MID-TERM EXAMINATION (Course Name: B.Tech CSE-AI/ ECE-AI) (Semester 5) (OCTOBER, 2023) OFF LINE mode, <Subjective>

Subject Cod			Subject: Machine Lear	ning(ML)				
Time: 1 ½H			М	aximum M	arks:30			
vote: Q. 1 is	compulsory.							
Q1					12 5*41			
	(a) Explain differe	nce between a hard m	argia and a soft mar	gin in Sun	(2.5*4)			
	Vector Machine (S	SVM).	largin and a sort mar	gin in Sup	port			
	(6)Suppose that we	have already trained	a classification tree,	K-Nearest				
-	Neighbors, and logi	istic regression. When	the data is relatively	large (e.g	, larger			
	than one million po	ints), which one of the	e techniques is slowe	er in makir	ng a			
	prediction?. Explain	prediction?. Explain with reason.						
	(ć) Explain the role	of data splitting in the	context of data prep	processing	. Why is			
	it essential to divide	e data into training an	d testing sets before	model tra	ining?			
w.	(d) What role does	standardization play in	n data preprocessing	and how	does it			
	impact the training	of machine learning n	nodels?	, and now	doesit			
	The state of the s	or machine rearring n	TOUCIS:					
2 (Attern	npt any Two Parts)	UNIT-1			(5,5)			
	nat are different cause		achine Learning? Wh	at are	10,0)			
fallou	its?							
(b) Eni	umerate and elucidate	e various techniques e	employed in data pre					
		significance of each in	the context of mach	nine				
	sing, highlighting the	significance of each ir	the context of mach	nine	dias I			
proces	sing, highlighting the	significance of each in	the context of mach	nine				
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(c) Des Machin learnin	ssing, highlighting the hig. scribe the role of Explore Learning and explaing.	oratory Data Analysis (in why it is considered UNIT-2	(EDA) In the context of a crucial step in the	of machine	(5,5)			
(c) Des Machin learnin (Attem	ssing, highlighting the right of Explorer Learning and explang. Ipt any Two Parts) ose you are building	oratory Data Analysis (in why it is considered UNIT-2 a logistic regression i	(EDA) in the context of a crucial step in the	of machine whether	(5,5) .			
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	G		No	0.20			
	Н		Yes	0.91			
* N	1		No	0.74			
	J		Yes	0.34			
diabetes. not have Calculate model?	that you cho han 0.4, you w If it is less tha diabetes. Accuracy, Pre	n or equal to	e that the cor 0.4, then you , Specificity ar	responding po would conclu and F1 Score ba	atient has ude that they ased on the		
follows. Th tiredness (Table-1	neir symptom:	s are cough (C), fever (F), c	difficulty brea	thing (B), an	d	
	Cough (C)	Fever (F)	Difficulty breathing (8)	Tiredness (T)	Diagnosis		
Patient 1	по	yes -	yes -	yes	Sick	1 L	1
Patient 2	yes ·	yes	no	yes	Sick		1
atient 3	yes	no	yes .	yes	Sick		77.0
Patient 4	yes	yes .	yes .	no	Sick •		
Patient 5	yes · · ·	no	no	yes	Healthy		
Patient 6	no	yes	yes · *	no 7	Healthy		
Patient 7	no	yes	no	no	Healthy	1	-
Patient 8	no	no	no	yes	Healthy	013/4	
Diagnosis) t	ni impurity ir this data. Wh the simplifie	at is the acci	uracy of this				
x		Υ		No.		-	
		5		The State of the	N. T. C.	DE T	1.
2		11	774	1			
3		14	•				
		18		2.40			
4							
ing least so	quare metho X independe	d to find the	eguation of dependent v	the regressi	on model fit culate R squ	ted are	

End-Term Examination (CBCS)(SUBJECTIVE TYPE)(OffLine) Course Name: < B.Tech CSE AI/ ECE-AI >, Semester: <5th > (November-December, 2023)

Subject Code: BAI 301	Subject: Machine Learning
Time: 3 Hours	Maximum Marks : 60
Note: Q. 1 is compulsory. Attempt one question ea	ch from the Units I, II, III & IV.

	: Q. 1 is compulsory. Attempt one question each from the Units I, II, III & IV.					
Q1.		(2.5*8=20				
	(a) With reference to logistic regression, if you will travel along with ROC curve cutoff value will change? Explain?					
	(b) A classifier is designed to identify if patients of a clinic need to go through the the diagnostic steps after the first round of testing. What classification metr more or less appropriate? Why?	ic would be				
	(c) Is it a good idea to focus on hyperparameter optimization when we can also the quality or quantity of the training data?					
	 (d) A decision tree classifier learned from a fixed training set achieves 100% ac the test set. Which of the algorithms (Logistic regression, An SVM with a pokernel, k-Nearest neighbors, Naïve Bayesclassifier) trained using the same to is guaranteed to give a model with 100% accuracy? Explain. (e) What is the curse of dimensionality? 	lynomial				
	(f) If a Decision Tree is underfitting the training set, is it a good idea to try scal input features? Explain?	ing the				
	(g) A correlation between age and health of a person found to be 0.99 . On the basis of this you would tell the doctors that age causes health? Explain with reason?					
	(h) Suppose you are using Ridge Regression and you notice that the training er validation error are almost equal and fairly high. Would you say that the mofrom high bias or high variance? Should you increase the regularization hyperparameter α or reduce it? Explain?					
	UNIT-1					
Q2.	 (a) Discuss the potential consequences of not handling outliers during data preprocessing (b) How can feature engineering contribute to improving the performance of machine learning models, and what are some common techniques used in feature engineering? (c) How can you use quantiles of a distribution to detect its outliers? 	(3+3+4=10)				
Q3.	(a)What are different steps in data pre-processing required before any model building? Why these steps are required? (b) Explain the difference between the one-hot and label encoding methods? (c) Discuss the trade-offs between removing outliers and transforming them during the data preprocessing stage	(3+3+4=10)				
Ъ	UNIT-2					
Q4.	A dataset of patients who have tested positive or negative for COVID-19 follows. Their symptoms are cough (C), fever (F), difficulty breathing (B), and tiredness (T).	(10				

		Cough (C)	Fever (F)	Difficulty	Tiredness	Diagnosis	
			The second secon	breathing	(T)	1	14
				(B)			147
	Patient 1	no	yes	yes	yes	Sick	
	Patient 2	yes	yes	no	yes	Sick	
	Patient 3	yes	no	yes	yes	Sick	
	Patient 4	yes	yes	yes	no	Sick	
	Patient 5	yes	no	no	yes	Healthy	
	Patient 6	no	yes	yes	no	Healthy	
	Patient 7	no	yes	no	no	Healthy	
	Patient 8	no	no	no	yes	Healthy	
	Build a naive	Bayes mode	el that predic	ts the diagno	osis from the	symptoms. Use	
	the naive Ba	yes algorithr	n to find the	following pr	obabilities:		
	Note: For th	a fallowing s	westions the	o symptoms	that are NO	mentioned are	_ =
	completely	unknown to	uestions, the	ple, if we kno	ow that the r	natient has a	
				fever, it doe			
	doesn't have		d about then	rever, it doe	.s not mean	ine patrom	
	uocsii t iiave	a level.					7
	a. The proba	bility that a	patient is sic	k given that t	the patient h	as a cough	1
	b. The proba	bility that a	patient is sic	k given that	the patient is	not tired	7
	~!				ba natiant b	os a sough and a	
		ibility that a	patient is sic	k given that i	ine patient n	as a cough and a	
	fever						
	d The weeks	- h:1:4:4h-a-a-a	mationt la pio	li aluan that	the nations b	as a sough and a	
		o difficulty b		k given that	the patient h	as a cough and a	
Q5.	(a)If it takes	one hour to	train a Decis	ion Tree on a	training set	containing 1	(4+6
	million insta	inces, roughl	y how much	time will it to	ake to train a	nother Decision	=10)
	Tree on a tra	aining set co	ntaining 10 n	nillion instan	ces?		
							1
		a dataset of	labeled poir	nts in a 2D sp	ace:		
l	Dataset:						
		3), Label: Cla					
		B), Label: Cla					
		2), Label: Cla					
		B), Label: Cla					1
		3), Label: Cla					
		4), Label: Cla					
		6), Label: Cla					
		7), Label: Cla					
				eled point: (3			
	5 5 5		ghbors (KNN) algorithm to	o classify the	new point. Set	
	k=3 for this	•					
						ne new point and	
			_		nave regardi	ng the choice of k	
1	and the imp	act of differ	ent distance	metrics.			

	UNIT-3	(10)
Q6.		(10)
•	Given the following set of points:	
	1 (2 3) (5 8) (1 3) (8 8) (7 3) (6.4) (1.6) (4.7)	
- 1	Compute full iterations of k-means clustering for K=2 with initial clusters	
	$ C_{-} + C_{$	
	Make sure to write down the necessary distances (Euclidean distance), explain	
	the steps you follow, and to describe the resulting clusters (centroid and	
	points) at the end of each iterations.	
	points) at the end of cachinerations	1.0
Q7.	(a) What are the main motivations for reducing a dataset's dimensionality?	(6+4
ų٬.	What are the main drawbacks?	=10)
	(b) Can PCA be used to reduce the dimensionality of a highly nonlinear	
		1
	dataset? Explain?	
	UNIT-4	¥.
Q8.	(a) A boosted strong learner L is formed by three weak learners, L1, L2, and L3	(4+2+4=1
що.	Their weights are 1.0.4, and 1.2, respectively. For a particular point, L1 and	0)
	L2 predict that its label is positive, and L3 predicts that it's negative. What is	
	the final prediction the learner L makes on this point?	
	the final prediction the rearrier 2 makes an analy	
	(b) What is the difference between hard and soft voting classifiers?	1
	(c) If your AdaBoost ensemble underfits the training data, which	
	hyperparameters should you tweak and how?	
Q9.	(a)If you have trained five different models on the exact same training data,	(5+5=10)
	and they all achieve 95% precision, is there any chance that you can combine	
	these models to get better results? If so, how? If not, why?	
	these models to get better results. It say no to the say in	
	(b) Describe Random Forest in the context of ensemble learning. How does	
	Random Forest utilize the concept of ensemble learning, and what	- 4
	distinguishes it from a single Decision Tree? Discuss the benefits of employing	
	distinguishes it from a single Decision free: Discuss the Benefits of the Control	
	an ensemble of trees in Random Forest?	
		40.00