8

(CO1)

(PO1) 4×3

(CO2)

(PO2)

ISLAMIC UNIVERSITY OF TECHNOLOGY (IUT)

ORGANISATION OF ISLAMIC COOPERATION (OIC)

Department of Computer Science and Engineering (CSE)

MID SEMESTER EXAMINATION DURATION: 1 HOUR 30 MINUTES

SUMMER SEMESTER, 2021-2022 FULL MARKS: 75

CSE 4835: Pattern Recognition

Programmable calculators are not allowed. Do not write anything on the question paper.

Answer all <u>3 (three)</u> questions. Figures in the right margin indicate full marks of questions whereas corresponding CO and PO are written within parentheses.

- 1. a) Define Pattern Recognition (PR). Mention three applications of PR in any domain. Discuss some of the expected invariances of a good feature.
 - b) A common practice in training Deep Neural Networks is to divide the dataset into three splits as- the training, validation and test set. Once the training procedure is completed, it can produce three types of results (in terms of accuracy) as- the training accuracy, validation accuracy, and test accuracy.
 - i. Discuss the merit of drawing conclusion regarding the performance of a network solely based on one of these three types of accuracies.
 - ii. According to the discussion in the previous question, do you recommend to solely depend on any of these three accuracies; or to depend on a pair of them; or to depend on all three of them? Justify your recommendation.
 - iii. Justify the necessity of using three splits instead of using two splits or one split (using the entire dataset for training).
 - c) i. Consider the multi-layer fully connected neural network given in Figure 1. All 2.5×2 activations are sigmoid and the optimizer is stochastic gradient descent. All the weights and biases are initialized to zero and an input $x \in \mathbb{R}^{n \times 1}$ is forward propagated within the network. What is the value of output \hat{y} ?

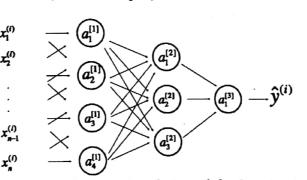
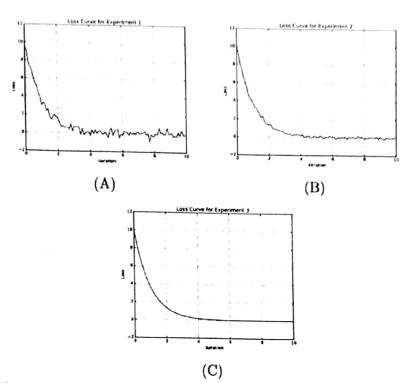


Figure 1: A multi-layer Neural Network for Question 1.c)

ii. Suppose you are solving a classification task to determine whether a leaf image belongs to the 'healthy' or 'diseased' class. A Convolutional Neural Network (CNN) is designed with a single output neuron. Let the output of this neuron be z. The final output of the network \hat{y} is given by: $\hat{y} = \sigma(ReLU(z))$. An Input with a final value $\hat{y} \ge 0.5$ is classified as 'diseased'. Comment on the possible output of this model.





3×3 (CO2) (PO2)

Figure 2: Loss curves for Experiment A, Experiment B, and Experiment C

Figure 2(A) shows a loss curve produced by a 'black box optimizer'. Figures 2(B) and 2(C) show state of the loss curve after tuning a hyperparameter.

- i. Which hyperparameter is likely to be modified here?
- ii. Out of these experiments, which one corresponds to the largest magnitude of the hypermeter? Justify.
- iii. The loss curve for Experiment C seems to be the most desirable. Despite this, is there any reason for which someone would choose the hyperparameter in Experiment B for training a model?
- b) The Code Snippet 1 mentions the basic idea of Adam optimizer.

```
moment1 = 0
moment1 = 0
for t = 1 ... numSteps:
dw = compute\_gradient(w)
moment1 = \beta_1 \times moment1 + (1 - \beta_1) \times dw
moment2 = \beta_2 \times moment2 + (1 - \beta_2) \times dw
w -= learningRate \times \frac{moment1}{\sqrt{moment2} + 10^{-7}}
```

Code snippet 1: For question 2.b)

	i.	Explain the role of the moment1 and moment2 terms in ensuring better convergence.	5 (CO1)
	ii.	Identify the limitation of the Code Snippet 1 and propose an improved solution.	(PO1) 5 (CO2)
c)	Discuss the concept of space warping in learning complex decision boundary.		
			(CO1) (PO1)

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- 3. a) Draw a proper flow-chart by arranging the *keywords* mentioned below according to their roles in solving a classification problem. Discuss their roles and relation with each other properly.
- 5+5 (CO1) (PO1)

Keywords: {Score, Weight vector, Gradient Descent, Loss Function, Input data, Backpropagation, Regularization}

b) Consider a Convolutional Neural Network (CNN) defined by the layers in Table 1. The network takes a RGB image of size 128 × 128 as input and classifies it to one of the 10 available classes. Fill the size of the activation map, number of learnable parameters, and the number of floating-point operations (multiply-add) at each layer.

15 (CO2)

(PO2)

The notation follows the convention:

- CONV-k-N denotes a convolutional layer with N filters, each having size $k \times k$. Padding and stride parameters are always 0 and 1 respectively.
- POOL-k indicates a $k \times k$ pooling layer with stride k and padding 0.
- FC-N stands for a fully-connected layer with N neurons.

Table 1: Network specification for Question 3.b)

Layer	Activation map dimensions	Number of learnable parameters	Number of Multiple-add operation
Input			
CONV-9-32			
POOL-2			
CONV-5-64			
POOL-2			
CONV-3-128			
POOL-2			
FC-256			
FC-64			
Output		j ²	