

**Instructions:** All questions are compulsory. Use of non-programmable scientific calculator is allowed.

- Q.1** (a) Define deep learning and explain how it differs from traditional machine learning approaches. (07)
- (b) What are the different types of activation functions popularly used? Explain each of them. (07)
- OR**
- (a) Explain in detail the basic structure of Perceptron. (07)
- (b) Describe the backpropagation algorithm and its significance in training ANNs. (07)
- Q.2** (a) Explain, in detail, the basic structure of Multi-Layer Perceptron. (07)
- (b) A neuron has two binary inputs and uses a threshold activation function. Its weights are  $w_1 = 2$  and  $w_2 = 1$ , and the bias is  $b = -2.5$ . Determine the output of the neuron for each of the following inputs: (0, 0), (1, 0), (0, 1), and (1, 1). Also, identify the logical function (if any) this neuron is computing. (07)
- OR**
- (a) Explain the working of dropout, including their mathematical operations and regularizing effects. (07)
- (b) What is the vanishing and exploding gradient problem? Provide an example where the gradients vanish for the sigmoid activation function. (07)
- Q.3** (a) Describe the theoretical and practical impact of batch normalization, including their mathematical formulations and roles in model training. (07)
- (b) A Convolutional Neural Network (CNN) receives an input image of size  $28 \times 28 \times 3$ . The network is composed of the following layers: (07)
- Convolutional Layer 1: 16 filters of size  $5 \times 5$ , stride 1, padding 2
  - Max Pooling Layer 1:  $2 \times 2$  window, stride 2
  - Convolutional Layer 2: 32 filters of size  $3 \times 3$ , stride 1, padding 1
  - Max Pooling Layer 2:  $2 \times 2$  window, stride 2
  - Fully Connected Layer: 256 neurons
  - Output Layer: 10 neurons for classification
- Answer the following:
- Calculate the output dimensions (height  $\times$  width  $\times$  depth) after each layer.
  - Calculate the total number of trainable parameters (weights + biases) in each convolutional and fully connected layer.
  - What is the total number of trainable parameters in the entire network?
- OR**
- (a) Compare and contrast Convolutional Neural Networks (CNNs) and Deep Neural Networks (DNNs) in terms of their architectural characteristics, methods of processing data, and common application areas. (07)
- (b) Discuss the concept of receptive fields in CNNs. How does it relate to feature extraction and spatial hierarchy in image data? (07)

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- Q.4** (a) Explain the various types of sequence modeling based on input and output sequence structures. Include relevant examples for each category. (07)
- (b) Explain the working of a vanilla Recurrent Neural Network (RNN) and describe its layer structure. Use a neat and labeled diagram to illustrate how it maintains memory of previous inputs over time (07)

**OR**

- (a) Discuss the strengths and limitations of Recurrent Neural Networks (RNNs) in handling sequential data. (07)
- (b) What are Gated Recurrent Units (GRUs)? Explain the architecture of a GRU cell with the help of a labeled diagram. (07)

**Q.5** Attempt any SEVEN out of TWELVE: (14)

- (1) What is the purpose of the bias input in a Perceptron, and why is it necessary?
- (2) Compute the total number parameters (Weights and bias) when Layer sizes: 12, 14, 14, 14, 2.
- (3) What is the key idea behind Momentum optimization?
- (4) How does AdaGrad differ from standard gradient descent?
- (5) What makes Adam the most popular optimizer in deep learning?
- (6) Why is padding commonly used in convolutional layers, and what is its general effect on the output dimensions?
- (7) Describe the concept of "weight sharing" in convolutional layers and explain its main benefits.
- (8) What problem did ResNet primarily address, and what architectural innovation did it introduce to solve it?
- (9) Distinguish between Max Pooling and Average Pooling in terms of their operation and typical use cases.
- (10) Distinguish between sequence data and time series data.
- (11) What distinguishes a Gated Recurrent Unit (GRU) from a Long Short-Term Memory (LSTM) network?
- (12) Illustrate vanishing gradient problem in RNN.

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