

FIT5215 Deep Learning

Quiz for:
Convolutional Neural Network

Trung Le and Tutor Team

Department of Data Science and AI
Faculty of Information Technology, Monash University
Email: trunglm@monash.edu

Question 1

*Given an 3D input tensor with shape [3, 32, 32] over which we apply a conv2D with **16 filters** each of which has shape [3,5,5], strides [3,3], and padding=0. What is the shape of the output tensor?*

- ☐ A. [10, 10, 16]
- ☐ B. [11, 11, 16]
- ☐ C. [16 , 11, 11]
- ☐ D. [16 , 10, 10]

Question 1

*Given an 3D input tensor with shape [3, 32, 32] over which we apply a conv2D with **16 filters** each of which has shape **[3,5,5]**, strides **[3,3]**, and padding=**0**. What is the shape of the output tensor?*

- ☐ A. [10, 10, 16]
- ☐ B. [11, 11, 16]
- ☐ C. [16, 11, 11]
- ☐ D. [16 ,10, 10] [x]

$$\left\lfloor \frac{32 + 2 \times 0 - 5}{3} \right\rfloor + 1 = 10$$

Question 2

*Given an 3D input tensor with shape $[3, 32, 32]$ over which we apply a conv2D with **filters** which has shape $[32, 3, 5, 5]$, strides $[3,3]$, and padding=2. What is the shape of the output tensor?*

- ☐ A. $[32, 10, 10]$
- ☐ B. $[32, 11, 11]$
- ☐ C. $[11, 11, 16]$
- ☐ D. $[10, 10, 16]$

Question 2

*Given an 3D input tensor with shape [3, 32, 32] over which we apply a conv2D with **filters** which has shape [32, 3, 5, 5], strides [3,3], and padding=2. What is the shape of the output tensor?*

$$\left\lfloor \frac{32 + 2 \times 2 - 5}{3} \right\rfloor + 1 = 11$$

- ☐ A. [32,10,10]
- ☒ B. [32,11, 11] [x]
- ☐ C. [11,11,16]
- ☐ D. [10,10,16]

Question 3

*Given an 3D input tensor with shape [10, 64, 64] over which we apply a **max pooling** layer with kernel size [3,3], strides [2,2], and padding=1. What is the **shape** of the **output tensor**?*

- ☐ A. [10, 31, 31]
- ☐ B. [10, 32, 32]
- ☐ C. [22, 22, 10]
- ☐ D. [32, 32, 10]

Question 3

*Given an 3D input tensor with shape [10, 64, 64] over which we apply a **max pooling** layer with kernel size [3,3], strides [2,2], and padding=1. What is the **shape** of the **output tensor**?*

- ☐ A. [10, 31, 31]
- ☒ B. [10, 32, 32] [x]
- ☐ C. [22, 22, 10]
- ☐ D. [32, 32, 10]

$$\left\lfloor \frac{64 + 2 \times 1 - 3}{2} \right\rfloor + 1 = 32$$

Question 4

*Assume that the tensor before the last tensor of a CNN has shape $[32, 10, 32, 32]$ and we apply **20 filters** each of which has the shape **$[10, 5, 5]$** and **strides**= **$[2,2]$** with **padding** = **1** to obtain the last tensor. What is the **shape** of the output tensor?*

- ☐ A. $[20, 15, 15]$
- ☐ B. $[20, 16, 16]$
- ☐ C. $[32, 20, 16, 16]$
- ☐ D. $[32, 20, 15, 15]$

Question 4

Assume that the tensor before the last tensor of a CNN has shape $[32, 10, 32, 32]$ and we apply **20 filters** each of which has the shape **$[10, 5, 5]$** and **strides= $[2,2]$** with **padding = 1** to obtain the last tensor. What is the **shape** of the output tensor?

- ☐ A. $[20, 15, 15]$
- ☐ B. $[20, 16, 16]$
- ☐ C. $[32, 20, 16, 16]$
- ☐ D. $[32, 20, 15, 15]$ **[x]**

- Given an image in a minibatch, we convolve each $[10, 32, 32]$ with $[10, 5, 5]$ to achieve a 15×15 feature map.

$$\left\lfloor \frac{32 + 2 \times 1 - 5}{2} \right\rfloor + 1 = 15$$

- There are **20** filters $\rightarrow [20, 15, 15]$
- There are **32** inputs in a minibatch $\rightarrow [32, 20, 15, 15]$

Question 5

*Assume that the tensor before the last tensor of a CNN has shape $[64, 10, 32, 32]$ and we apply **5 filters** each of which has the shape $[10, 5, 5]$ and strides= $[2, 2]$ with padding = **0** to obtain the last tensor. We flatten this tensor to a **fully connected** (FC) layer. What is the **number of neurons** on this FC layer?*

- ☐ A. 14 x 14
- ☐ B. 14 x 14 x 5
- ☐ C. 64 x 16 x 16 x 5
- ☐ D. 64 x 14 x 14 x 5

Question 5

*Assume that the tensor before the last tensor of a CNN has shape [64, 10, 32, 32] and we apply **5 filters** each of which has the shape **[10, 5, 5]** and strides = **[2, 2]** with padding = **0** to obtain the last tensor. We flatten this tensor to a **fully connected (FC)** layer. What is the **number of neurons** on this FC layer?*

- ☐ A. 14 x 14
- ☐ B. 14 x 14 x 5 **[x]**
- ☐ C. 64 x 16 x 16 x 5
- ☐ D. 64 x 14 x 14 x 5

- Given an image in a minibatch, we convolve each [10 , 32, 32] with [10, 5, 5] to achieve a 14x14 feature map.

$$\left\lfloor \frac{32 + 2 \times 0 - 5}{2} \right\rfloor + 1 = 14$$

- There are **5** filters \rightarrow **[5 , 14, 14]**
- We flatten [5 , 14, 14] and obtain 14x14x5 neurons

Question 6

What likely happen if using a large filter (e.g., 7x7, 9x9) with a deep model (e.g., 20 layers) if there are few images?

- ☐ A. Overfitting
- ☐ B. Underfitting

Question 6

What likely happen if using a large filter (e.g., 7x7, 9x9) with a deep model (e.g., 20 layers) if there are few images?

☒ A. Overfitting [x]

☐ B. Underfitting

- Larger filter → More parameters → Overfitting problem
- Small filter → Fewer parameters → Underfitting problem
- If we use large enough model (several layers), 3x3 filter will be a common choice.

Question 7

Which is a good CNN model architecture?

- ☐ A. Input layer → Convolutional layer (Activation) → Pooling layer → FC layer → Output
- ☐ B. Input layer → Pooling layer → Convolutional layer (Activation) → FC layer → Output
- ☐ C. Input layer → FC Layer → Pooling layer → Convolutional layer (Activation) → Output
- ☐ D. Input layer → Convolutional layer (Activation) → FC layer → Pooling layer → Output

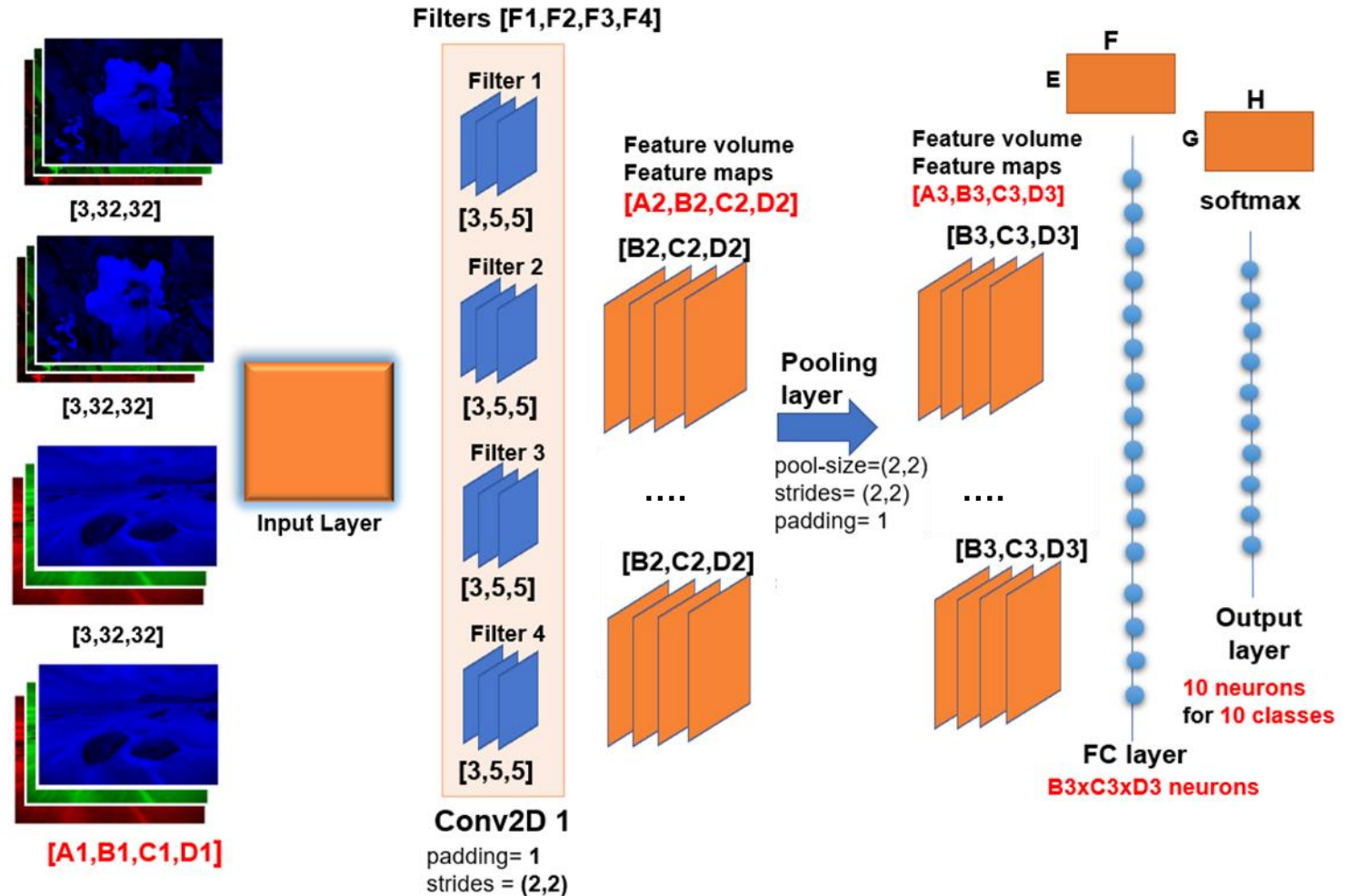
Question 7

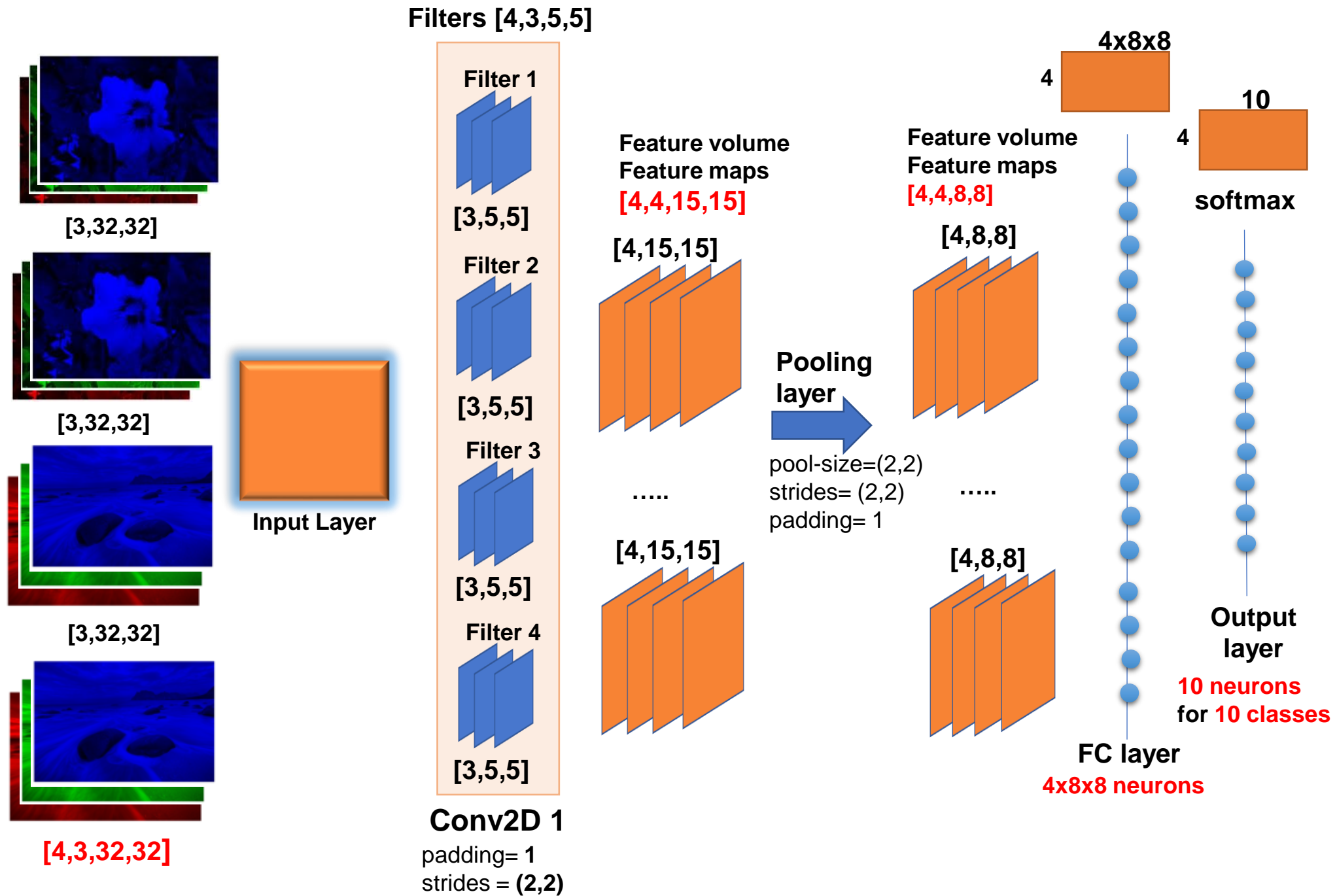
Which is a good CNN model architecture?

- ☒ A. Input layer → Convolutional layer (Activation) → Pooling layer → FC layer → Output **[x]**
- ☐ B. Input layer → Pooling layer → Convolutional layer (Activation) → FC layer → Output
- ☐ C. Input layer → FC Layer → Pooling layer → Convolutional layer (Activation) → Output
- ☐ D. Input layer → Convolutional layer (Activation) → FC layer → Pooling layer → Output

Question 8

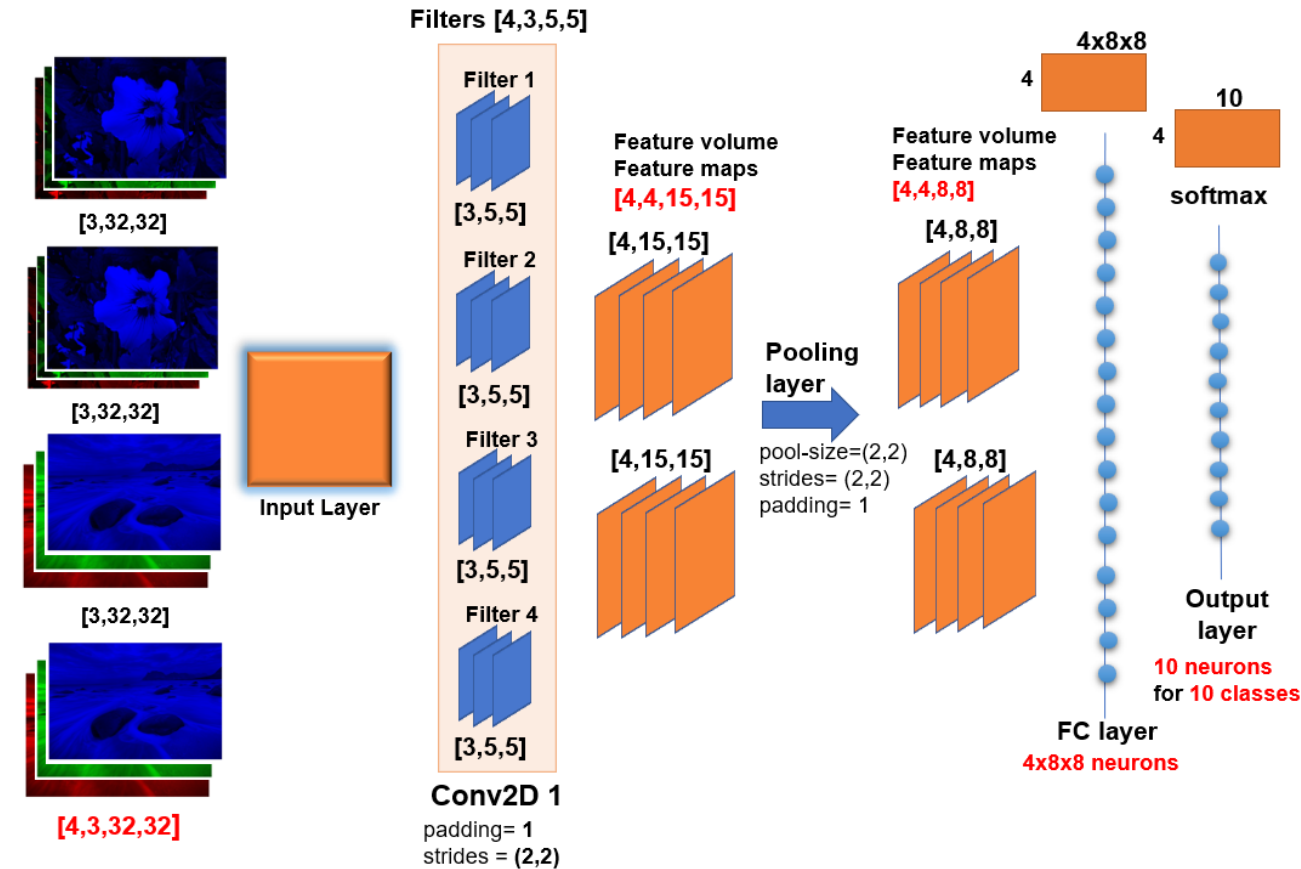
- Consider the following CNN in operation. What are $[A1, B1, C1, D1]$, $[F1, F2, F3, F4]$, $[A2, B2, C2, D2]$, $[A3, B3, C3, D3]$, $[E, F]$, $[G, H]$?





Question 9

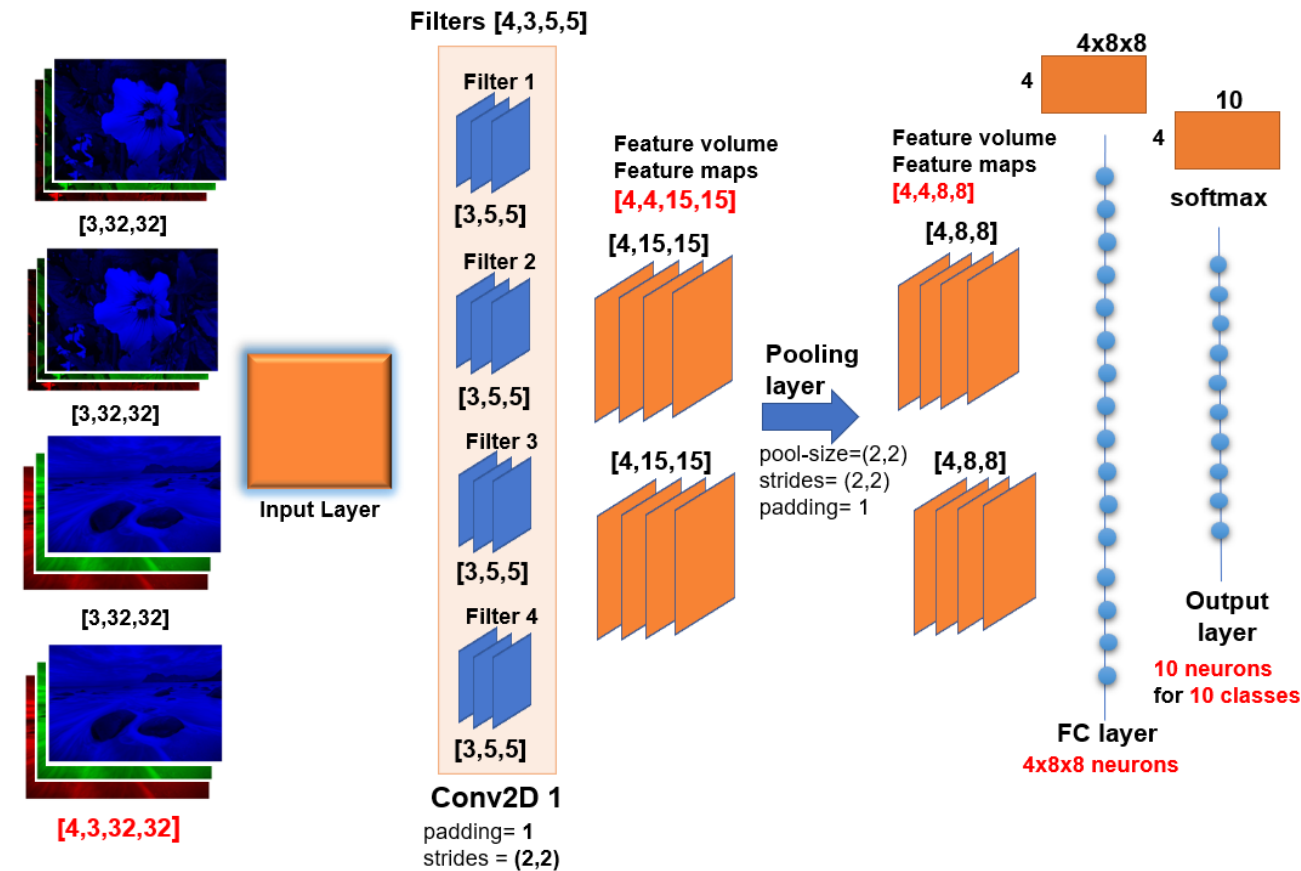
Consider the following CNN in operation. Assume that the 2D tensor $[4, 10]$ at the output layer is the one **after applied softmax**. What is the meaning of the second row in this tensor?



- ☐ A. The prediction probabilities of the second image in the batch.
- ☐ B. The logits of the second image in the batch.
- ☐ C. The prediction probabilities of the third image in the batch.
- ☐ D. The logits of the third image in the batch.

Question 9

Consider the following CNN in operation. Assume that the 2D tensor $[4, 10]$ at the output layer is the one **after applied softmax**. What is the meaning of the second row in this tensor?



- ☐ A. The prediction probabilities of the second image in the batch. [x]
- ☐ B. The logits of the second image in the batch.
- ☐ C. The prediction probabilities of the third image in the batch.
- ☐ D. The logits of the third image in the batch.

Question 10

What is the purpose of the Conv2D layers? (MC)

- ☐ A. Process 1D tensors.
- ☐ B. Process 2D and 3D tensors.
- ☐ C. Provide some filters to detect the patterns inside an input tensor.
- ☐ D. Reduce the input size to make it more manageable.

Question 10

What is the purpose of the Conv2D layers?

- ☐ A. Process 1D tensors.
- ☒ B. Process 2D and 3D tensors. [x]
- ☒ C. Provide some filters to detect the patterns inside an input tensor. [x]
- ☐ D. Reduce the input size to make it more manageable.

Question 11

What are correct about the pooling layers? (MC)

- ☐ A. Preserve the crucial information in the input.
- ☐ B. Increase the input size to have more information.
- ☐ C. Provide some filters to detect the patterns inside an input tensor.
- ☐ D. Reduce the input size to make it more manageable.

Question 11

What are correct about the pooling layers? (MC)

- ☐ A. Preserve the crucial information in the input. [x]
- ☐ B. Increase the input size to have more information.
- ☐ C. Provide some filters to detect the patterns inside an input tensor.
- ☐ D. Reduce the input size to make it more manageable. [x]

Question 12

What are correct about the BatchNorm layers? (MC)

- ☐ A. Mitigate the overfitting.
- ☐ B. In training phase, it uses the population statistics (mean and std).
- ☐ C. In training phase, it uses the batch statistics (mean and std).
- ☐ D. In testing phase, it uses the population statistics (mean and std).
- ☐ E. In testing phase, it uses the batch statistics (mean and std)

Question 12

What are correct about the BatchNorm layers? (MC)

- ☐ A. Mitigate the overfitting. **[x]**
- ☐ B. In training phase, it uses the population statistics (mean and std).
- ☐ C. In training phase, it uses the batch statistics (mean and std). **[x]**
- ☐ D. In testing phase, it uses the population statistics (mean and std). **[x]**
- ☐ E. In testing phase, it uses the batch statistics (mean and std)

Question 13

Given an implementation as below. What is the shape printed?

```
cnn_part = nn.Sequential(  
    nn.Conv2d(3, 32, kernel_size=3, stride= 2,padding=1),  
    nn.ReLU(),  
    nn.Conv2d(32, 64, kernel_size=3, stride =1, padding=1),  
    nn.ReLU()  
)  
  
x= torch.randn(1,3,224,224)  
print(cnn_part(x).shape)
```

- ☐ A. [1, 64, 112, 112]
- ☐ B. [32, 64, 112, 112]
- ☐ C. [32, 64, 224, 224]
- ☐ D. [1, 64, 224, 224]

Question 13

Given an implementation as below. What is the shape printed?

```
cnn_part = nn.Sequential(  
    nn.Conv2d(3, 32, kernel_size=3, stride= 2,padding=1),  
    nn.ReLU(),  
    nn.Conv2d(32, 64, kernel_size=3, stride =1, padding=1),  
    nn.ReLU()  
)  
  
x= torch.randn(1,3,224,224)  
print(cnn_part(x).shape)
```

- ☐ A. [1, 64, 112, 112] **[x]**
- ☐ B. [32, 64, 112, 112]
- ☐ C. [32, 64, 224, 224]
- ☐ D. [1, 64, 224, 224]

Question 14

Given an implementation as below. What is the shape printed?

```
cnn_part = nn.Sequential(  
    nn.Conv2d(3, 32, kernel_size=3, stride= 2,padding=1),  
    nn.ReLU(),  
    nn.Conv2d(32, 64, kernel_size=3, stride =1, padding=1),  
    nn.BatchNorm2d(64),  
    nn.ReLU(),  
    nn.MaxPool2d(kernel_size=2, stride=2, padding = 0)  
)  
  
x= torch.randn(64,3,224,224)  
print(cnn_part(x).shape)
```

- ☐ A. [1, 64, 112, 112]
- ☐ B. [64, 64, 56, 56]
- ☐ C. [64, 64, 224, 224]
- ☐ D. [1, 64, 56, 56]

Question 14

Given an implementation as below. What is the shape printed?

```
cnn_part = nn.Sequential(  
    nn.Conv2d(3, 32, kernel_size=3, stride= 2,padding=1),  
    nn.ReLU(),  
    nn.Conv2d(32, 64, kernel_size=3, stride =1, padding=1),  
    nn.BatchNorm2d(64),  
    nn.ReLU(),  
    nn.MaxPool2d(kernel_size=2, stride=2, padding = 0)  
)  
  
x= torch.randn(64,3,224,224)  
print(cnn_part(x).shape)
```

- ☐ A. [1, 64, 112, 112]
- ☒ B. [64, 64, 56, 56] [x]
- ☐ C. [64, 64, 224, 224]
- ☐ D. [1, 64, 56, 56]

Question 15

Given an implementation as below. What is the shape printed?

```
cnn_part = nn.Sequential(  
    nn.Conv2d(3, 32, kernel_size=3, stride= 2,padding=1),  
    nn.ReLU(),  
    nn.Conv2d(32, 64, kernel_size=3, stride =1, padding=1),  
    nn.BatchNorm2d(64),  
    nn.ReLU(),  
    nn.MaxPool2d(kernel_size=2, stride=2, padding = 0),  
    nn.Flatten(1),  
    nn.Linear(64*56*56, 26)  
)  
  
x= torch.randn(64,3,224,224)  
print(cnn_part(x).shape)
```

- ☐ A. [64, 64*56*56]
- ☐ B. [64, 26]
- ☐ C. [64, 64, 56, 56]
- ☐ D. [1, 26]

Question 15

Given an implementation as below. What is the shape printed?

```
cnn_part = nn.Sequential(  
    nn.Conv2d(3, 32, kernel_size=3, stride= 2,padding=1),  
    nn.ReLU(),  
    nn.Conv2d(32, 64, kernel_size=3, stride =1, padding=1),  
    nn.BatchNorm2d(64),  
    nn.ReLU(),  
    nn.MaxPool2d(kernel_size=2, stride=2, padding = 0),  
    nn.Flatten(1),  
    nn.Linear(64*56*56, 26)  
)  
  
x= torch.randn(64,3,224,224)  
print(cnn_part(x).shape)
```

- ☐ A. [64, 64*56*56]
- ☐ B. [64, 26] **[x]**
- ☐ C. [64, 64, 56, 56]
- ☐ D. [1, 26]