	Precision and Tuple no.	1	2	3	4	5	6	7	8	9	10	11	12		
	Actual	P	P	<u>Р</u>	P	N	N	P	N	N	P	P	N		
	Label														
	Predicted Lebel	P	P	N	N	N	N	P	P	P	N	P	N		
)	Discuss attri	bute s	electi	on ci	iteria	for t	the de	ecisi	on tr	ee al	gorith	nm.			
)	Explain the s	suppo	rt vec	tor m	nachir	ne in	detai	il.							
	Apply SVM	for X	OR p	robl	em.										
)	How to impr	ove th	ne per	form	ance	of th	e cla	ssifi	er m	odel'	?				
)	Write a note on Ensembling techniques.					ique	S.								
	Consider following for each class		_	ulate	e weig	ghted									
	Consider foll		_	eulate		ghted cted		rage	Prec		and				
1	Consider fold for each class	s. Als	_	eulate	e weig Predi	ghted cted	d aver	rage	Prec	ision	and				
	Consider foll	s. Als	o calc	eulate	e weig Predic Class 1	ghted cted	Clas	rage	Prec	ision	and				
	Consider fold for each class	cl:	o calc	eulate	e weig Predic Class 1	ghted cted	Class	rage s 2	Prec	lass 3	and				
	Consider fold for each class	cl:	ess 1 ess 2 ess 3	culate	Predictions 15	ghted cted	Clas	rage s 2 2 15 3	Prec	lass 3 3 8 45	and	Weig			
	Consider following for each class	Cla Cla	ass 1 ass 2 ass 3	alue	Predictions 15 7 2 of K	ghted cted	Clas	s 2 2 15 3	Prec	lass 3 8 45 KNN	steps	Weig	hted a		
	Consider following for each class Actual How to select Consider the	Claret the last	ass 1 ass 2 ass 3 best v	alue	Predictions 1 15 7 2 of K	in K	Class NN.	s 2 2 15 3 Exp	Precolation I	lass 3 8 45 KNN	steps	Weigi 	e di	avera;	
	Consider follows for each class Actual How to select	Claret the last	ass 1 ass 2 ass 3 best v	alue	Predictions 1 15 7 2 of K	in K	Class NN.	s 2 2 15 3 Exp	Precolation I	lass 3 8 45 KNN	steps	Weigi 	e di	avera;	
	Consider following for each class Actual How to select Consider the	Claret the last	ass 1 ass 2 ass 3 best v	alue	Predictions 1 15 7 2 of K	in K	Class NN.	s 2 2 15 3 Exp	Precolation I	lass 3 8 45 KNN	steps	Weigi 	e di	avera;	
	Consider following for each class Actual How to select Consider the	Claret the last	ass 1 ass 2 ass 3 best v	alue	Predictions 1 15 7 2 of K	in K	Class NN.	s 2 2 15 3 Exp	Precolation I	lass 3 8 45 KNN	steps	Weigi 	e di	avera;	

Predicted

Actual

	Α	В	С
Α	8	2	0
В	1	9	0
С	1	2	7

- 9) Give steps of Ensemble Learning. Explain majoring voting.
- 10) Construct Decision Tree upto three levels for following data.

Day	Weather	Temperature	Humidity	Wind	Play?
1	Sunny	Hot	High	Weak	No
2	Cloudy	Hot	High	Weak	Yes
3	Sunny	Mild	Normal	Strong	Yes
4	Cloudy	Mild	High	Strong	Yes
5	Rainy	Mild	High	Strong	No
6	Rainy	Cool	Normal	Strong	No
7	Rainy	Mild	High	Weak	Yes
8	Sunny	Hot	High	Strong	No
9	Cloudy	Hot	Normal	Weak	Yes
10	Rainy	Mild	High	Strong	No

11) Construct a decision Tree for the following data

Instance	A1	A2	A3	Classification
1	True	Hot	High	No
2	True	Hot	High	No
3	False	Hot	High	Yes
4	False	Cool	Normal	Yes
5	False	Cool	Normal	Yes
6	True	Cool	High	No
7	True	Hot	High	No
8	True	Hot	Normal	Yes
9	False	Cool	Normal	Yes
10	False	Cool	High	Yes

Apply KNN and find Label (Liked) for (Apple: false, Mango:True, Grapes: True), Consider, K=3

Apple Mango Grapes Liked

A True True True False

B True False False True

C False True True False

D False True False True

E True False False True

13) Apply Naïve Bais and find Label (Liked) for (Apple: false, Mango:True, Grapes: True)

	Apple	Mango	Grapes	Liked
A	True	True	True	False
В	True	False	False	True
С	False	True	True	False
D	False	True	False	True
Е	True	False	False	True

14). Solve following Numerical example for Linear SVM:
Positively labelled data points (3,1)(3,-1)(6,1)(6,-1) and Negatively labelled data points (1,0)(0,1)(0,-1)(-1,0)

Construct the Ensemble learning Model for following cases. Consider any example.

By manipulating the training set, By manipulating the input features, By manipulating

By manipulating the training set, By manipulating the input features, By manipulating learning algorithms.

Construct a decision tree using CART algorithm with the below training dataset shown in the table.

Instance	A1	A2	A3	Classification
1	True	Hot	High	No
2	True	Hot	High	No
3	False	Hot	High	Yes
4	False	Cool	Normal	Yes
5	False	Cool	Normal	Yes
6	True	Cool	High	No
7	True	Hot	High	No
8	True	Hot	Normal	Yes
9	False	Cool	Normal	Yes
10	False	Cool	High	Yes

Unit V

- 1) Explain the K-means algorithm in detail.
- Two clusters have the following data points. Calculate their intra cluster distance. Also calculate total inertia of the data point assignment.

He	1	1	1	1	1	1	1	1	1	1	1	1
igh	8	7	6	7	8	8	8	8	8	8	8	7
t	5	0	8	9	2	8	0	0	3	0	0	7
We	7	5	6	6	7	7	7	7	8	8	6	7
igh t	2	6	0	8	2	7	1	0	4	8	7	6
As sig ne d clu ste r	2	1	1	2	2	2	2	2	2	2	2	2

	Differentiate between the Partitioning method and hierarchical method of clustering.						
4)	Give advantages based method	and disadvantages	of Partitioned metho	ods, Hierarchical methods, Density			
5)	Consider flowing	g instances given as	input to K-means c	lustering algorithm, k=3.			
	Find members of	f these 3 clusters after	er two iterations				
	$X=\{(2,10), (2,5)\}$, (8,4), (5,8), (7,5), ((6,4), (1,2), (4,9)}				
6)	Compare Single	link, Average link a	nd Complete link c	lustering with example of any one			
	type.						
7)				s on this. Assume that minimum			
		d (s = 2) and minimum	um confident thresh	old $(c = 60\%)$			
	Transaction I		Items				
	T1 T2	Hot Dogs, Bur		4			
	T3	Hot Dogs, Bur Hot Dogs, Cok		-			
	T4	Chips, Coke	to, ompo	┪			
	T5	Chips, Ketchu					
	T6	Hot Dogs, Cok	Hot Dogs, Coke, Chips				
8)	Consider flowing instances given as input to K-means clustering algorithm, k=3.						
Ο,	Find members of these 3 clusters after two iterations						
0)			-				
0)	Find members of		er two iterations				
9)	Find members of $X = \{(2,10), (2,5)\}$	f these 3 clusters after, (8,4), (5,8), (7,5), (basket analysis using	er two iterations (6,4), (1,2), (4,9)}	nd discussed the key terms			
	Find members of $X=\{(2,10), (2,5)\}$ Explain Market I	f these 3 clusters after, (8,4), (5,8), (7,5), (basket analysis using t.	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i	f these 3 clusters after, (8,4), (5,8), (7,5), (basket analysis using t.	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da	f these 3 clusters after, (8,4), (5,8), (7,5), (basket analysis using t. ta is as follows: List of Item IDs	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da	f these 3 clusters after, (8,4), (5,8), (7,5), (basket analysis using t. ta is as follows: List of Item IDs 11, 12, 13	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2	ta is as follows: List of Item IDs 11, 12, 13 12, 14	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1	f these 3 clusters after (8,4), (5,8), (7,5), (basket analysis using t. ta is as follows: List of Item IDs 11, 12, 13 12, 14 12, 13	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4	f these 3 clusters after, (8,4), (5,8), (7,5), (basket analysis using t. ta is as follows: List of Item IDs 11, 12, 13 12, 14 12, 13 11, 12, 14	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5	f these 3 clusters after (8,4), (5,8), (7,5), (basket analysis using the transfer of the second state of t	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5 Y6	f these 3 clusters after (8,4), (5,8), (7,5), (basket analysis using the transfer of the trans	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5 Y6 T7	f these 3 clusters after (8,4), (5,8), (7,5), (basket analysis using the transfer of the trans	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5 Y6	f these 3 clusters after (8,4), (5,8), (7,5), (basket analysis using the transfer of the trans	er two iterations (6,4), (1,2), (4,9)}				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5 Y6 T7 T8 T9	f these 3 clusters after (8,4), (5,8), (7,5), (9 basket analysis using the transfer of the tra	er two iterations (6,4), (1,2), (4,9)} g association rule ar				
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5 Y6 T7 T8 T9	f these 3 clusters after (8,4), (5,8), (7,5), (10 basket analysis using the context of the second state of	er two iterations (6,4), (1,2), (4,9)} g association rule ar	nd discussed the key terms			
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5 Y6 T7 T8 T9 Find the frequen Or with FP grow	f these 3 clusters after (8,4), (5,8), (7,5), (10 basket analysis using the context of the second state of	er two iterations 6,4), (1,2), (4,9)} g association rule ar	nd discussed the key terms			
9)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5 Y6 T7 T8 T9 Find the frequen Or with FP grow How Apriori alg	ta is as follows: List of Item IDs I1, I2, I3 I2, I4 I2, I3 I1, I2, I3 I1, I2, I3 I1, I2, I3 I1, I3, I3 I1, I4 I1, I5 II, I5 III, II	er two iterations (6,4), (1,2), (4,9)} g association rule are the association rule in in detail.	nd discussed the key terms			
9) 10) 11)	Find members of X={(2,10), (2,5)} Explain Market I associated with i Transactional da TID T1 T2 T3 T4 T5 Y6 T7 T8 T9 Find the frequen Or with FP grow How Apriori alg	f these 3 clusters after (8,4), (5,8), (7,5), (basket analysis using t. ta is as follows: List of Item IDs I1, I2, I3 I2, I4 I2, I3 I1, I2, I4 I1, I3 I1, I3 I1, I2, I3 I1, I3 I1, I2, I3 t itemset and general of the algorithm.	er two iterations (6,4), (1,2), (4,9)} g association rule are the association rule in in detail.	nd discussed the key terms			

	(a1, a4, a5, a6)
	(a1, a2, a4, a7)
	(a2, a3, a4, a7)
	(a2, a3, a5, a6)
	Find Association Rules for given support and confidence (3, 50%).
13)	Give steps of the Apriori Algorithm
14)	Compare Apriori and FP Growth.
15)	What are the advantages of FP growth algorithm over Apriori?
Unit VI	
1)	Apply Artificial neural network for disease prediction. Comment using performance metrics.
2)	Give steps for building a Recommender System. Explain in detail.
3)	Apply and analyse machine learning algorithm for medical diagnosis of diabetes based on patient data such as medical history, lab test results, and lifestyle factors. Comment on the type of algorithm can be applied and the performance evaluation parameters used for it.
4)	Apply machine learning specifically for predicting the stock prices, analyze the performance evaluation parameters and the type of machine learning algorithm suitable for it.
5)	How can machine learning be applied to Product Recommendation on E-commerce Websites? Explain steps involved in Product Recommendation using machine learning
6)	Apply and analyse machine learning algorithm for medical diagnosis of diabetes based on patient data such as medical history, lab test results, and lifestyle factors. Comment on the type of algorithm can be applied and the performance evaluation parameters used for it.
7)	Apply Artificial neural network for disease prediction. Comment using performance metrics.
8)	Design a Recommender System for an ecommerce platform. Explain all steps involved in the process.
9)	Apply machine learning specifically for predicting the stock prices, analyze the performance evaluation parameters and the type of machine learning algorithm suitable for it.
10)	How can machine learning be applied to Product Recommendation on E-commerce Websites? Explain steps involved in Product Recommendation using machine learning
11)	Which algorithm is best suited for stock prediction? Explain in detail.
12)	Apply Artificial neural network for disease prediction. Comment using performance
L	ı

_	metrics.