



## Deep Learning

### Assignment- Week 8

TYPE OF QUESTION: MCQ/MSQ

Number of questions: 10

Total mark: 10 X 1 = 10

#### QUESTION 1:

Which of the following is false about CNN?

- a. Output should be flattened before feeding it to a fully connected layer
- b. There can be only 1 fully connected layer in CNN
- c. We can use any many convolutional layers in CNN
- d. None of the above

**Correct Answer: b**

**Detailed Solution:**

**Direct from classroom lecture**

#### QUESTION 2:

The input image has been converted into a matrix of size 64 X 64 and a kernel/filter of size 5x5 with a stride of 1 and no padding. What will be the size of the convoluted matrix?

- a. 5x5
- b. 59x59
- c. 64x64
- d. 60x60

**Correct Answer: d**

**Detailed Solution:**

The size of the convoluted matrix is given by  $C \times C$  where  $C = ((I - F + 2P) / S) + 1$ , where  $C$  is the size of the Convoluted matrix,  $I$  is the size of the input matrix,  $F$  the size of the filter matrix and  $P$  the padding applied to the input matrix. Here  $P=0$ ,  $I=64$ ,  $F=5$  and  $S=1$ . Therefore, the answer is 60x60.

#### QUESTION 3:



Filter size of 3x3 is convolved with matrix of size 4x4 (stride=1). What will be the size of output matrix if valid padding is applied:

- a. 4x4
- b. 3x3
- c. 2x2
- d. 1x1

**Correct Answer: c**

**Detailed Solution:**

**This type is used when there is no requirement for Padding. The output matrix after convolution will have the dimension of  $((n - f + 2P)/S + 1) \times ((n - f + 2P)/S + 1)$**

---

#### **QUESTION 4:**

Let us consider a Convolutional Neural Network having three different convolutional layers in its architecture as:

**Layer-1:** Filter Size – 3 X 3, Number of Filters – 10, Stride – 1, Padding – 0

**Layer-2:** Filter Size – 5 X 5, Number of Filters – 20, Stride – 2, Padding – 0

**Layer-3:** Filter Size – 5 X 5, Number of Filters – 40, Stride – 2, Padding – 0

Layer 3 of the above network is followed by a fully connected layer. If we give a 3-D image input of dimension 39 X 39 to the network, then which of the following is the input dimension of the fully connected layer.

- a. 1960
- b. 2200
- c. 4563
- d. 13690



**Correct Answer: a**

**Detailed Solution:**

the input image of dimension  $39 \times 39 \times 3$  convolves with 10 filters of size  $3 \times 3$  and takes the Stride as 1 with no padding. After these operations, we will get an output of  $37 \times 37 \times 10$ .

Output of layer 2 would be  $17 \times 17 \times 20$

Output of layer 3 would be  $7 \times 7 \times 40$ . Flattening this gives 1960.

---

### **QUESTION 5:**

Suppose you have 40 convolutional kernel of size  $3 \times 3$  with no padding and stride 1 in the first layer of a convolutional neural network. You pass an input of dimension  $1024 \times 1024 \times 3$  through this layer. What are the dimensions of the data which the next layer will receive?

- a.  $1020 \times 1020 \times 40$
- b.  $1022 \times 1022 \times 40$
- c.  $1021 \times 1021 \times 40$
- d.  $1022 \times 1022 \times 3$

**Correct Answer: b**

**Detailed Solution:**

The layer accepts a volume of size  $W_1 \times H_1 \times D_1$ . In our case,  $1024 \times 1024 \times 3$

Requires four hyperparameters: Number of filters  $K=40$ , their spatial extent  $F=3$ , the stride  $S=1$ , the amount of padding  $P=0$ .

Produces a volume of size  $W_2 \times H_2 \times D_2$  i.e. where:  $W_2 = (W_1 - F + 2P) / S + 1 = (1024 - 3) / 1 + 1 = 1022$ ,  $H_2 = (H_1 - F + 2P) / S + 1 = (1024 - 3) / 1 + 1 = 1022$ , (i.e. width and height are computed equally by symmetry),  $D_2 = \text{Number of filters } K = 40$ .

---



---

**QUESTION 6:**

Consider a CNN model which aims at classifying an image as either a rose, or a marigold, or a lily or an orchid (consider the test image can have only 1 of the classes at a time) . The last (fully-connected) layer of the CNN outputs a vector of logits,  $L$ , that is passed through a \_\_\_\_ activation that transforms the logits into probabilities,  $P$ . These probabilities are the model predictions for each of the 4 classes. Fill in the blanks with the appropriate option.

Fill in the blanks with the appropriate option.

- a. Leaky ReLU
- b. Tanh
- c. ReLU
- d. Softmax

**Correct Answer: d**

**Detailed Solution:**

**Softmax works best if there is one true class per example, because it outputs a probability vector whose entries sum to 1.**

---

**QUESTION 7:**

Suppose your input is a 300 by 300 color (RGB) image, and you use a convolutional layer with 100 filters that are each 5x5. How many parameters does this hidden layer have (without bias)

- a. 2501
- b. 2600
- c. 7500
- d. 7600

**Correct Answer: c**

**Detailed Solution:**

**As we have a RGB Image so each filter would be 3D, whose dimension is  $5 * 5 * 3 = 75$**

**Now we have 100 such filters. Now, as there is no bias so, total number of parameters =  $5 * 5 * 3 * 100 = 7500$**



---

**QUESTION 8:**

Which of the following activation functions can lead to vanishing gradients?

- a. ReLU
- b. Sigmoid
- c. Leaky ReLU
- d. None of the above

**Correct Answer: b**

**Detailed Solution:**

**For sigmoid activation, a large change in the input of the sigmoid function will cause a small change in the output. Hence, the derivative becomes small. When more and more layers uses such activation, the gradient of the loss function becomes very small making the network difficult to train.**

---

**QUESTION 9:**

Statement 1: Residual networks can be a solution for vanishing gradient problem

Statement 2: Residual networks provide residual connections straight to earlier layers

Statement 3: Residual networks can never be a solution for vanishing gradient problem

Which of the following option is correct?

- a. Statement 2 is correct
- b. Statement 3 is correct
- c. Both Statement 1 and Statement 2 are correct
- d. Both Statement 2 and Statement 3 are correct

**Correct Answer: c**

**Detailed Solution:**



Residual networks can be a solution to vanishing gradient problems, as they provide residual connections straight to earlier layers. This residual connection doesn't go through activation functions that “squashes” the derivatives, resulting in a higher overall derivative of the block.

---

**QUESTION 10:**

Input to SoftMax activation function is [0.5,0.5,1]. What will be the output?

- a. [0.28,0.28,0.44]
- b. [0.022,0.956, 0.022]
- c. [0.045,0.910,0.045]
- d. [0.42, 0.42,0.16]

**Correct Answer: a**

**Detailed Solution:**

$$\text{SoftMax, } \sigma(x_j) = \frac{e^{x_j}}{\sum_{k=1}^n e^{x_k}} \text{ for } j=1,2,\dots,n$$

$$\text{Therefore, } \sigma(0.5) = \frac{e^{0.5}}{\sum_{k=1}^n e^{x_k}} = 0.28 \text{ and similarly the other values}$$

---

---

\*\*\*\*\*END\*\*\*\*\*