

# Birla Institute of Technology and Science, Pilani

## Work Integrated Learning Programmes Division

M. Tech. in AI & ML

I Semester 2023-2024

Mid-Semester Test  
(EC2 - Makeup)

Course Number	AIMLCZG51
Course Name	DEEP NEURAL NETWORK
Nature of Exam	Closed Book
Weight-age for grading	30
Duration	2 hrs

* Pages	2
* Questions	6

### Instructions

1. All questions are compulsory.
  2. All answers must be directed to the question in short and simple paragraphs or bullet points; use visuals/diagrams wherever necessary.
  3. Assumptions made if any, should be stated clearly at the beginning of your answer.
- 
1. Consider a single layer perceptron having 2 inputs and 1 output. Let threshold be 0.5, learning rate be 0.6, bias be -2 and weight values  $w_1 = 0.3$  and  $w_2 = 0.7$ . Given the input patterns in the table, compute the value of the output and train using perceptron learning rule for one epoch. [5]

E.g. #	$x_1$	$x_2$	$y$
1	1	1	+1
2	1	0	+1
3	0	1	-1
4	0	0	+1

### Rubrics and answer

- z - 0.25, h - 0.25,  $\Delta$  - 0.25, new - 0.25 marks
- Each row in table carries 1 mark each.
- Any wrong equation or wrong computation, reduce appropriately, but maintain a minimum of 0.25 for each row.

$$z = w_1x_1 + w_2x_2 + b$$

$$h = \text{sign}(z)$$

$$\Delta w_1 = \eta(t - h)x_1$$

$$\Delta w_2 = \eta(t - h)x_2$$

$$\Delta b = \eta(t - h)$$

$$w_{\text{new}} \leftarrow w_{\text{old}} + \Delta w$$

$x_1$	$x_2$	$y$	$b$	$w_1$	$w_2$	$z$	$h$	$y, h$	$\Delta$	new $b, w_1, w_2$
1	1	+1	-2	0.3	0.7	0.3 + 0.7 - 2 =-1	-1	N	$\delta b = 0.6(1 - (-1)) = 1.2$ $\delta w_1 = 0.6(1 + 1)1 = 1.2$ $\delta w_2 = 0.6(1 + 1)1 = 1.2$	$b = -2 + 1.2 = 0.8$ $w_1 = 0.3 + 1.2 = 1.5$ $w_2 = 0.7 + 1.2 = 1.9$
1	0	+1	0.8	1.5	1.9	1.5 + 0 + 0.8 =2.3	+1	Y		
0	1	-1	0.8	1.5	1.9	0 + 1.9 + 0.8 = 2.7	+1	N	$\delta b = 0.6(-1 - 1) = -1.2$ $\delta w_1 = 0.6(-1 - 1)0 = 0$ $\delta w_2 = 0.6(-1 - 1)1 = -1.2$	$b = 0.8 - 1.2 = -0.4$ $w_1 = 1.5$ $w_2 = 1.9 - 1.2 = 0.7$
0	0	+1	-0.4	1.5	0.7	0+0- 0.4 = -0.4	-1	N	$\delta b = 0.6(1 + 1) = 1.2$ $\delta w_1 = 0.6(1 + 1)0 = 0$ $\delta w_2 = 0.6(1 + 1)0 = 0$	$b = -0.4 + 1.2 = 0.8$ $w_1 = 1.5$ $w_2 = 0.7$

2. Derive the equation for the derivative of categorical cross-entropy loss  $L$  with respect to the weighted sum  $Z$ , for a three-class classification problem. Assume single hidden layer and  $d$  input neurons. [5]

#### Rubrics and answer

for class k, output  $Z_k = \sum_{i=1}^3 w_{jk} \cdot a_i$

softmax function  $\hat{y}_k = \frac{e^{Z_k}}{\sum_{j=1}^3 e^{Z_j}}$  0.5 mark

categorical cross-entropy loss  $L = -\sum_{k=1}^3 y_k \cdot \log(\hat{y}_k)$  1 mark

$$\frac{\partial L}{\partial Z_k} = \frac{\partial L}{\partial \hat{y}_k} \cdot \frac{\partial \hat{y}_k}{\partial Z_k} \quad 1 \text{ mark}$$

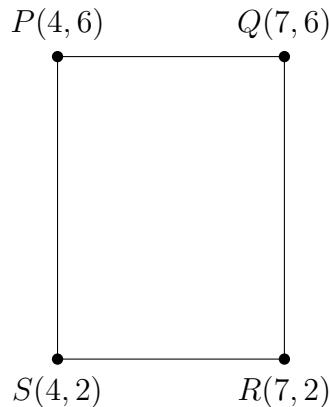
$$\frac{\partial L}{\partial \hat{y}_k} = -\frac{y_k}{\hat{y}_k} \quad 1 \text{ mark}$$

$$\frac{\partial \hat{y}_k}{\partial Z_k} = \hat{y}_k \cdot (1 - \hat{y}_k) \quad 1 \text{ mark}$$

$$\frac{\partial L}{\partial Z_k} = -y_k \cdot (1 - \hat{y}_k) \quad 0.5 \text{ mark}$$

3. Construct an MLP for the given complex decision boundary.

[6]



**Rubrics and answer**

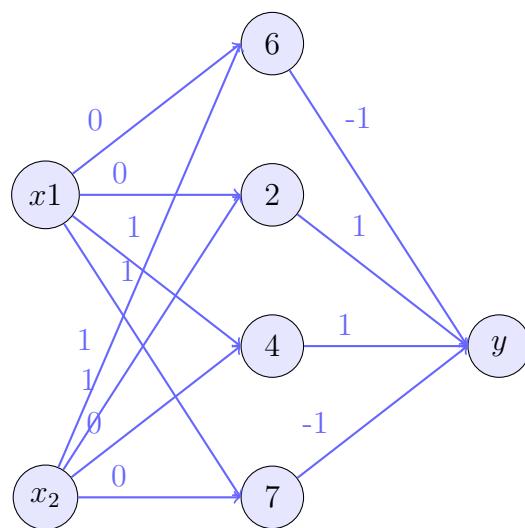
- Each equation, 1 mark each.
- MLP, hidden layer with correct parameters - 1 mark
- MLP, last layer with correct parameters - 1 mark

$$PQ \rightarrow 0x_1 + 1x_2 = 6$$

$$SR \rightarrow 0x_1 + 1x_2 = 2$$

$$PS \rightarrow 1x_1 + 0x_2 = 4$$

$$QR \rightarrow 1x_1 + 0x_2 = 7$$



4. Find the minimum value of  $p$  for the equation  $t = (2p + 3)^2$  using SGD. Assume the initial value of  $p$  as 6 and learning rate as 0.1. Do 3 iterations.

[4]

**Rubrics and answer**

- Each equation, 1 mark each.

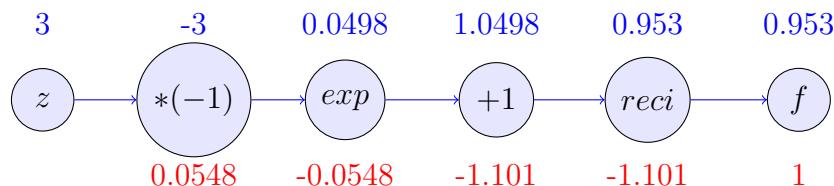
$$\begin{aligned}
dt/dp &= 2(2p + 3)2 = 8p + 12 \\
p_1 &\leftarrow 6 - 0.1(8 * 6 + 12) = 0 \\
p_2 &\leftarrow 0 - 0.1(8 * 0 + 12) = -1.2 \\
p_3 &\leftarrow (-1.2) - 0.1(8 * (-1.2) + 12) = -1.44
\end{aligned}$$

5. Draw the computational graph for the equation  $f = 1/(1 + e^{(-z)})$ . Show the computations of derivatives of  $f$  wrt  $z$  in the graph. Using the graph, compute the value of  $f$  and the derivatives if  $z = 3$ . [5]

Rubrics and answer

- FP graph 1 mark
- BP graph 1 mark
- Output computation 1 mark
- Gradient computation 1 mark

$$\begin{array}{lll}
p = -z = -3 & \frac{\partial p}{\partial z} = -1 & \frac{\partial f}{\partial p} = (-1)(-0.0548) = 0.0548 \\
q = e^p = e^{-3} = 0.0498 & \frac{\partial q}{\partial p} = e^p & \frac{\partial f}{\partial q} = e^{-3}(-1.101) = -0.0548 \\
r = p + 1 = 0.0498 + 1 = 1.0498 & \frac{\partial r}{\partial p} = 1 & \frac{\partial f}{\partial r} = 1(-1.101) = -1.101 \\
f = \frac{1}{r} = \frac{1}{1.0498} = 0.953 & \frac{\partial f}{\partial r} = \frac{-1}{r^2} & \frac{\partial f}{\partial r} = \frac{-1}{0.953^2} = -1.101
\end{array}$$



6. Given an error surface, compute the value that minimizes the error with respect to  $(w_1, w_2, w_3)$ . Compute the minimum possible value of error. [5]

$$E(w_1, w_2, w_3) = (w_1 - w_2)^3 - 2(w_1^2 - w_2) + w_1^2 + w_2^2$$

Rubrics and answer

$$\begin{aligned}
\frac{\partial E}{\partial w_1} &= 3(w_1 - w_2)^2 - 2w_1 \\
\frac{\partial E}{\partial w_2} &= -3(w_1 - w_2)^2 + 2w_2 + 2 \\
\frac{\partial E}{\partial w_3} &= 0
\end{aligned}$$

- Each equation 1 mark
- if the student writes or attempts that the equations have to be equated to zero and compute the value of parameters - 2 marks