Name:

**Enrolment No:** 



Semester

Max. Marks: 100

Time

: III

:03 Hours

## UNIVERSITY OF PETROLEUM AND ENERGY STUDIES School of Computer Science

End Semester Examination, December 2020

Course : D	Pata Mining & Prediction by Machines
Program : B	B. Tech CSE AIML
<b>Course Code</b>	: CSAI 2005
Instructions	:

## **SECTION A**

		Marks	
Q1	Why data mining is in high demand and what kind of data can be mined?	05	<b>CO1</b>
Q2	What is the difference between classification and regression? Why data classification is known as a two-step process?	05	CO3
Q3	<ul> <li>Given two objects represented by the tuples (22, 1, 42,10) and (20, 0, 36, 8):</li> <li>a) Compute the Euclidian distance between the two objects.</li> <li>b) Compute the Manhattan distance between the two objects.</li> </ul>	05	CO2
Q4	What are the different methods of measuring central tendency of data set?	05	CO2
Q5	How bagging method can be useful to improve the accuracy of classification machine models?	05	CO3
Q6	Give an application example of where the border between normal objects and outliers is often unclear, so that the degree to which an object is an outlier has to be well estimated.	05	CO4
	SECTION B		
Q7	<ul> <li>Suppose that the data mining task is to cluster the following eight points (with (x, y) representing location into three clusters:</li> <li>A1(2, 10), A2(2, 5), A3(8, 4), B1(5, 8), B2(7, 5), B3(6, 4), C1(1, 2), C2(4, 9):</li> <li>The distance function is Euclidean distance. Suppose initially we assign A1, B1, and C1 as the center of each cluster, respectively.</li> <li>a) Write down k-means algorithm</li> <li>b) Use k-means algorithm for the three cluster centers after the first round execution</li> <li>c) Find the final three clusters</li> </ul>	4+4+2 =10	CO4

Q8	What do v	ou mean by Pro	cess Stand	ardization	n? Brief	ly explain	the CRIS	P-DM	4.0	001
ς-	phases and	-			·	<b>J</b>			10	CO1
Q9	Explain KNN algorithm. Why it is also called Lazy Learner? What are the points to be subjected when choosing the value of k? For the below problem predict for the class of Davis using KNN and assume the value of k=3.									
		Custo	omer Age	Income (K)	No. of cards	Response	]			
		John	35	35	3	Yes	1			
		Rache	el 22	50	2	No			10	
		Ruth	63	200	1	No				CO3
		Tom	59	170	1	No	-			
		Neil	25	40	4	Yes	1			
		David	37	50	2	?				
	people. Fol	lowing is the conf	-	ix for the s	same. Predict	ed		onsists of 100		
	people. Fol	lowing is the cont	fusion matr	ix for the s	same. Predict	ed Positive				
	people. Fol	lowing is the conf	fusion matr	ix for the s	same. Predict	Positive			10	CO4
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suitabi	e examp	105.						
				Section C				
a) b)	<ul> <li>Consider the data as two-dimensional data points. Given a new data point, x= (1.4,1.6) as a query, rank the database points based on similarity with the query using Euclidean distance, Manhattan distance, Minkowski distance.</li> <li>Following dataset is used to learn a decision tree which predicts if a student passed data mining and prediction by machine (Yes or No), based on their previous GPA (High, Medium, or Low) and whether or not they studied. Draw the decision tree for the same. Also, show the calculations regarding entropy and information gain.</li> </ul>							
	GPA		Studie	d	Passed			
	Low		False		No			
	Low		True		Yes			
	Mediu	m	False		No			
	Mediu	m	True		Yes			
	High		False		Yes			
	High		True		Yes		20	СО
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