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## SCHOOL OF COMPUTER SCIENCE AND ENGINEERING CONTINUOUS ASSESSMENT TEST - II

WINTER SEMESTER 2024-2025

SLOT: C2 + TC2

**Programme Name & Branch** 

: M.Tech (Integrated) & Computer Science and Engineering

Course Code and Course Name : MDI3003 and Advanced Predictive Analytics

Faculty Name(s)

: Dr. G. N. Balaji, Dr. Archana T, Dr. UMA PRIYA D

Class Number(s)

: VL2024250502376, 2378, 2380

**Date of Examination** 

: 18.03.2025

**Exam Duration** 

: 90 minutes

Maximum Marks: 50

## **Answer All Questions**

- M Max mark; CO Course Outcome; BL Blooms Taxonomy Level (1 Remember, 2 -Understand, 3 - Apply, 4 - Analyse, 5 - Evaluate, 6 - Create)
- CO3: Gain the insights from data through Exploratory data analysis for feature engineering
- CO4: Compare the underlying predictive modeling techniques. Analyse the performance of the model and quality of results.
- CO5: Explore predictive models to enhance the prediction performance.

Q. No			Question			M	CO	BL
1./	Given data = {2.5 1.6, 0.9}. apply Property coordinates	10	3	3				
2.	new coordinate system where the most significant variations are captured.  A family enjoys different weekend activities such as going to the cinema, playing tennis, shopping, or staying in. Their choice depends on several factors: the weather, whether parents are present, and their financial status. However, they do not follow a strict routine, making it challenging to predict their decisions. Use the given data, to uncover the decision-making pattern of the family. By applying a decision tree model, determine how different factors influence their weekend plans.						3	
	Weekend W1	Weather Sunny	Parents Yes	Money Rich	Decision Cinema	10	4	5
	W2	Sunny •	No	Rich	Tennis			
	W3	Windy	Yes	Rich	Cinema			
	W4	Rainy	Yes	Poor	Cinema			
	W5	Rainy	No	Rich	Stay In			
3.	The bank cor (Employed/Unen	nsiders thre nployed), Cr historical loa	ee categorical edit Score ( n data, the bai	l factors: l Good/Poor), nk aims to pr	n applicants' profiles. Employment Status and Existing Debt edict whether a new = 3	10	4	3
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	Dataset:				and the second s			
	Applicant	Employment	Credit Score	Existing Debt	Loan Status			
	A1	Employed	Good	No	Approved	And the second s		
	A2	Unemployed	Good	No	Approved			
	A3	Employed	Poor	No	Approved			
	A4	Employed	Good	Yes	Approved			
	A5	Unemployed	Poor	No	Rejected			
	A6	Employed	Poor	Yes	Rejected			
	A7	Unemployed	Poor	Yes	Rejected			
	A8	Employed	Good	No	Approved			
	A9	Unemployed	Good	Yes	Approved			
	AS	Offering 10			The state of the s	1	1	1
<b>F.</b>	classified as transactions ensemble mo Explain how method you	legitimate) ar flagged as fraud odel using baggir each technique o would prioritize	d). Given this sc ng and boosting t contributes to re for this applicati	enario, how wot o optimize detect ducing misclassif on.	Approved  ulent transactions itives (legitimate ald you design an tion performance? fication and which	10	5	
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