Code No.: 16147 (G) N/O

VASAVI COLLEGE OF ENGINEERING (AUTONOMOUS), HYDERABAD

Accredited by NAAC with A++ Grade

B.E. VI-Semester Main & Backlog Examinations, May/June-2023

Fundamentals of Machine Learning (OE-IV)

(Common to Civil & Mech.) Time: 3 hours

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.	Stem of the question		M	L	СО	D.
1.	List any four problem statements in AI.	+				PO
2.	Justify the importance of Test error (or Generalization error) metric in evaluating ML models?	n	2	1 2	1	1
3.	Illustrate the significance of ML Performance metric.		4	2	1	1
4.	List different Python data types along with an example.	1	2	2	2	1
5.	Define a neuron. List few functions that	2	2	1	2	1
,			!	1	3	1
6.	The following neural network takes two binary valued inputs X_1 and X_2 , and the activation function is the threshold function $(h(x) = 1 \text{ if } x > 0; 0 \text{ otherwise})$. Which of the following logical functions does it compute?	2		2	3	1
	-4					
	(x ₁) +8	106				
	OR b) AND c) NAND d) none					
1						
	What are Support Vectors in SVM?	2	1	4	ı	1
P	X and Y are Boolean random variables. Given: $P(X=1) = 0.3$; $P(X=0) = 0.7$; $P(Y=1 \mid X=1) = 0.4$; $P(Y=0 \mid X=1) = 0.6$; $P(Y=1 \mid X=0) = 0.6$; $P(Y=0 \mid X=0) = 0.4$; Calculate $P(X=1 \mid Y=0)$ using Bayes rule.	2	2	4		1
D	Discuss briefly about Density based clustering?	2	1	-		
W	hat does k mean in k-means clustering?		1	5		
	Part-B $(5 \times 8 = 40 \text{ Marks})$	~	1	3	1	
a) Co	ompare and contrast Supervised and Unsupervised Learning.					

b)	What are the addifferent ways to	dvanta o creat	ges of using te a Numpy a	g Numpy ov array from a	ver Python list?	lists? What are	the 4	1	1		1
2. a)	Discuss how Li Linear Regression	inear on.	Regression	works. Write	e a python	code to implem	nent 4	3	2		2
b)	Compare and co	ontrast	Linear Regi	ression with	Logistic Reg	gression.	4	3	2		2
3. a)	Explain the deta	ails of	Perceptron a	algorithm for	Boolean Al	ND function.	4	3	3	}	1
b)	A the new	Trone V	ico a ciamoio	l activation f	function. Per		pass $O_2 = 0$	3	3	3	2
	4	D		33	- 1 50-	01					
	72	- / (2)*		*(1)	(6)	0,					
	W	13 = C	0.10 was	3 20.30	W35=0.6	o was = 0.70					
	W	14 = 0	0.80	ц = O. ЦО	No.4-B	60 W46 - 0.80					
		446									
			$x_1 = 0.50$ $36 = 0.75$	b ₁ = b ₂	0.35 20.60						
[4. a)	I Lain a Mairia D	follow	ing dataset f	edict if tennis	res considero	ed for playing to ted on a particula y: high, wind: v	u day	4 3	3	4	2
14. a)	Using Naïve B which is { outl	follow Bayes a look :	ing dataset falgorithm prosumny, Temp	For the feature edict if tennis	res considere s can be play ld , Humidity	y: high, wind: v	u day	4 3	3	4	2
(4. a)	Using Naïve B which is { outl	follow Bayes a look :	ing dataset falgorithm prosumny, Temperature	For the feature dict if tennisperature : mil	res considero	ed on a particula	u day	4 3	3	4	2
(4. a)	Using Naïve B which is { outloo	follow Bayes a look :	ing dataset falgorithm prosumny, Temperature Hot	For the featuredict if tennisperature: mil	res consideres can be play Id , Humidity Wind Weak	y: high, wind: v	u day	4 3	3	4	2
(4. a)	Using Naïve B which is { outloo	follow Bayes a look :	ing dataset falgorithm prosumny, Temperature Hot Hot	For the featuredict if tennisperature: mil	res consideres can be play	y : high, wind : v	u day	4 3	3	4	2
[4. a)	Using Naïve B which is { outloo Sunn Overca	follow Bayes a look :	ing dataset falgorithm prosumny, Temperature Hot Hot Hot	For the featuredict if tennisperature: mil	res consideres can be play Id , Humidity Wind Weak Strong	y : high, wind : v	u day	4 3	3	4	2
(4. a)	Using Naïve B which is { outloo Sunn Overce Rain	follow Bayes a look :	ing dataset falgorithm prosunny, Temperature Hot Hot Hot Mild	For the featuredict if tennisperature: mil	res consideres can be play Id , Humidity Wind Weak Strong Weak	y : high, wind : v	u day	4 3	3	4	2
[4. a)	Using Naïve B which is { outloo Sunn Overca Rain	follow Bayes a look:	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Mild Cool	For the feature dict if tennis perature : mil	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak	y : high, wind : v Play tennis No No Yes Yes	u day	4 3	3	4	2
(4. a)	Using Naïve B which is { outloo Sunn Overca Rain Rain	follow Bayes a look :	ing dataset falgorithm prosunny, Temperature Hot Hot Hot Mild	For the feature dict if tennis perature: mil	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak Weak	y : high, wind : v Play tennis No No Yes Yes Yes No Yes	u day	4 3	3	4	2
(4. a)	Using Naïve B which is { outloo Sunn Overca Rain Rain overca	follow Bayes a look:	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Mild Cool Cool	For the feature edict if tennis perature : mil	res consideres can be play ld , Humidity Wind Weak Strong Weak Weak Weak Strong	y: high, wind: v Play tennis No No Yes Yes Yes No Yes No	u day	4 3	3	4	2
(4. a)	Using Naïve B which is { outlook Sunn Overca Rain Rain Overca Sunn	follow Bayes a look:	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Mild Cool Cool	For the feature edict if tennis perature : mil Humidity High High High High Normal Normal	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak Strong Strong Strong Weak Weak	y: high, wind: v Play tennis No No Yes Yes Yes No Yes No Yes No Yes	u day	4 3	3	4	2
(4. a)	Using Naïve B which is { outloo Sunn Overca Rain Rain overca	follow Bayes a look:	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Cool Cool Mild	For the feature edict if tennis perature : mil Humidity High High High Normal Normal Normal High	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak Weak Strong Strong Weak Weak Weak Weak	y: high, wind: v Play tennis No No Yes Yes Yes No Yes No Yes No Yes Yes No Yes No Yes	u day	4 3	}	4	2
(4. a)	Using Naïve B which is { outlook Sunn Overca Rain Rain Overca Sunn Overca Sunn	follow Bayes a look: Ny yay aast na na sast	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Cool Cool Cool Mild Cool	For the feature edict if tennis perature : mile the mile that the mile t	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak Strong Strong Weak Weak Weak Weak Strong	Play tennis No No Yes Yes No Yes No Yes	u day	4 3	}	4	2
(4. a)	Using Naïve B which is { outlook Sunn Sunn Overca Rain Rain Overca Sunn Sunn Rain	follow Bayes a look: Ny hy hy hast he hast hy	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Cool Cool Cool Mild Cool Mild Cool Mild Mild	For the feature edict if tennis perature : mile the mile that the mile t	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak Strong Strong Weak Weak Strong Strong Strong Strong Strong	Play tennis No No Yes Yes No Yes No Yes	u day	4 3	3	4	2
[4. a)	Using Naïve B which is { outlook Sunn Sunn Overca Rain Rain Overca Sunn Sunn Overca Sunn	follow Bayes a look: Nok IV IV asst n n asst n n n n n n n sast n n n sast	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Cool Cool Cool Mild Cool Mild Mild Mild Mild Mild Hot	For the feature edict if tennis perature : mile tennis perature : mi	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak Strong Strong Weak Weak Weak Weak Weak Weak Weak Weak	Play tennis No No Yes Yes Yes No Yes	u day	4 3	3	4	2
14. a)	Using Naïve B which is { outlook Sunn Sunn Overca Rain Rain Overca Sunn Sunn Overca Sunn Overca Sunn Overca	follow Bayes a look: Ny Ny Sast Sast Sast Sast Sast Sast Sast Sast	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Cool Cool Cool Mild Cool Mild Mild Mild Mild Mild	For the feature edict if tennis perature : mile the mile that the mile t	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak Strong Strong Weak Weak Strong Strong Strong Strong Strong	Play tennis No No Yes Yes No Yes No Yes	u day	4 3	3	4	2
14. a)	Using Naïve B which is { outlook Sunn Sunn Overca Rain Rain Overca Sunn Sunn Overca Sunn Coverca Rain Rain Rain Rain Rain Rain Rain Rai	follow Bayes a look: Dok Ty Ty Ty Ty Ty Ty Ty Ty Ty T	ing dataset falgorithm prosumny, Temperature Hot Hot Hot Cool Cool Cool Mild Cool Mild Mild Mild Mild Mild Mild Mild Mil	For the feature dict if tennis perature: mile perat	res consideres can be play Id , Humidity Wind Weak Strong Weak Weak Strong Strong Strong Weak Weak Strong Weak Strong	Play tennis No No Yes Yes Yes No Yes	veak}	4 3	2	4	2

	on each of 12 in	Subject	A						
		1	1.0	В					
		2		1.0					
		3	3.0	2.0					
		4		4.0					
		5	5.0	7.0					
		6	3.5	5.0					
		7	4.5	5.0					
		8	4.8	6.0					
	-	9	5.2	7.0					
			5.5	.6.0					
		10	6.0	7.0					
	-	11	6.4	8.0					
		12	6.8	7.0 -					
16. a)	Discuss the star								
16. a)	Explain the following Learning: i) Precision iv) F1 Score	ing performance	e metrics used	g Problem. I for Classific Specificity	ation in Machine	4 4	2 2	1 2	1
	Explain the following Learning: i) Precision	ing performance	e metrics used	l for Classific	ation in Machine	4 4			
b)	Explain the following Learning: i) Precision iv) F1 Score	on ii) Retthe following:	e metrics used	for Classific		4			
b) (. a) b)	Explain the following Learning: i) Precision iv) F1 Score Answer any two of the Illustrate how Back appropriate example. Compare and Contra	on ii) Retailed the following: Propagation and the control of the	e metrics used ecall iii) lgorithm world and Clusterian	for Classific Specificity ks in a Neura		4	2	3	2
b) a) b)	Explain the following Learning: i) Precision iv) F1 Score Answer any two of the Illustrate how Back appropriate example.	on ii) Retailed the following: Propagation and the control of the	e metrics used ecall iii) lgorithm world and Clusterian	for Classific Specificity ks in a Neura		4	2	2	1

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

(i)	Blooms Town	TO. Flogram
ii)	Blooms Taxonomy Level – 1 Blooms Taxonomy Level – 2	20%
iii)	Blooms Taxonomy Level – 2	40%
	Tuxonomy Level – 3 & 4	40%
