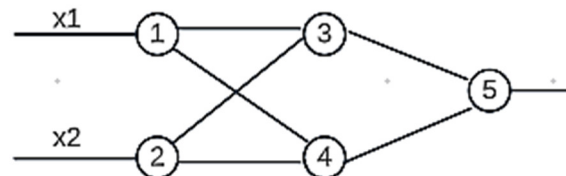


- OR iii. Assume that the neurons have a sigmoid activation function. Perform a forward pass and a backward pass on the network. Assume that the actual output of y is 0.5 and learning rate is 1. Perform another forward pass.



Take

$x_1=0.35$        $x_2=0.9$   
 $w_{13}=0.1$        $w_{14}=0.4$   
 $w_{23}=0.8$        $w_{24}=0.6$        $w_{35}=0.3$        $w_{45}=0.9$

- Q.5 i. Differentiate between single-channel and multi-channel data in ConvNet. **3** 1 1 2 4 2  
 ii. Explain how convolutional layers and pooling layers act as strong priors in image recognition tasks. **7** 2 1 2 4 2  
 OR iii. Describe the effect of increasing the stride in a convolutional layer. How does this impact the size of the output feature map? **7** 2 1 2 4 2  
 Q.6 i. What is the attention mechanism in neural networks? **3** 1 1 2 5 2  
 ii. Explain the architecture of Generative Adversarial Networks (GANs). **7** 2 1 2 5 2  
 OR iii. What are the key ethical considerations in the development and deployment of AI and neural networks? **7** 2 1 2 5 2

\*\*\*\*\*

Total No. of Questions: 6

Total No. of Printed Pages:4

Enrollment No.....



Faculty of Engineering  
 End Sem Examination Dec 2024

RA3EL26 Neural Network

Programme: B.Tech.

Branch/Specialisation: RA

Duration: 3 Hrs.

Maximum Marks: 60

Note: All questions are compulsory. Internal choices, if any, are indicated. Answers of Q.1 (MCQs) should be written in full instead of only a, b, c or d. Assume suitable data if necessary. Notations and symbols have their usual meaning.

- |  | Marks | BL | PO      | CO | PSO |
|--|-------|----|---------|----|-----|
| Q.1 i. The three primary layers in a neural network system are:                  | 1     | 1  | 1 2 4,5 | 1  | 2   |
| (a) Input, hidden, and output layers   |       |    |         |    |     |
| (b) Axon, dendrite, and soma   |       |    |         |    |     |
| (c) Training, validation, and testing layers                                     |       |    |         |    |     |
| (d) Activation, feedforward, and feedback layers                                 |       |    |         |    |     |
| ii. Which of the following accurately describes a Single-Layer Perceptron (SLP)? | 1     | 1  | 1 2 4   | 1  | 2   |
| (a) It consists of one hidden layer between the input and output                 |       |    |         |    |     |
| (b) It can solve both linearly and non-linearly separable problems               |       |    |         |    |     |
| (c) It can only classify linearly separable problems                             |       |    |         |    |     |
| (d) It uses the sigmoid activation function exclusively                          |       |    |         |    |     |
| iii. Which of the following is true about Recurrent Neural Networks (RNNs)?      | 1     | 1  | 1 2 4   | 2  | 2   |
| (a) They do not have loops   |       |    |         |    |     |
| (b) They are suitable for processing sequential data                             |       |    |         |    |     |
| (c) They work only with feedforward connections                                  |       |    |         |    |     |
| (d) They require no memory of previous states                                    |       |    |         |    |     |
| iv. In a feedforward neural network, information flows:                          | 1     | 1  | 5       | 2  | 2   |
| (a) In cycles  |       |    |         |    |     |
| (b) Back and forth   |       |    |         |    |     |
| (c) In one direction   |       |    |         |    |     |
| (d) Randomly between neurons   |       |    |         |    |     |

P.T.O.



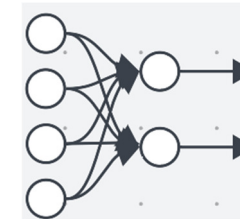
**Marking Scheme**  
**RA3EL26 (T) Neural Network (T)**

- Q.1 i) a) Input, hidden, and output layers **1**
- ii) c) It can only classify linearly separable problems **1**
- iii) b) They are suitable for processing sequential data **1**
- iv) c) Only in one direction **1**
- v) b) To randomly deactivate neurons during training to prevent overfitting **1**
- vi) a) Modifying a pre-trained model by adjusting its parameters to a new task **1**
- vii) b) To perform element-wise multiplication with the input data **1**
- viii) b) A separate kernel is used for each channel **1**
- ix) c) To make AI decisions interpretable and understandable **1**
- x) b) Unfair or discriminatory decision-making **1**

- Q.2 i. Draw a simple neural network having one input layer, 2 hidden layer and one output layer.....**1.5 marks**  
Also write the names of various terminologies related to Neural network. ....**1.5 marks**
- ii. Explain biological neuron with the help of a diagram. ....**4 marks**  
Also, write any 3 differences between biological neuron and artificial neuron.....**3 marks**
- OR iii. Implement XOR logic function using McCulloch-Pitts Model. ....**5 marks**  
Find the values of threshold and weights. ....**2 marks**

- Q.3 i. Explain Radial Basis Function (RBF) networks.....**2 marks** **2**

- ii. What are activation Functions. ....**4 marks** **8**  
Explain various types of activation functions used in the neural network. ....**4 marks**
- OR iii. Consider the network shown in fig which considers four training samples each of length 4 and two output units **8**



Train the SOM network by determining the class membership of the input data

X1= [1,0,1,0]      X2= [1,0,0,0]

X3= [1,1,1,1]      X4= [0,1,1,0]

Output units:

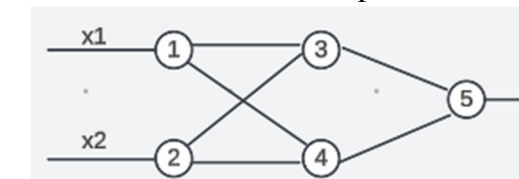
[Unit 1] = [0.3,0.5,0.7,0.2]

[Unit 2] = [0.6,0.5,0.4,0.2]

Take learning rate= 0.6

**Finding answer.....8 marks**

- Q.4 i. What is transfer learning in neural networks.....**2 marks** **2**
- ii. Discuss the role of regularization techniques like dropout and weight decay in neural networks. ....**4 marks** **8**  
Dropout.....**4 marks**  
Weight decay.....**4 marks**
- OR iii. Assume that the neurons have a sigmoid activation function. ....**8**  
Perform a forward pass and a backward pass on the network. Assume that the actual output of y is 0.5 and learning rate is 1. Perform another forward pass.



Take x1=0.35      x2= 0.9

w13= 0.1      w14= 0.4

[2]

$$w_{23} = 0.8 \quad w_{24} = 0.6$$

$$w_{35} = 0.3 \quad w_{45} = 0.9$$

**Finding answer.....8 marks**

- Q.5 i. Differentiate between single-channel and multi-channel data in ConvNet. ....3marks **3**
- ii. Explain how convolutional layers and pooling layers act as strong priors in image recognition tasks. convolutional layers .....3 marks pooling layers .....4 marks **7**
- OR iii. Describe the effect of increasing the stride in a convolutional layer. **Finding answer.....4 marks** How does this impact the size of the output feature map? .....3 marks **7**
- Q.6 i. What is the attention mechanism in neural networks? .....3 marks **3**
- ii. Explain the architecture of Generative Adversarial Networks (GANs). .....7 marks **7**
- OR iii. What are the key ethical considerations in the development and deployment of AI and neural networks? .....7 marks **7**

\*\*\*\*\*

**Q. 3 (iii)****Given Data:****1. Input vectors:**

- $X_1 = [1, 0, 1, 0]$
- $X_2 = [1, 0, 0, 0]$
- $X_3 = [1, 1, 1, 1]$
- $X_4 = [0, 1, 1, 0]$

[3]

**2. Output unit weights:**

- Unit 1:  $W_1 = [0.3, 0.5, 0.7, 0.2]$
- Unit 2:  $W_2 = [0.6, 0.5, 0.4, 0.2]$

**3. Learning rate:  $\eta = 0.6$** **Step 1: Process Input  $X_1 = [1, 0, 1, 0]$** **1.1 Compute Euclidean Distance**

- For Unit 1 ( $W_1 = [0.3, 0.5, 0.7, 0.2]$ ):

$$D_1 = \sqrt{(1 - 0.3)^2 + (0 - 0.5)^2 + (1 - 0.7)^2 + (0 - 0.2)^2}$$

$$D_1 = \sqrt{(0.7)^2 + (-0.5)^2 + (0.3)^2 + (-0.2)^2} = \sqrt{0.49 + 0.25 + 0.09 + 0.04} = \sqrt{0.87} \approx 0.933$$

- For Unit 2 ( $W_2 = [0.6, 0.5, 0.4, 0.2]$ ):

$$D_2 = \sqrt{(1 - 0.6)^2 + (0 - 0.5)^2 + (1 - 0.4)^2 + (0 - 0.2)^2}$$

$$D_2 = \sqrt{(0.4)^2 + (-0.5)^2 + (0.6)^2 + (-0.2)^2} = \sqrt{0.16 + 0.25 + 0.36 + 0.04} = \sqrt{0.81} = 0.9$$

**1.2 Determine the Winning Unit**

- $D_1 \approx 0.933, D_2 = 0.9$
- Unit 2 wins because it has the smaller distance.

**1.3 Update Weights for Winning Unit (Unit 2)**

Using the weight update rule:

$$w_i(t+1) = w_i(t) + \eta \cdot (x_i - w_i(t))$$

- For  $W_2 = [0.6, 0.5, 0.4, 0.2]$ :

$$w_2(1) = 0.6 + 0.6 \cdot (\downarrow 0.6) = 0.6 + 0.6 \cdot 0.4 = 0.84$$

$$w_2(2) = 0.5 + 0.6 \cdot (0 - 0.5) = 0.5 + 0.6 \cdot (-0.5) = 0.2$$

$$w_2(3) = 0.4 + 0.6 \cdot (1 - 0.4) = 0.4 + 0.6 \cdot 0.6 = 0.76$$

$$w_2(4) = 0.2 + 0.6 \cdot (0 - 0.2) = 0.2 + 0.6 \cdot (-0.2) = 0.08$$

Updated weights for Unit 2:

$$W_2 = [0.84, 0.2, 0.76, 0.08]$$



[2]

Step 2: Process Input  $X_2 = [1, 0, 0, 0]$ 

## 2.1 Compute Euclidean Distance

- For Unit 1 ( $W_1 = [0.3, 0.5, 0.7, 0.2]$ ):

$$D_1 = \sqrt{(1 - 0.3)^2 + (0 - 0.5)^2 + (0 - 0.7)^2 + (0 - 0.2)^2}$$

$$D_1 = \sqrt{(0.7)^2 + (-0.5)^2 + (-0.7)^2 + (-0.2)^2} = \sqrt{0.49 + 0.25 + 0.49 + 0.04} = \sqrt{1.27} \approx 1.127$$

- For Unit 2 ( $W_2 = [0.84, 0.2, 0.76, 0.08]$ ) (updated weights):

$$D_2 = \sqrt{(1 - 0.84)^2 + (0 - 0.2)^2 + (0 - 0.76)^2 + (0 - 0.08)^2}$$

$$D_2 = \sqrt{(0.16)^2 + (-0.2)^2 + (-0.76)^2 + (-0.08)^2} = \sqrt{0.0256 + 0.04 + 0.5776 + 0.0064} = \sqrt{0.6496} \approx 0.805$$

## 2.2 Determine the Winning Unit

- $D_1 \approx 1.127, D_2 \approx 0.805$
- Unit 2 wins again because it has the smaller distance.

## 2.3 Update Weights for Winning Unit (Unit 2)

Using the update rule for  $W_2 = [0.84, 0.2, 0.76, 0.08]$ :

$$w_i(t+1) = w_i(t) + \eta \cdot (x_i - w_i(t))$$

- $w_2(1) = 0.84 + 0.6 \cdot (1 - 0.84) = 0.84 + 0.06 = 0.936$
- $w_2(2) = 0.2 + 0.6 \cdot (0 - 0.2) = 0.2 + 0.6 \cdot (-0.2) = 0.08$
- $w_2(3) = 0.76 + 0.6 \cdot (0 - 0.76) = 0.76 + 0.6 \cdot (-0.76) = 0.304$
- $w_2(4) = 0.08 + 0.6 \cdot (0 - 0.08) = 0.08 + 0.6 \cdot (-0.08) = 0.032$

Updated weights for Unit 2:

$$W_2 = [0.936, 0.08, 0.304, 0.032]$$

Step 3: Process Input  $X_3 = [1, 1, 1, 1]$ 

## 3.1 Compute Euclidean Distance

- For Unit 1 ( $W_1 = [0.3, 0.5, 0.7, 0.2]$ ):

$$D_1 = \sqrt{(1 - 0.3)^2 + (1 - 0.5)^2 + (1 - 0.7)^2 + (1 - 0.2)^2}$$

$$D_1 = \sqrt{(0.7)^2 + (0.5)^2 + (0.3)^2 + (0.8)^2} = \sqrt{0.49 + 0.25 + 0.09 + 0.64} = \sqrt{1.47} \approx 1.213$$

- For Unit 2 ( $W_2 = [0.936, 0.08, 0.304, 0.032]$ ):

$$D_2 = \sqrt{(1 - 0.936)^2 + (1 - 0.08)^2 + (1 - 0.304)^2 + (1 - 0.032)^2}$$

$$D_2 = \sqrt{(0.064)^2 + (0.92)^2 + (0.696)^2 + (0.968)^2}$$

[3]

$$D_2 = \sqrt{0.0041 + 0.8464 + 0.4844 + 0.9370} = \sqrt{2.2719} \approx 1.507$$

## 3.2 Determine the Winning Unit

- $D_1 \approx 1.213, D_2 \approx 1.507$
- Unit 1 wins because it has the smaller distance.

## 3.3 Update Weights for Winning Unit (Unit 1)

Using the update rule for  $W_1 = [0.3, 0.5, 0.7, 0.2]$ :

$$w_i(t+1) = w_i(t) + \eta \cdot (x_i - w_i(t))$$

- $w_1(1) = 0.3 + 0.6 \cdot (1 - 0.3) = 0.3 + 0.6 \cdot 0.7 = 0.72$
- $w_1(2) = 0.5 + 0.6 \cdot (1 - 0.5) = 0.5 + 0.6 \cdot 0.5 = 0.8$
- $w_1(3) = 0.7 + 0.6 \cdot (1 - 0.7) = 0.7 + 0.6 \cdot 0.3 = 0.88$
- $w_1(4) = 0.2 + 0.6 \cdot (1 - 0.2) = 0.2 + 0.6 \cdot 0.8 = 0.68$

Updated weights for Unit 1:

$$W_1 = [0.72, 0.8, 0.88, 0.68]$$

Step 4: Process Input  $X_4 = [0, 1, 1, 0]$ 

## 4.1 Compute Euclidean Distance

- For Unit 1 ( $W_1 = [0.72, 0.8, 0.88, 0.68]$ ):

$$D_1 = \sqrt{(0 - 0.72)^2 + (1 - 0.8)^2 + (1 - 0.88)^2 + (0 - 0.68)^2}$$

$$D_1 = \sqrt{(0.72)^2 + (0.2)^2 + (0.12)^2 + (0.68)^2}$$

$$D_1 = \sqrt{0.5184 + 0.04 + 0.0144 + 0.4624} = \sqrt{1.0352} \approx 1.017$$

[2]

- For Unit 2 ( $W_2 = [0.936, 0.08, 0.304, 0.032]$ ):

$$D_2 = \sqrt{(0 - 0.936)^2 + (1 - 0.08)^2 + (1 - 0.304)^2 + (0 - 0.032)^2}$$

$$D_2 = \sqrt{(0.936)^2 + (0.92)^2 + (0.696)^2 + (0.032)^2}$$

$$D_2 = \sqrt{0.876 + 0.8464 + 0.4844 + 0.0010} = \sqrt{2.2078} \approx 1.486$$

#### 4.2 Determine the Winning Unit

- $D_1 \approx 1.017, D_2 \approx 1.486$
- Unit 1 wins because it has the smaller distance.

### Final Results

The updated weights are:

- Unit 1:  $W_1 = [0.72, 0.8, 0.88, 0.68]$
- Unit 2:  $W_2 = [0.936, 0.08, 0.304, 0.032]$

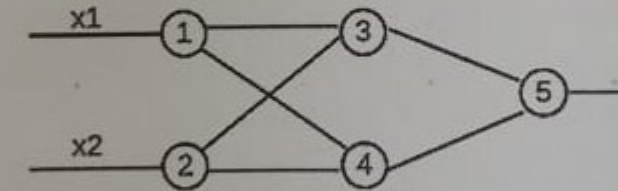
[3]

Q. 4 (iii)

[4]

- iii. Assume that the neurons have a sigmoid activation function. Perform a forward pass and a backward pass on the network. Assume that the actual output of  $y$  is 0.5 and learning rate is 1. Perform another forward pass.

8



Take

$$x_1 = 0.35$$

$$x_2 = 0.9$$

$$w_{13} = 0.1$$

$$w_{14} = 0.4$$

$$w_{23} = 0.8$$

$$w_{24} = 0.6$$

$$w_{35} = 0.3$$

$$w_{45} = 0.9$$

#### Given Data:

- Inputs:  $x_1 = 0.35, x_2 = 0.9$
- Weights:
  - $w_{13} = 0.1, w_{14} = 0.4, w_{23} = 0.8, w_{24} = 0.6, w_{35} = 0.3, w_{45} = 0.9$
- Actual output:  $y = 0.5$
- Activation function: Sigmoid  $\sigma(z) = \frac{1}{1+e^{-z}}$
- Learning rate:  $\eta = 1$

[2]

## Step 1: Forward Pass (Initial Weights)

### Hidden Layer Calculations:

The net input to hidden layer neurons  $h_3$  and  $h_4$  is:

$$h_j = \sigma(\text{Net}_j) \quad \text{where} \quad \text{Net}_j = x_1 \cdot w_{1j} + x_2 \cdot w_{2j}$$

1. For  $h_3$ :

$$\text{Net}_3 = (0.35)(0.1) + (0.9)(0.8) = 0.035 + 0.72 = 0.755$$

$$h_3 = \sigma(0.755) = \frac{1}{1 + e^{-0.755}} \approx 0.680$$

2. For  $h_4$ :

$$\text{Net}_4 = (0.35)(0.4) + (0.9)(0.6) = 0.14 + 0.54 = 0.68$$

$$h_4 = \sigma(0.68) = \frac{1}{1 + e^{-0.68}} \approx 0.663$$

### Output Layer Calculation:

The net input to output neuron  $o_5$  is:

$$\text{Net}_5 = h_3 \cdot w_{35} + h_4 \cdot w_{45}$$

$$\text{Net}_5 = (0.680)(0.3) + (0.663)(0.9) = 0.204 + 0.5967 = 0.8007$$

The output at  $o_5$  is:

$$o_5 = \sigma(0.8007) = \frac{1}{1 + e^{-0.8007}} \approx 0.690$$

[3]

## Step 2: Backward Pass (Error Calculation and Weight Updates)

### Error at the Output Layer:

The error is given as the difference between actual output  $y = 0.5$  and predicted output  $o_5 = 0.690$ :

$$\delta_5 = (y - o_5) \cdot \sigma'(0.8007)$$

The derivative of sigmoid:

$$\sigma'(z) = \sigma(z) \cdot (1 - \sigma(z))$$

$$\sigma'(0.8007) = 0.690 \cdot (1 - 0.690) = 0.690 \cdot 0.310 = 0.2139$$

Thus:

$$\delta_5 = (0.5 - 0.690) \cdot 0.2139 = -0.190 \cdot 0.2139 \approx -0.0406$$

### Update Weights to Output Layer:

Using the weight update rule:

$$\Delta w_{j5} = \eta \cdot \delta_5 \cdot h_j$$

1. For  $w_{35}$ :

$$\Delta w_{35} = 1 \cdot (-0.0406) \cdot 0.680 \approx -0.0276$$

$$w_{35} = 0.3 + (-0.0276) = 0.2724$$

2. For  $w_{45}$ :

$$\Delta w_{45} = 1 \cdot (-0.0406) \cdot 0.663 \approx -0.0269$$

$$w_{45} = 0.9 + (-0.0269) = 0.8731$$

[2]

**Error Propagation to Hidden Layer:**

The error at hidden layer neurons  $h_3$  and  $h_4$  is:

$$\delta_j = \sigma'(h_j) \cdot (\delta_5 \cdot w_{j5})$$

1. For  $h_3$ :

$$\sigma'(h_3) = h_3 \cdot (1 - h_3) = 0.680 \cdot (1 - 0.680) = 0.680 \cdot 0.320 = 0.2176$$

$$\delta_3 = 0.2176 \cdot (-0.0406) \cdot 0.3 \approx -0.0026$$

2. For  $h_4$ :

$$\sigma'(h_4) = h_4 \cdot (1 - h_4) = 0.663 \cdot (1 - 0.663) = 0.663 \cdot 0.337 = 0.2235$$

$$\delta_4 = 0.2235 \cdot (-0.0406) \cdot 0.9 \approx -0.0082$$

**Update Weights to Hidden Layer:**

1. For  $w_{13}$ :

$$\Delta w_{13} = 1 \cdot (-0.0026) \cdot x_1 = -0.0026 \cdot 0.35 \approx -0.00091$$

$$w_{13} = 0.1 + (-0.00091) = 0.0991$$

2. For  $w_{14}$ :

$$\Delta w_{14} = 1 \cdot (-0.0082) \cdot x_1 = -0.0082 \cdot 0.35 \approx -0.00287$$

[3]

$$w_{14} = 0.4 + (-0.00287) = 0.3971$$

3. For  $w_{23}$ :

$$\Delta w_{23} = 1 \cdot (-0.0026) \cdot x_2 = -0.0026 \cdot 0.9 \approx -0.00234$$

$$w_{23} = 0.8 + (-0.00234) = 0.7977$$

4. For  $w_{24}$ :

$$\Delta w_{24} = 1 \cdot (-0.0082) \cdot x_2 = -0.0082 \cdot 0.9 \approx -0.00738$$

$$w_{24} = 0.6 + (-0.00738) = 0.5926$$

**Updated weights:**

- $w_{13} = 0.0991$
- $w_{14} = 0.3971$
- $w_{23} = 0.7977$
- $w_{24} = 0.5926$
- $w_{35} = 0.2724$
- $w_{45} = 0.8731$