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Code No.: 14267 N/O

VASAVI COLLEGE OF ENGINEERING (AUTONOMOUS), HYDERABAD

Accredited by NAAC with A++ Grade

B.E. IV-Semester Main & Backlog Examinations, July-2023

Machine Learning

Time: 3 hours

(Common to CSE & AIML)

Max. Marks: 60

Note: Answer all questions from Part-A and any FIVE from Part-B

Part-A $(10 \times 2 = 20 \text{ Marks})$

Q. No.				f the que	$\times 2 = 20 M$				-	
1.	Consider se	et of hypoth	eses H1 a	nd III	1		M	L	CC	P
	more genera	al than the	other hype	III HZ, FI	nd among	them which is	2	1	1	1,
	H1 = {'?',']	Vormal' '2	other hype	duesis.						.,
	Н2 — ССТУ	worman , ?	, '+'}							
	$H2 = {\text{`War}}$									
2.	Find the H	ypothesis v	which bes	t fits the	given set	of samples using				
					g Bot	or samples using	g 2	3	1	1,2
	Day Outl		perature	Humidi	y Wind	PlayTennis				
	D2 Sun		Hot Hot	High	Weak	No			7	
	D3 Over	cast	Hot	High High	Strong					
- 1	D4 Rai		Aild	High	Weak Weak	Yes				
3.	Calculate the		SECTION OF		weak	Yes bias. The input				
1	activation fur	action.	,,	ooj. Ose	omary Sig	n bias. The input veight values are moid function as		1	the state of	
	Consider the		et of train	ing exam	oles:		2	1	2	1,2
		Instance	Classifi	cation	a ₁ a ₂				-	1,2
		1 2	+		ГТ		100			
		3	+		T T					
		5	+		F					
		6	_		T					
re	a) What is the espect to the t	ne entropy	of this co	-11		examples with				
				roution.						
	et there be 5 l	-					2	3	3	124
		$P(h_i \mid D)$	$P(F \mid h_i)$	$P(L \mid h_i)$	$P(R \mid h_i)$			_	5	,2,4
		0.4	1	0	0					
		0.2	0	1	0				,	
		0.1	0	0	• 1					
				The state of the s						
		0.1	0 .	1	0			ν'		
		0.1	0	1	0	ward (F). What		· ·		

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6.					
	Define Vapnik Chervoniks (VC) dimension. How VC dimension is related with no of training examples used for learning.	2	2	3	1,2
7.	Write the steps to find a Maximum A Posterior (MAP) hypothesis using Brute-force method.	2	2	4	1,2
8.	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	2	3	4	1,2,4
	Find the Q(s1,aright)?				
9.	Consider a string before a crossover s1 `= 111 <u>1010</u> 101 s2 `= 1110110101 Compute the offspring's after the crossover	2	3	5	1,2,3
10.	Find the out _{fut} dimension of Conv layer 1 if the input image is 228*228 and kenel size is 5*5 with zero padding and stride is 2.	2	3	5	1,2,4
1	Part- B (5×8 = 40 $Marks$)				
11 0)	Explain the issues trace and the instance of t				
11. a)	Explain the issues that need to be considered while designing the machine learning algorithm.	4	1	1	1,2
11. a) b)	Consider the training data in the following table where Play is a classattribute. In the table, the Humidity attribute has values "L" (for low) or "H" (for high), Sunny has values "Y" (for weak), and Play has values "Yes" or "No".	4	3	1	1,2
5 07	Consider the training data in the following table where Play is a classattribute. In the table, the Humidity attribute has values "L" (for low) or "H" (for high), Sunny has values "Y" (for yes) or "N" (for no), Wind hasvalues "S' (for strong) or "W" (for weak), and Play has values "Yes" or "No".				
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	Major		The second secon						
	CS	Experience	Tie	Hired'	?				
	CS	programming	pretty	NO					
	CS	programming	pretty	NO					
	CS	management	pretty	YES	E TETT	1.5			
		management	ugly	YES	D THE VI				
	business	programming	pretty	YES					
	business	programming	ugly	YES					
	business	management	pretty	NO					
	business	management	pretty	NO					
3. a) II b)	lustrate Expect	ation Maximizati	-			4	2	3	1
		Bayes r	R: The le	cture started by cturer arrives late cture concerns n cturer is Manuel nny	e	4	3	3	1,2
	P(s)=0.3	8	M [F	P(M)=0.6				/	<i>i</i> .
P	(L M^S)=0.05 (L M^~S)=0.1 (L ~M^S)=0.1 (L ~M^S)=0.2	P(T P(T	L)=0.3 -L)=0.8	P(R P(R	M)=0.3 ~M)=0.6				<i>k</i> /
Fin Dis	(L M^-S)=0.1 (L ~M^S)=0.1 (L ~M^-S)=0.2	P(T)	-L)=0.8 - ^ T)?	R P(R	M)=0.3 ~M)=0.6	4	2	4	1,2
Fin	(L M^-S)=0.1 (L ~M^S)=0.1 (L ~M^-S)=0.2	P(T) P(T) P(S^~M^L~R) cance of locally	^T)? weighted regre	ession.	M)=0.3 ~M)=0.6	4 4	2	4 4	1,2
Fin Dis	(L M^-S)=0.1 (L ~M^S)=0.1 (L ~M^-S)=0.2 In the value of I seems the significant of I in the control of I	P(S^~M^L~R	^T)? weighted regre	ession. Weight	M)=0.3 ~M)=0.6	4 4	2 3	4 4	
Fin Dis	(L M^-S)=0.1 (L ~M^S)=0.1 (L ~M^S)=0.2 In the value of I in the cuss the significant in the cus of the cus of the cus of the customer in the cust	P(S^~M^L~R cance of locally theight	^T)? weighted regree Age 45	ession. Weight	M)=0.3 ~M)=0.6	4 4	2 3	4 4	
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Fin Dis	(L M^-S)=0.1 (L -M^S)=0.1 (L -M^S)=0.2 Index the value of I ID I I I I I I I I I I I	P(S^~M^L~R cance of locally Height 5 5.11 5.6 5.9 4.8	Age 45 26 30 34 40	ession. Weight 77 47 55 59 72	M)=0.3 ~M)=0.6	4 4	2 3	4 4	
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Fin Dis	(L M^-S)=0.1 (L -M^S)=0.1 (L -M^S)=0.2 and the value of I secuss the significant of I 2 3 4 5 6 7 8	P(S^~M^L~R cance of locally Height 5 5.11 5.6 5.9 4.8 5.8 5.3 5.8 5.5	Age 45 26 30 34 40 36 19 28 23	ession. Weight 77 47 55 59 72 60 40 60 45	M)=0.3 ~M)=0.6	4 4	2 3	4 4	
Fin Dis	(L M^-S)=0.1 (L -M^S)=0.1 (L -M^S)=0.2 Id the value of I scuss the signifi ID 1 2 3 4 5 6 7 8 9	P(S^~M^L~R cance of locally v Height 5 5.11 5.6 5.9 4.8 5.8 5.3 5.8	Age 45 26 30 34 40 36 19 28	ession. Weight 77 47 55 59 72 60 40 60	M)=0.3 ~M)=0.6	4 4	2 3	4 4	

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b) What is Deep learning? Explain the Recurrent Neural Network (RNN)	4	2	5	1,2
architecture and training steps myorved				
Model.	4	1	1	1,2
What is an Inductive Bias? Explain the List then Eliminate Algorithm.	4	3	2	1,2,3
b) What is Artificial Neural Network? Calculate the output where the threshold t=0.0 and take four combinations of inputs for x and y where bias=-1.	7	3		
10,				
01,				
00,				
1 1.				
-1				
W = 0.3				
x w=0.5 t=0.0				
$\mathbf{W} = 0.5$				
W === -0.4				
	-			
y				
17. Answer any <i>two</i> of the following:	4	1	3	1,2
a) What is E -exhaustive? What are the true error and training error in	4	1	,	1,2
version space? Give with an example.		2	4	1,2
The design of the following Instance-based learning techniques.	4	2	4	1,2
. Case based Reasoning II. Radial basis networks				
CODY I was below with filter values. Compute the output	t 4	3	5	1,2,3
c) Consider the CNN image below with meet of convolution 1 with zero padding and stride 1.And mention the	2			
outcome after applying ReLu Function.				
Fitter				
Input Image				
3 8				
8 4 🚫 1 0 -1				
1 0 -1				
5 4 4 5 4				

M: Marks; L: Bloom's Taxonomy Level; CO; Course Outcome; PO: Programme Outcome

D. D.	Toyol 1	20%
i)	Blooms Taxonomy Level – 1	30%
ii)	Blooms Taxonomy Level – 2	50%
iii)	Blooms Taxonomy Level – 3 & 4	3070