VASAVI COLLEGE OF ENGINEERING (AUTONOMOUS), HYDERABAD

Accredited by NAAC with A++ Grade

M.Tech. (C.S.E.) II-Semester Main Examinations, September-2022 Data Mining

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 ques	tion	M	L	CO	PO	
1.	For the following vectors x and y, Compute 0	Cosine similarity.	2	1	1	1,2	
	x = (0, 1, 2, 1) $y = (1, 2, 0, 3)$	Micheller can of box					
2.	Draw a diagram to show the steps of KDD.	present a town of sensing symbol many	2	1	1	1,2	
		of the second of					
3.	Compute Entropy at a node with 10 samples	ntropy at a node with 10 samples and C1=2, C2=8.					
4.	Compare the role of Training data and Test of	role of Training data and Test data in a Classification model.					
5.	Given below is a simple ANN with 2 inputs -3, +2, +2	$X1, X2 \in \{0, 1\}$ and edge weight	2	2	3	1,2	
	$h\left(x\right) = \begin{cases} 1 & if \ x \\ 0 & other$	≥ 0 rwise					
	3						
	(X1)—+2—→(h	(x)					
	+2	· manuscript					
	X2 +2	on A. A. Shirts belond may bouled	YOUR				
	1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1		mixe				
	Which logical function does the ANN comp	ute?	H zer				
6.	Differentiate linearly separable and non-linear	arly separable problems. Give one	2	1	3	1,2	
	example for each.						
7.	Find the support count of {a}, {b}, {a, c}, {b} based on the following table:	a, b}.	2	2	4	1,2	
		emsets					
		i, b, c}					
	NAME AND ADDRESS OF THE PROPERTY OF THE PROPER	ı, c}					
		a, d}					
	4 {{	o, e, f}					
8.	What is Association Rule Mining? List Association Rule Mining.	two real-world applications of	2	1	4	1,2	
9.	Define Minpts and Epsilon used in DBSCA	N algorithm.	2	1	5	1,2	
10.	What is Dendrogram. Give an example.		2	1	5	1,2	
10.	what is Dehutogram. Give an example.			1		1,4	

Code No.: 22958

	Why we need techniques.	d to perform da	ita pre-proce	essing? Ex	xplain data pre-	processing	4	2	1	1,2
b)		nave the followi	ng 2-D data	set:			4	2	1	1,2
	I have now		A1		A2					and f
	x1	1-3142 78713	1.5	10 10 10 10 10 10 10 10 10 10 10 10 10 1	1.7	ralo assen	1			
	x2		2	- 12 - 3	1.9					
	х3		1.6		1.8					
	x4		1.2		1.5					
	x5		1.5		1.0	W X S. (3) / 7				
	Consider the	data as 2-D data	noints Giv	en a new	data point, $x = ($	1416) 25				
2. a)	a query, rank Euclidean dis	the database patance, and Man	points based hattan distan	on simila	arity with the quadratused in F	uery using	3	2	2	1,2,
. u)	classification		algorium to	create a N	die-set used in r	cuic-based	3	4	2	1,2,
b)		main with three es T and F. Giv			C. Each attributestances.	te has two	5	3	2	1,2,
		A	BC	TARGET	Parking the T	17.10				
		T	TT	YES						
		T	TF	NO						
		T	FT	YES						
		F	TT	YES						
		F	TF	NO	NO FINE	/ 13				
		F	FF	YES						
	Calculate the	information gai	n for the attr	ributes A	and C.					
3. a)		ide that a person	having chil	ls, fever, i	s seen by a doct mild headache a ithm for predicti	nd without	5	3	3	1,2,
o. a)	· ·	has the flue? (C	So I tavio isc							
o. a)	running nose	running nose	headache	fever	has flu		tabal dis			
). a)	running nose			fever Y	has flu N		nereal abie			
o. a)	running nose	running nose	headache		The state of the s	y separable and of (a).	die die oo no			
o. a)	running nose chills Y	running nose	headache mild	Y	N	y separable into of (a), rog table:	nasal objection wolle			
o. a)	chills Y Y	running nose N Y	headache mild no	Y N	N Y	y separable int of (a), nog table: however, table:	die die no me wolle			
o. a)	chills Y Y Y N N	running nose N Y N Y N Y	headache mild no strong	Y N Y	N Y Y	y separable unit of (n), nog table: homeographic	nesali side on co collo			
o. a)	chills Y Y Y N N N	running nose N Y N Y N Y N Y	headache mild no strong mild no strong	Y N Y Y N	N Y Y Y N Y	y separable int of (a), nog table: however, able: however, able: 1	albe albe ap ma volle			
э. а)	running nose chills Y Y Y N N N N	running nose N Y N Y N Y Y Y	headache mild no strong mild no strong strong	Y N Y Y N	N Y Y Y N N	y separable and of (a), and of (a), one table: Yoursection 1	naz il dise naz zia wollo			
э. а)	chills Y Y Y N N N	running nose N Y N Y N Y N Y	headache mild no strong mild no strong	Y N Y Y N	N Y Y Y N Y	y separable and of (a), and of	usa ulbe op me wollo			

т.	a)	Construct the Fp	tree from the belov	w transaction d	atabase.	ALC:	5	3	4	1,2,3
		100 \{f, a, b \} 200 \{a, b \} 300 \{b, f \} 400 \{b, c \}	s bought c, d, g, i, m, pl d, c, f, l, m, ol f, h, j, ol c, k, s, pl d, c, e, l, p, m, nl							
	b)	An are not to the second secon	sociation rule mining	narama e timunuali	on the data	with continuous	3	2	4	1,2,3
5.	a)		o find the distance between the clusters given below. A2=(3,5), A3=(8,4), A4=(5,8)}, C2= { A5=(7,5), A6=(4,4), B(9)}					3	5	1,2,3
	b)	Explain the step	s in DBSCAN algor	rithm to perform	n density-l	pased clustering.	3	2	5	1,2,
	a)		ollowing data using			La se propio	4	3	1	1,2
	b)		s of Decision tree in	duction algori	thm.		4	2	2	1,2
7.		Answer any <i>two</i> of the following:								
	a)						4	3	3	1,2,
				Strongth		_				
		Name	Acid Durability	Strength	Class					
		Type-1	7	7	Class F					
		Type-1	7	7	F					
		Type-1 Type-2	7 7	7	F					
	b)	Type-1 Type-2 Type-3 Type-4 Trace the result with support thr	7 7 3 1 s of using the Aprioreshold S=33.34% arequent itemsets for	7 4 4 4 ori algorithm or and Confidence	F F T T the groces	=60%. Show the	4	3	4	1,2,
	b)	Type-1 Type-2 Type-3 Type-4 Trace the result with support thr candidate and fr	7 7 3 1 s of using the Aprioreshold S=33.34% arequent itemsets for	7 4 4 4 ori algorithm or and Confidence	F F T T the groces	=60%. Show the	4	3	4	1,2,
	b)	Type-1 Type-2 Type-3 Type-4 Trace the result with support thr candidate and fr frequent itemset	7 7 3 1 s of using the Aprioreshold S=33.34% are equent itemsets for exts	7 4 4 ari algorithm or and Confidence each database s	F F T T the groces	=60%. Show the	4	3	4	1,2,
	b)	Type-1 Type-2 Type-3 Type-4 Trace the result with support thr candidate and fr frequent itemset	7 7 3 1 s of using the Aprio eshold S=33.34% are equent itemsets for exts	7 4 4 ari algorithm or and Confidence each database s	F F T T the groces	=60%. Show the	4	3	4	1,2,
	b)	Type-1 Type-2 Type-3 Type-4 Trace the result with support thr candidate and fr frequent itemset	7 7 3 1 s of using the Aprio eshold S=33.34% are equent itemsets for exts Items HotDogs, Buns, F	7 4 4 ori algorithm or and Confidence each database s	F F T T the groces	=60%. Show the	4	3	4	1,2,
	b)	Type-1 Type-2 Type-3 Type-4 Trace the result with support thr candidate and fr frequent itemset Transaction ID T1 T2 T3 T4	7 7 3 1 s of using the Aprior eshold S=33.34% are equent itemsets for ets Items HotDogs, Buns, HotDogs, Buns HotDogs, Coke, Coke, Coke, Coke, Coke	7 4 4 ori algorithm or and Confidence each database s	F F T T the groces	=60%. Show the	4	3	4	1,2,
	b)	Type-1 Type-2 Type-3 Type-4 Trace the result with support thr candidate and fr frequent itemset Transaction ID T1 T2 T3 T4 T5	7 7 3 1 s of using the Aprio eshold S=33.34% are equent itemsets for ets Items HotDogs, Buns, HotDogs, Buns HotDogs, Coke, Chips, Coke Chips, Ketchup	7 4 4 4 ori algorithm or and Confidence each database s	F F T T the groces	=60%. Show the	4	3	4	1,2,
	b)	Type-1 Type-2 Type-3 Type-4 Trace the result with support thr candidate and fr frequent itemset Transaction ID T1 T2 T3 T4	7 7 3 1 s of using the Aprior eshold S=33.34% are equent itemsets for ets Items HotDogs, Buns, HotDogs, Buns HotDogs, Coke, Coke, Coke, Coke, Coke	7 4 4 4 ori algorithm or and Confidence each database s	F F T T the groces	=60%. Show the	4	3	4	1,2,

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

i)	Blooms Taxonomy Level – 1	20%
ii)	Blooms Taxonomy Level – 2	40%
iii)	Blooms Taxonomy Level – 3 & 4	40%