigmoid activation function because of the skip-connection can help to reduce gradient vanishing.
- C. 1x1 Conv in skip-connection is used to change number of output channels.
- D. A ResNet model consists of many ResNet blocks, each ResNet block consists of many residual blocks, each residual block includes several convolutional and activation layers.

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- D. A ResNet model consists of many ResNet blocks, each ResNet block consists of many residual blocks, each residual block includes several convolutional and activation layers. [x]

Given an adversarial example x_{adv} of a clean example x w.r.t. model $f, y \in \{1,2,...,M\}$ is the true label. Which statements are correct? (MC).

- □ A. x_{adv} and x look very similar under human perspective
- □ B. x_{adv} and x look very different under human perspective
- \square C. $argmax_{1 \leq m \leq M} f_m(x_{adv}) = y$
- \square D. $argmax_{1 \leq m \leq M} f_m(x_{adv}) \neq y$

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- \square D. $argmax_{1 \leq m \leq M} f_m(x_{adv}) \neq y$ [x]

Given a constraint of an adversarial example as follow: $x_{adv} \in B_{\epsilon}(x) = \{x': ||x'-x||_{\infty} \le \epsilon\}$. Which statements are correct? (MC)

- \square A. This constraint to make sure that x_{adv} and x look very similar under human perspective
- \square B. This constraint to make sure that x_{adv} and x look very different under human perspective
- \square C. This constraint to make sure that $argmax_{1 \le m \le M} f_m(x_{adv}) = argmax_{1 \le m \le M} f_m(x)$
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that means epsilon is the pixel difference threshold

Given a DL model $f(x; \theta)$ parameterized by θ where $f(x; \theta)$ represents the prediction probabilities of x associated with a ground-truth label $y \in \{1, ..., M\}$, we find an adversarial example by $x_{adv} = argmax_{x' \in B_{\epsilon}(x)} l(f(x'; \theta), y)$. Which statements are correct? (MC)

- \square A. We maximally increase the chance to predict x_{adv} with label y.
- \square B. We maximally decrease the chance to predict x_{adv} with label y.
- \square C. We maximally increase the chance to predict x_{adv} with any else label $y' \neq y$.
- D. It is a targeted attack.
- E. It is an untargeted attack.

Given a DL model $f(x;\theta)$ parameterized by θ where $f(x;\theta)$ represents the prediction probabilities of x associated with a ground-truth label $y \in \{1, ..., M\}$, we find an adversarial example by $x_{adv} = argmax_{x' \in B_{\epsilon}(x)} l(f(x';\theta), y)$. Which statements are correct? (MC)

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