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# B.M.S. College of Engineering, Bengaluru-560019

Autonomous Institute Affiliated to VTU

## January / February 2025 Semester End Main Examinations

**Programme: B.E.**

**Semester: V**

**Branch: Artificial Intelligence and Machine Learning**

**Duration: 3 hrs.**

**Course Code: 24AM5PCDEL**

**Max Marks: 100**

**Course: Deep Learning**

**Instructions:** 1. Answer any FIVE full questions, choosing one full question from each unit.  
2. Missing data, if any, may be suitably assumed.

UNIT - I			CO	PO	Marks
1	a)	Elaborate on the learning rules of a single layer perceptron.	CO2	PO1	<b>05</b>
	b)	A single-layer perceptron is trained to classify two classes using the Leaky ReLU activation function with $\alpha = 0.01$ . The perceptron receives an input vector $x=[2, -3]^T$ and the weight vector is $w=[0.5, -1.0]^T$ , with a bias $b=0.2$ . Find: i. The net input $z$ to the activation function. ii. The output of the Leaky ReLU activation function for the given input.	CO1	PO2	<b>10</b>
	c)	Differentiate between Online and Batch learning.	CO3	PO2	<b>05</b>
		<b>OR</b>			
2	a)	Differentiate between a Single-layer Perceptron and a Multi-layer Perceptron.	CO2	PO2	<b>05</b>
	b)	A Multi-layer Perceptron is designed with the following architecture to solve the XOR problem. • Input layer with 2 neurons ( $x_1, x_2$ ) • Hidden layer with 2 neurons with sigmoid activation function • Output layer with one neuron operating on sigmoid activation function Given the parameters: Input to Hidden layer weights $\mathbf{W}_{\text{hidden}} = \begin{bmatrix} 10 & 10 \\ -10 & -10 \end{bmatrix}, \quad \mathbf{b}_{\text{hidden}} = \begin{bmatrix} -5 \\ 15 \end{bmatrix}$ Hidden to output layer weights $\mathbf{W}_{\text{output}} = \begin{bmatrix} 10 \\ 10 \end{bmatrix}, \quad \mathbf{b}_{\text{output}} = -15$ Determine the output of the network for the input combination: $(x_1, x_2)=(0,0)$ . Show all intermediate steps.	CO3	PO4	<b>10</b>
	c)	Write the procedural steps of a Perceptron Convergence algorithm.	CO2	PO1	<b>05</b>

**Important Note:** Completing your answers, compulsorily draw diagonal cross lines on the remaining blank pages. Revealing of identification, appeal to evaluator will be treated as malpractice.

<b>UNIT - II</b>																		
3	a)	Given the true probability distribution (P) and the predicted probability distribution (Q) over three outcomes: <table border="1" style="display: inline-table; vertical-align: middle;"> <tr> <th>Outcome (x)</th><th>P(x)</th><th>Q(x)</th></tr> <tr> <td>1</td><td>0.4</td><td>0.3</td></tr> <tr> <td>2</td><td>0.4</td><td>0.5</td></tr> <tr> <td>3</td><td>0.2</td><td>0.2</td></tr> </table>	Outcome (x)	P(x)	Q(x)	1	0.4	0.3	2	0.4	0.5	3	0.2	0.2		CO3	PO3	<b>08</b>
Outcome (x)	P(x)	Q(x)																
1	0.4	0.3																
2	0.4	0.5																
3	0.2	0.2																
		Determine the associated cross entropy.																
	b)	List and explain any 2 activation functions of a Deep Neural Network (DNN).		CO2	PO1	<b>06</b>												
	c)	Illustrate the functioning of a DNN.		CO2	PO3	<b>06</b>												
		<b>OR</b>																
4	a)	A simple Neural Network is designed with one hidden layer functioning on ReLU (Rectified Linear Unit) activation function and one output layer having Weights of the hidden layer $W = [[0.5, -0.5][-0.2, 0.8]]$ with bias $b = [[0.1][0.2]]$ , input vector $x = [[-2][3]]$ and true target label $y = 1$ . i. Compute the activations of the hidden layer using the ReLU activation function. ii. Calculate the output from the network if the output layer uses a linear activation. iii. Calculate the Mean Squared Error (MSE) loss.		CO3	PO4	<b>08</b>												
	b)	Elaborate on the Vanishing gradient problem and ways to mitigate it.		CO2	PO1	<b>06</b>												
	c)	Differentiate between Gradient Descent and Nesterov Accelerated Gradient Descent.		CO2	PO2	<b>06</b>												
		<b>UNIT - III</b>																
5	a)	How does early stopping act as regularizer? Explain.		CO2	PO2	<b>10</b>												
	b)	Elaborate the technique of Adaboost optimization.		CO3	PO1	<b>10</b>												
		<b>OR</b>																
6	a)	Are dropouts always helpful during network training? Explain.		CO2	PO2	<b>10</b>												
	b)	Elaborate on the technique of RMSprop optimization.		CO3	PO1	<b>10</b>												
		<b>UNIT - IV</b>																
7	a)	Design a Convolutional Neural Network (CNN) with the following specification to classify a gray scale image with the following specifications. Represent the size of the intermediate feature maps precisely on the architectural design. <ul style="list-style-type: none"> <li>• Input: (64, 64, 1)</li> <li>• Conv1: 16 filters, kernel: 5x5, activation function: ReLU</li> <li>• Pool1: 2x2 Max Pooling</li> <li>• Conv2: 32 filters, kernel: 3x3, activation function: ReLU</li> <li>• Pool2: 2x2 Max Pooling</li> <li>• Flatten</li> <li>• Dense: 64 neurons, activation function: ReLU</li> <li>• Output: 5 neurons, activation function: Softmax</li> </ul>		CO3	PO4	<b>10</b>												
	b)	What are advantages of padding in a CNN? Explain.		CO2	PO2	<b>05</b>												
	c)	Elaborate on backpropagation through convolutions.		CO1	PO1	<b>05</b>												
		<b>OR</b>																

	8	a)	<p>Consider a Convolutional Neural Network (CNN) with the following configuration:</p> <p>An input image of size 4x4:</p> <table border="1"> <tr><td>1</td><td>2</td><td>0</td><td>3</td></tr> <tr><td>4</td><td>5</td><td>1</td><td>0</td></tr> <tr><td>2</td><td>0</td><td>3</td><td>1</td></tr> <tr><td>1</td><td>2</td><td>4</td><td>0</td></tr> </table> <p>A single 2x2 filter (kernel) with weights:</p> <table border="1"> <tr><td>1</td><td>-1</td></tr> <tr><td>0</td><td>1</td></tr> </table> <ol style="list-style-type: none"> <li>Perform convolution operation on the given input image using the specified filter and stride = 1. Write the resulting feature map.</li> <li>Apply ReLU activation function to the result obtained in (i). Represent the updated feature map in the matrix form.</li> <li>Perform max pooling on the feature map obtained from (ii) using 2*2 pooling window and stride = 1. Determine the resulting feature map.</li> </ol>	1	2	0	3	4	5	1	0	2	0	3	1	1	2	4	0	1	-1	0	1	CO3	PO4	10
1	2	0	3																							
4	5	1	0																							
2	0	3	1																							
1	2	4	0																							
1	-1																									
0	1																									
		b)	Elaborate on types of padding.	CO2	PO2	05																				
		c)	Illustrate the concept of convolution as a matrix multiplication.	CO1	PO3	05																				
<b>UNIT - V</b>																										
	9	a)	<p>Compute the values of input gate and output gate of a LSTM cell by considering the given data:</p> $h_{t-1} = \begin{bmatrix} 0.5 \\ 0.1 \end{bmatrix}, x_t = \begin{bmatrix} 0.8 \\ 0.3 \end{bmatrix}$ $W_i = \begin{bmatrix} 0.2 & 0.4 & 0.6 & 0.1 \\ 0.5 & 0.3 & 0.7 & 0.2 \end{bmatrix}, b_i = \begin{bmatrix} 0.1 \\ 0.2 \end{bmatrix}$ $W_o = \begin{bmatrix} 0.3 & 0.7 & 0.5 & 0.9 \\ 0.4 & 0.2 & 0.6 & 0.8 \end{bmatrix}, b_o = \begin{bmatrix} 0.05 \\ 0.1 \end{bmatrix}$	CO3	PO4	10																				
		b)	Illustrate the architecture of Gated Recurrent Units (GRU) cell along with block diagram and related mathematical expressions.	CO2	PO2	10																				
<b>OR</b>																										
	10	a)	<p>Compute the values of update gate and the Candidate Hidden State <math>\tilde{h}_t</math> of a GRU cell by considering the given data:</p> <p>Current Input (<math>x_t</math>): <math>x_t = \begin{bmatrix} 0.6 \\ 0.2 \end{bmatrix}</math> Previous Hidden State (<math>h_{t-1}</math>): <math>h_{t-1} = \begin{bmatrix} 0.3 \\ 0.5 \end{bmatrix}</math></p> <p>Update Gate Weights and Bias (<math>W_z, b_z</math>):</p> $W_z = \begin{bmatrix} 0.4 & 0.3 & 0.5 & 0.6 \\ 0.7 & 0.2 & 0.4 & 0.3 \end{bmatrix}, b_z = \begin{bmatrix} 0.1 \\ 0.2 \end{bmatrix}$ <p>Candidate Hidden State Weights and Bias (<math>W_h, b_h</math>):</p> $W_h = \begin{bmatrix} 0.6 & 0.2 & 0.8 & 0.4 \\ 0.5 & 0.7 & 0.3 & 0.6 \end{bmatrix}, b_h = \begin{bmatrix} 0.05 \\ 0.1 \end{bmatrix}$	CO3	PO4	10																				

	b)	Illustrate the architecture of Long Short Term Memory (LSTM) cell along with related block diagram and mathematical expressions.	CO2	PO2	<b>10</b>
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