



**MONASH**  
University

FIT 3181/5215 Deep Learning

**Quiz for:**  
**Advanced Convolutional Neural Networks**

**Teaching team**

Department of Data Science and AI  
Faculty of Information Technology, Monash University  
Email: [trunglm@monash.edu](mailto:trunglm@monash.edu)



# Question 1

*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.

# Question 1

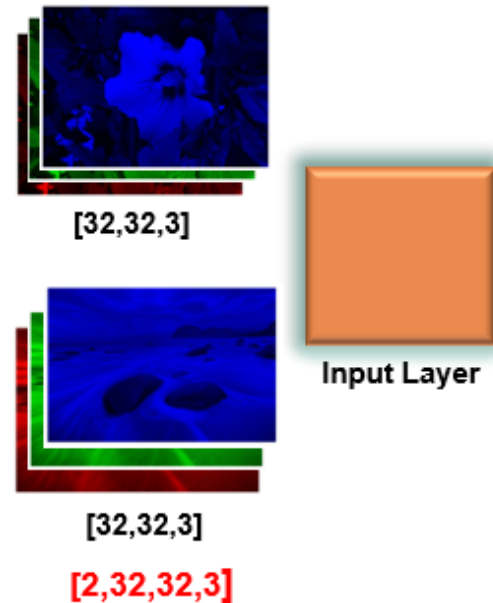
*Which statements are correct? (MC)*

- ☐ 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.

# Question 2

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

CNN in Operation



Filters  $[3, 3, 3, 4]$



Conv2D 1  
padding= same  
strides = (2,2)

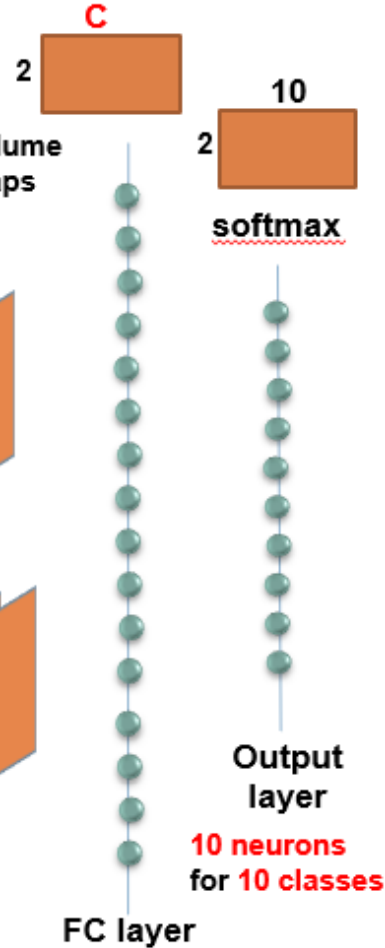
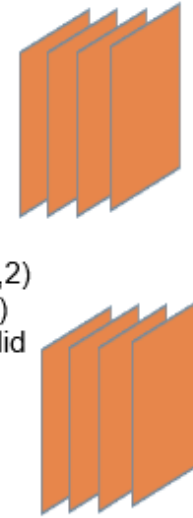
Feature volume  
Feature maps  
**A**



Pooling layer

pool-size=(2,2)  
strides= (2,2)  
padding= valid

Feature volume  
Feature maps  
**B**

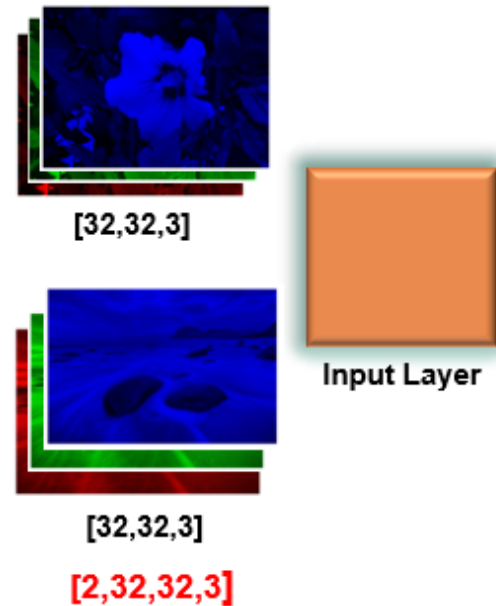


- ☐ A.  $[15, 15, 4]$ ,  $[8, 8, 4]$ ,  $8 \times 8 \times 4$
- ☐ B.  $[2, 15, 15, 4]$ ,  $[2, 8, 8, 4]$ ,  $2 \times 8 \times 8 \times 4$
- ☐ C.  $[2, 16, 16, 4]$ ,  $[2, 8, 8, 4]$ ,  $2 \times 8 \times 8 \times 4$
- ☐ D.  $[2, 16, 16, 4]$ ,  $[2, 8, 8, 4]$ ,  $8 \times 8 \times 4$

# Question 2

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

## CNN in Operation



Filters [3,3,3,4]

Conv2D 1  
padding= same  
strides = (2,2)

Feature volume  
Feature maps  
[2,16,16,4]

[16,16,4]

[16,16,4]

$$\left\lfloor \frac{32 - 1}{2} \right\rfloor + 1 = 16$$

Pooling layer

pool-size=(2,2)  
strides= (2,2)  
padding= valid

Feature volume  
Feature maps  
[2,8,8,4]

[8,8,4]

[8,8,4]

$$\left\lfloor \frac{16 - 2}{2} \right\rfloor + 1 = 8$$

FC layer  
8x8x4 neurons

8x8x4  
2  
10  
softmax

Output layer

10 neurons  
for 10 classes

- ☐ 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 [x]

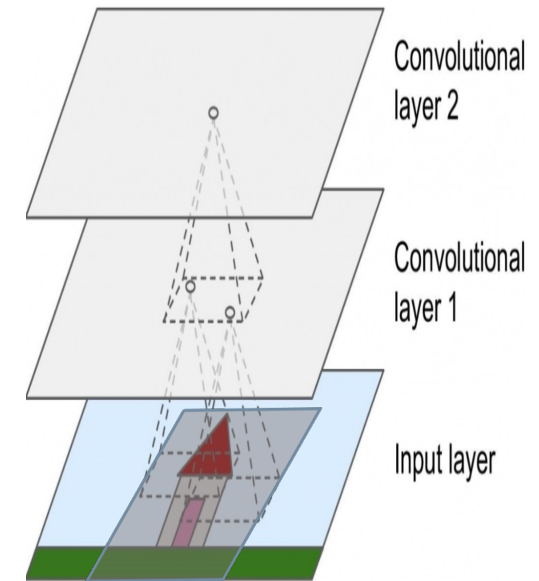
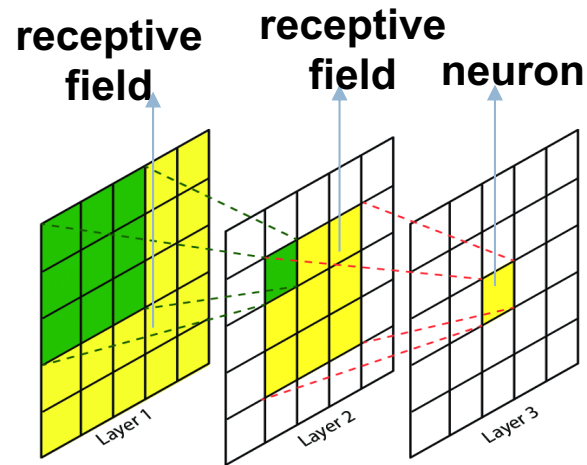
# Question 3

*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.

# Question 3

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

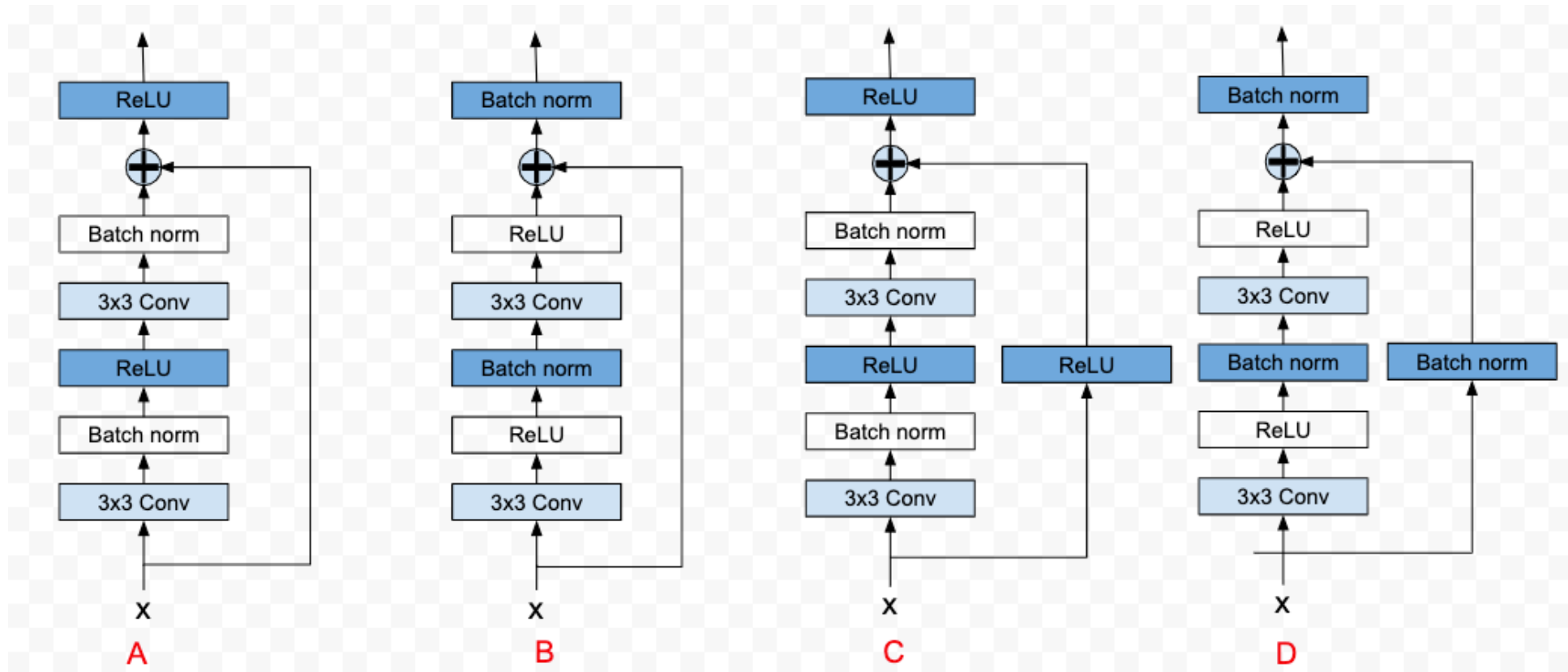


- ☐ A. Receptive field of neurons on higher layers become smaller.
- ☐ B. The value of a neuron is not computationally relevant to its receptive field.
- ☐ C. Receptive field of neurons on higher layers become larger. **[x]**
- ☐ D. The value of a neuron is computationally relevant to its receptive field.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



# Question 4

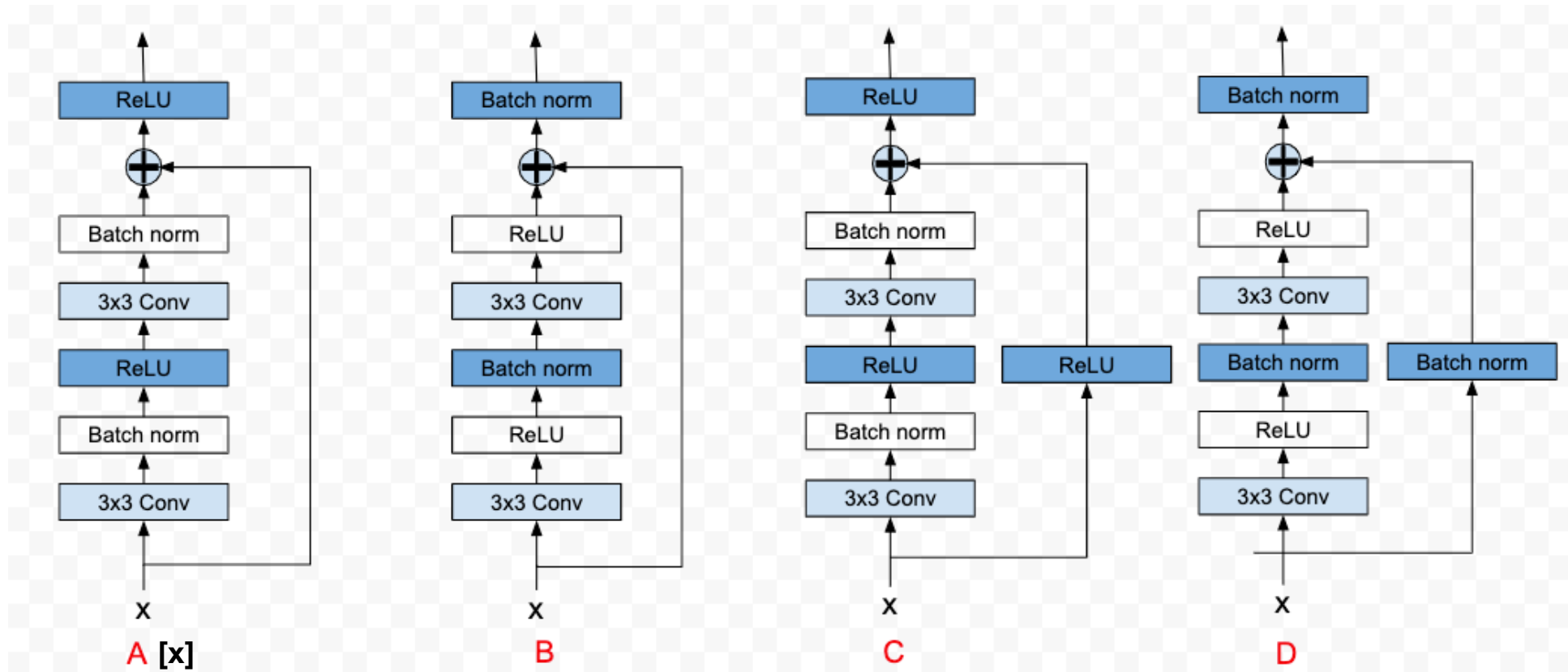
*Which illustration is correct for the residual block? (SC).*





# Question 4

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



# Question 5

*Given an implementation of the residual block as below? What is the shape of Y (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)
```

# Question 5

*Given an implementation of the residual block as below? What is the shape of Y (SC).*

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

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# Question 6

*Given an implementation of the residual block as below? What is the shape of Y (SC).*

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

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class Residual(tf.keras.Model):
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blk = Residual(num_channels=6)
X = tf.random.uniform((10, 32, 32, 3))
Y = blk(X)
print(Y.shape)
```

# Question 6

*Given an implementation of the residual block as below? What is the shape of Y (SC).*

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

```
class Residual(tf.keras.Model):
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        self.conv2 = tf.keras.layers.Conv2D(num_channels, kernel_size=3, padding='same')
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        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=6)
X = tf.random.uniform((10, 32, 32, 3))
Y = blk(X)
print(Y.shape)
```

# Question 7

*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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- ☐ 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. **[x]**
- ☐ 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]**



## Question 8

*Given an adversarial example  $x_{adv}$  of a clean example  $x$  w.r.t. model  $f$ ,  $y \in \{1, 2, \dots, M\}$  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
- ☐ C.  $\operatorname{argmax}_{1 \leq m \leq M} f_m(x_{adv}) = y$
- ☐ D.  $\operatorname{argmax}_{1 \leq m \leq M} f_m(x_{adv}) \neq y$

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- ☐ A.  $x_{adv}$  and  $x$  look very similar under human perspective **[x]**
- ☐ B.  $x_{adv}$  and  $x$  look very different under human perspective
- ☐ C.  $\operatorname{argmax}_{1 \leq m \leq M} f_m(x_{adv}) = y$
- ☐ D.  $\operatorname{argmax}_{1 \leq m \leq M} f_m(x_{adv}) \neq y$  **[x]**

## Question 9

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

- ☐ A. This constraint to make sure that  $x_{adv}$  and  $x$  look very similar under human perspective
- ☐ B. This constraint to make sure that  $x_{adv}$  and  $x$  look very different under human perspective
- ☐ C. This constraint to make sure that  $\operatorname{argmax}_{1 \leq m \leq M} f_m(x_{adv}) = \operatorname{argmax}_{1 \leq m \leq M} f_m(x)$
- ☐ D. The highest absolute difference between pixels of  $x_{adv}$  and  $x$  is less than or equal  $\epsilon$

## Question 9

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

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- ☐ B. This constraint to make sure that  $x_{adv}$  and  $x$  look very different under human perspective
- ☐ C. This constraint to make sure that  $\operatorname{argmax}_{1 \leq m \leq M} f_m(x_{adv}) = \operatorname{argmax}_{1 \leq m \leq M} f_m(x)$
- ☐ D. The highest absolute difference between pixels of  $x_{adv}$  and  $x$  is less than or equal  $\epsilon$  **[x]**

# Question 10

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

- ☐ A. We maximally increase the chance to predict  $x_{adv}$  with label  $y$ .
- ☐ B. We maximally decrease the chance to predict  $x_{adv}$  with label  $y$ .
- ☐ 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.

## Question 10

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

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

# Question 11

Given a DL model  $f(x; \theta)$  parameterized by  $\theta$  where  $f(x; \theta)$  represents the prediction probabilities of  $x$  associated with a ground-truth label  $y \in \{1, \dots, M\}$ , we find an adversarial example by  $\mathbf{x}_{adv} = \underset{\mathbf{x}' \in B_\epsilon(\mathbf{x})}{\operatorname{argmin}} l(f(\mathbf{x}'; \theta), \mathbf{y}_\neq)$  with  $\mathbf{y}_\neq \neq y$ .

Which statements are correct? (MC)

- ☐ A. We maximally increase the chance to predict  $x_{adv}$  with label  $y$ .
- ☐ B. We maximally increase the chance to predict  $x_{adv}$  with label  $y_\neq$ .
- ☐ C. It is a targeted attack.
- ☐ D. It is an untargeted attack.



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Given a DL model  $f(x; \theta)$  parameterized by  $\theta$  where  $f(x; \theta)$  represents the prediction probabilities of  $x$  associated with a ground-truth label  $y \in \{1, \dots, M\}$ , we find an adversarial example by  $\mathbf{x}_{adv} = \mathbf{argmin}_{x' \in B_\epsilon(x)} l(f(\mathbf{x}'; \theta), \mathbf{y}_\neq)$  with  $y_\neq \neq y$ .

Which statements are correct? (MC)

- ☐ A. We maximally increase the chance to predict  $x_{adv}$  with label  $y$ .
- ☐ B. We maximally increase the chance to predict  $x_{adv}$  with label  $y_\neq$ . **[x]**
- ☐ C. It is a targeted attack. **[x]**
- ☐ D. It is an untargeted attack.