

### UNIT – I

#### **INTRODUCTION TO MACHINE LEARNING**

1.	a)	Define Machine Learning What are the advantages of Machine Learning.	[CO1][L1]	5M
	b)	How do you select Machine Learning model.	[CO1][L3]	5M
2.		Classify various types of Machine Learning & explain any ML model with an	[CO1][L2]	<b>10M</b>
		example.		
3.		Explain the applications of Machine Learning.	[CO1][L2]	<b>10M</b>
4.	a)	Explain concepts of learning by Rote & Induction with an example.	[CO1][L2]	5M
	b)	Briefly discuss various Dataset's in Machine Learning	[CO1][L2]	5M
5.	a)	Briefly explain the concepts of Reinforcement Algorithm	[CO1][L2]	5M
	b)	Discuss about various types of Data	[CO1][L1]	5M
6.		Explain different stages in Machine Learning	[CO1][L2]	<b>10M</b>
7.	a)	Briefly describe the procedure to handle duplicate records	[CO1][L3]	5M
	b)	Explain about feature engineering in Machine Learning	[CO1][L3]	5M
8.		What is the role of searching and learning in Machine Learning Explain Grid	[CO1][L4]	<b>10M</b>
		search & Random search with an example		
9.		Explain different Data collection Methods	[CO1][L2]	<b>10M</b>
10.	a)	What is the need of label encoding	[CO1][L1]	<b>2M</b>
	b)	What is meant by supervised and Unsupervised learning in ML	[CO1][L1]	<b>2M</b>
	c)	Define machine learning.	[CO1][L1]	<b>2M</b>
	d)	What is meant by Label Dataset in ML	[CO1][L1]	<b>2M</b>
	e)	What is meant by regression problem	[CO1][L1]	2M

#### UNIT – II

#### **NEAREST NEIGHBOR-BASED MODELS**

1.	a)	Explain the steps involved in Nearest Neighbour Models.	[CO2][L2]	5M
	b)	List out the applications of Manhattan Distance in ML.	[CO2][L1]	5M
2.		Explain Euclidian Distance measure with one example.	[CO2][L2]	<b>10M</b>
3.		Explain Manhattan Distance measure with example.	[CO2][L2]	<b>10M</b>
4.	a)	Explain the concept of Murkowski distance.	[CO2][L2]	5M
	b)	Explain hamming distance for any two binary and DNA patterns.	[CO2][L2]	5M
5.		Explain Non Metric Proximity measures with one example.	[CO2][L2]	<b>10M</b>
6.		Explain the steps involved in K-Nearest Neighbors algorithm	[CO2][L2]	<b>10M</b>

7. Classify a new customer as "high-spending" or "low-spending" based on their [CO2][L3] 10M age and income by using KNN algorithm.

Customer	Age	Income (Rupees)	Spending Class
А	25	30,000	Low
В	45	80,000	High
С	35	50,000	Low
D	50	90,000	High
E (New)	40	60,000	

A bank wants to classify customers as "Low-Risk" or "High-Risk" for loans. [CO2][L3] 10M 8. Classify Customer E using Manhattan Distance and Radius Distance Nearest Neighbour Algorithm. Assume suitable radius.

Customer	Ago	Income	Credit	Loan Amount	Risk
Customer	Age	(\$)	Score	(Rupees)	Level
А	25	30,000	700	10,000	Low
В	45	80,000	600	40,000	High
С	35	50,000	750	20,000	Low
D	55	90,000	580	50,000	High
E (New)	40	60,000	680	25,000	

- 9. Explain KNN Regression with one example. Also list out advantages [CO2][L2] 10M 10. a) List out the performance measures of Regression. [CO2][L1] **2M** List out the performance measures of Classification. b)
  - Define MAE and  $R^2$ . c)
  - How does KNN regression differ from KNN classification d)
  - List out common distance measures used in machine learning. [CO2][L1] **2M** e)

- [CO2][L1] **2M**
- [CO2][L1] **2M**
- [CO2][L2] **2M**

### UNIT – III

### **MODELS BASED ON DECISION TREES & THE BAYES CLASSIFIER**

1.	-	Explain the key steps involved in building a Decision Tree for classification.	[CO4][L2]	<b>10M</b>
2.		Discuss how impurity is measured in decision trees using Gini Index and	[CO3][L4]	<b>10M</b>
		Entropy with examples.		
3.		Describe the steps involved in Decision Tree is built for regression with one	[CO3][L3]	<b>10M</b>
		example.		
4.		Describe the steps involved in training a Random Forest for classification.	[CO4][L2]	<b>10M</b>
5.		Apply Random Forest Regression to a sample dataset and explain how it	[CO3][L3]	<b>10M</b>
		predicts continuous target values. Mention any evaluation metrics used.		
6.		Explain the key concepts of Bayes' Rule and how it is used in classification.	[CO4][L2]	<b>10M</b>
		Provide an example.		
7.		Describe the steps involved in building a Bayes Classifier for binary	[CO4][L3]	<b>10M</b>
		classification using Bayes' Rule.		
8.		What is the Naive Bayes Classifier Explain the assumption of class	[CO4][L2]	<b>10M</b>
		conditional independence and how it simplifies computation.		
9.		Apply the Naive Bayes Classifier to a real-world multi-class problem such as	[CO4][L3]	<b>10M</b>
		spam detection. Outline the model building, prediction, and evaluation steps.		
10.	a)	What is the main criterion used to split nodes in a decision tree during	[CO4][L1]	<b>2M</b>
		classification		
	b)	State one key difference between decision trees used for classification and for	[CO3][L2]	<b>2M</b>
		regression.		
	c)	What is the role of the bias-variance trade-off in decision trees	[CO3][L4]	<b>2M</b>
	d)	In the context of Bayes' Rule, what is meant by the 'prior probability'	[CO3][L1]	<b>2M</b>
	e)	What assumption does the Naive Bayes Classifier make about features	[CO4][L1]	2M

#### $\mathbf{UNIT} - \mathbf{IV}$

#### LINEAR DISCRIMINANTS FOR MACHINE LEARNING

- 1. Explain the steps involved in training a Perceptron classifier with a simple [CO5][L3] 10M example.
- 2. What is Linear Discriminant Analysis (LDA) Explain LDA steps for [CO5][L2] 10M classification.
- 3. Explain the Perceptron Learning Algorithm with steps. How does the [CO5][L3] 10M algorithm converge for linearly separable data
- 4. Describe the working of Support Vector Machines (SVMs) for linearly [CO5][L2] 10M separable data. Explain margin, hyper plane, and support vectors.
- 5. What happens when data is not linearly separable in SVM Explain how soft [CO5][L4] 10M margin and Kernel Trick help.
- 6. Explain the concept of the Kernel Trick in SVM with an example. Why is it [CO5][L3] 10M useful for non-linear classification
- 7. Compare and contrast Logistic Regression and Linear Regression. Give one [CO5][L4] 10M example each.
- 8. What is a Multi-Layer Perceptron (MLP) Describe its architecture and how it [CO5][L2] 10M extends the single-layer perceptron.
- 9. Explain the Backpropagation algorithm for training an MLP. Include forward [CO5][L3] 10M pass, error calculation, and weight update steps.
- 10. a) What is the main objective of the Perceptron Learning Algorithm[CO5][L1]2M
  - b) Define the term 'Support Vector' in Support Vector Machines (SVM). [CO5][L1] 2M
  - c) What is the role of the activation function in a Multi-Layer Perceptron [CO5][L2] 2M
  - d) Mention one key difference between Linear Regression and Logistic [CO5][L1] 2M Regression.
  - e) What does the Kernel Trick do in the context of SVM [CO5][L2] 2M

#### $\mathbf{UNIT} - \mathbf{V}$

#### **CLUSTERING**

1. What is clustering Explain the types of clustering methods with simple [CO6][L2] 10M examples. 2. Describe the K-Means clustering algorithm. What are the steps involved in [CO6][L3] 10M the algorithm Provide an example. 3. What is Fuzzy C-Means clustering How does it differ from K-Means [CO6][L3] 10M clustering Explain Agglomerative and Divisive hierarchical clustering methods. How are [CO6][L3] 10M 4. they different 5. What is Soft Clustering How is it used in clustering methods like Fuzzy C- [CO6][L4] 10M Means What is Matrix Factorization in clustering How does it help in clustering large [CO6][L3] 10M 6. datasets 7. Explain Rough Clustering and the Rough K-Means algorithm. How is it [CO6][L4] 10M different from traditional K-Means 8. What is the Expectation Maximization (EM) algorithm How does it work for [CO6][L3] 10M clustering data Explain Spectral Clustering. How does it work, and when is it useful for [CO6][L4] 10M 9. clustering a) What is the difference between Agglomerative and Divisive clustering [CO6][L1] 10. **2M** methods **b**) What is a centroid in K-Means clustering [CO6][L1] **2M** What is Soft Clustering Give an example. [CO6][L2] **2M c**) How does Fuzzy C-Means differ from K-Means [CO6][L2] **2M** d) What is Matrix Factorization in clustering Why is it useful [CO6][L2] **2M** e)

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#### **BIT BANK (OBJECTIVES)**

Subject with Code: Machine Learning (23CS0902)

Regulation: R23

Course & Branch: B.Tech – CSM,CAD & CAI

Year & Sem: II Year & II Sem

# UNIT-I

## **Introduction to Machine Learning**

1	Machine Learning is an application of .	[ ]
	A)Block Chain B) Neural Network C) Artificial Intelligence D) None	
2	Application of Machine Learning is.	[ ]
	A)Email filtering B) Sentimental Analysis C) Face Recognition D) All of the a	bove
3	Among the following option identify the one which is not a type of learning	[ ]
	A)Semi Supervised B) Supervised C) Un Supervised D) Reinforcement	
4	If the coefficient of determination is a positive value, then the regression equation	[ ]
	A)must have a positive slope B)must have a negative slope	
	C)either a positive or a negative slope D)None	
5	Type of matrix decomposition model is	[ ]
	A)Predictive Model B) Descriptive Model C) Logical Model D) Theorem M	Iodel
6	Identify the kind of learning algorithm for "facial identities for facial expressions".	[ ]
	A)Prediction <b>B</b> ) <b>Recognition patterns</b> C)Recognizing anomalies D) Generating	Patterns
7	Machine learning is a subset of which of the following.	[ ]
	A)Deep Learning B) Data Learning C) Artificial Intelligence D) Negative Fo	unction
8	Which of the following machine learning techniques helps in detecting the outliers in data	ι[ ]
	A)Classification B) Clustering C) Anomaly Detection D)Data Set	
9	The most significant phase in genetic algorithm is	[ ]
	A)Mutation B) Selection C) Fitness Function D)Crossover	
10	Which of the following are common classes of problems in machine learning	[ ]
	A)Regression B) Classification C) Clustering D) All of the above	
11	Which of the following is not a supervised learning	[ ]
	A)PCA B) Naïve Bayesian C) Linear Regression D) Decision Tree	
12	Which of the following is not a machine learning algorithm [	]
	A)SVM <b>B</b> ) SVG C)Random Forest D) KNN	
13	Among the following option identify the one which is used to create the most common	[ ]
	A)plot B) quickplot C) qplot D) quickerplot	
14	The Real-world machine learning use cases are	[ ]
	A)Digital Assistant B) Chatbots C)Fraud Detection D) All of the above	
15	An un supervised learning problems can be grouped as	[ ]
	A)Clustering B) Association C) Both A & B D) Decision Tree	
16	In Machine learning the model that must solve the given performance task is known as	[ ]
	A)Critic B) Generalize C) <b>Performance System</b> D) All of the above	
17	Linear regression is a regression techniques that is used to model data having a outcome	[ ]
	A)Linear, Binary B) Linear, Numeric C) Nonlinear, binary D) Nonlinear,	Numeric
18	In what type of learning labelled training data is used	[ ]

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	Un supervised B) Reinforcement C) Active learning D)Supervised		
,	at does dimensionality reduction reduce	[	1
	stochastics <b>B</b> ) collinerity C) performance D) entropy	L	1
,	pe of matrix decomposition model is	ſ	1
	Descriptive model <b>B) Predictive Model</b> C) Logical Model D) Validation data	L	1
	A is	ſ	1
	Forward features B) Backward features C) Feature extraction A) D)Feature I	L Predicti	ion
· · · · · · · · · · · · · · · · · · ·	e average squared difference between classifier predicted output and actual output	ſ	1
	<b>mean squared error</b> B) root mean squared errorC) mean absolute errorD) mean relati	L ive erro	J Nr
	e problem of finding hidden structure in unlabelled data is called	ſ	" 1
	<b>Unsupervised</b> B)Supervised C) Reinforcement D) A&B	L	1
<i>,</i>	tabase query is used to uncover this type of knowledge.	ſ	1
	Hidden B) Shallow C) Deep D) Multi dimension	L	1
,	hich learning Requires self-assessment to identify patterns within data	ſ	1
		L	1
		г	1
	ferent learning model does not include Mamoriza B) Analogy (C) Deduction <b>D</b> ) Introduction	L	]
,	Memorize B) Analogy C) Deduction <b>D</b> ) Introduction	г	1
	box plot, the upper and lower whisker length depends on Median B) Mean C) IQR D) All of the above	[	]
<i>,</i>		r	1
	ucture representation of raw input data into meaningful is called a mode	[	]
-	Pattern B) Data C) Object D) Set of data	r	1
	nture selection tries to eliminate feature which are	[	]
· · · · · · · · · · · · · · · · · · ·	Rich B) Redundant C) Irrelevant D) B & C	г	1
30 For		[	]
	Interactive <b>B)Predictive</b> C)Descriptive D)Prescriptive	г	1
	e K- mean algorithm is a	l	J
	Supervised <b>B</b> ) Un Supervised C) Reinforcement D) Weakly Supervised	г	1
	Unsupervised learning we have model	Ĺ	]
,	Interactive B) catch, finally C) Descriptive D)Prescriptive	г	1
	edicting whether a tumor is malignant or benign is an example	[ .::::::::::::::::::::::::::::::::::::	]
<i>,</i>	Unsupervised B) Supervised regression C) Supervised D) Categorical Class	sincati	on 1
	hich of the following measures is not used for a classification model	l	J
,	Accuracy B) Precision C) Recall D) Purity	г	1
	nplest form of n-grams where $n=1$ is called	[	]
,	Bigram B) Unigram C) Trigram D) none of the above	г	1
	sine similarity is most popular used in	[	]
· · · · · · · · · · · · · · · · · · ·	Image Classification <b>B</b> ) <b>Text</b> C) Feature Selection D) None	1	
	ferent learning methods does not include [	]	
· · · · · · · · · · · · · · · · · · ·	Analogy <b>B)Introduction</b> C) Deduction D) Memorization		1
	hich of the following is a widely used and effective machine learning algorithm based of	on the 1	dea
	pagging	L	]
,	Decision Tree B) Regression C)Classification D)Random Forest	r	
	hich of the following techniques cannot be used for normalization in text mining	[	]
· · · · · · · · · · · · · · · · · · ·	Stemming B) Lemmatization C) Stop word Removal D)None		
40 As	the number of training examples goes to infinity, your model trained on that data will	have:	-
		l	]
A)I	Lower variance <b>B</b> ) <b>Higher variance</b> C) Same variance D) None of the above		

# Unit II

# **Nearest Neighboring Neural Network**

6	0		
1. What is the primary purpose of proximity	measures in data analysis	[	]
A) To calculate distances between	data points		
B) To visualize data distributions			
C) To identify patterns in data			
D) To classify data into categories			
2. What is the name of the proximity measured	e that calculates the straight-line distance	between	two
points		[	]
A) Euclidean distance	B) Manhattan distance		
C) Minkowski distance	D) Cosine similarity		
3. Which proximity measure is used to calcu	late the similarity between two vectors	[	]
A) Cosine similarity	B) Jaccard similarity		
C) Euclidean distance	D) Manhattan distance		
4. What is the name of the proximity measured	e that calculates the distance between two	points in	n a
grid-based system		[	]
A) Manhattan distance	B) Euclidean distance		
C) Minkowski distance	D) Chebyshev distance		
5. What is the name of the proximity measured	e that calculates the similarity between tw	o sets [	]
A) Jaccard similarity	B) Cosine similarity		
C) Euclidean distance	D) Manhattan distance		
6. Which proximity measure is used to calcu	late the distance between two points in a h	nigh-	
dimensional space	[	]	
A) Euclidean distance	B) Manhattan distance		
C) Minkowski distance	D) Mahalanobis distance		
7. What is the name of the proximity measured	e that calculates the distance between two	points b	ased
on their ranks		[	]
A) Spearman distance	B) Kendall distance		
C) Euclidean distance	D) Manhattan distance		
8. What is a non-metric similarity function		[	]
A) A function that satisfies the triang	le inequality		
B) A function that does not satisfy	the triangle inequality		
C) A function that is used for cluster	ing		
D) A function that is used for classifi	cation		
9. Which of the following is an example of a	non-metric similarity function	[	]
A) Euclidean distance	B) Cosine similarity		
C) Jaccard similarity	<b>D</b> ) Edit distance		
10. What is the primary characteristic of a ne	on-metric similarity function	[	]
A) It is symmetric	B) It is non-negative		
C) It does not satisfy the triangle in	<b>nequality</b> D) It is used for clustering		
11. Which of the following non-metric simil	arity functions is used for comparing strin	gs [	]
A) Edit distance B) Cosine sim	ilarity C) Jaccard similarity D) Euclidea	ın distan	ce
12. What is the name of the non-metric simil	larity function that is used for comparing s	ets [	]
A) Jaccard similarity	B) Cosine similarity		
C) Euclidean distance	D) Manhattan distance		

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13. Which of the following non-metric similarity fu		L	]
A) Graph edit distance	B) Cosine similarity		
C) Jaccard similarity	D) Euclidean distance	-	
14. What is the primary advantage of using non-me	etric similarity functions	L	
A) They are more efficient to compute			
B) They are more accurate for certain ty	pes of data		
C) They are more interpretable			
D) They are more robust to outliers			
15. Which of the following is a limitation of non-m	-	[	]
A) They are not symmetric	B) They are not non-negative		
C) They do not satisfy the triangle inequa			-
16. What is the name of the non-metric similarity f	unction that is used for comparing ima	ges [	]
A) Image edit distance	B) Cosine similarity		
C) Jaccard similarity	D) Euclidean distance		
17. What is the primary purpose of measuring prox	imity between binary patterns [	]	
A) To classify data into categories	B) To cluster data into groups		
C) To identify similarities between patter	<b>ms</b> D) To predict continuous out	come	s
18. Which of the following proximity measures is a	commonly used for binary patterns	[	]
A) Euclidean distance	B) Manhattan distance		
C) Jaccard similarity	D) Cosine similarity		
19. What is the name of the proximity measure that	t calculates the number of identical bits	s betv	veen
two binary patterns		[	]
A) Hamming distance	B) Jaccard similarity	_	_
C) Cosine similarity	D) Euclidean distance		
20. Which of the following proximity measures is s	,	a feat	ure in
a binary pattern	-	[	
A) Jaccard similarity	B) Cosine similarity	L	-
C) Euclidean distance	D) Manhattan distance		
21. What is the name of the proximity measure that		binar	v
patterns based on the number of shared feature		[	1
A) Jaccard similarity	B) Cosine similarity	L	1
C) Euclidean distance	D) Simple matching coefficient		
22. Which classification algorithm uses Euclidean		[	]
A) K-Nearest Neighbors (KNN)	B) Support Vector Machines (SVM)	L	L
C) Decision Trees	D) Random Forest		
23. What is the primary advantage of using K-Near	,	ſ	1
A) It can handle high-dimensional data	B) It is robust to outliers	L	J
C) It is simple to implement	D) It can handle non-linear relationsl	nin	
24 What is the name of the classification algorithm		пр	1
A) Quadratic Discriminant Analysis (QD		L voic (1	
• •			LDA)
C) K-Nearest Neighbors (KNN)	D) Support Vector Machines (SVM)	г	1
25. Which classification algorithm uses cosine sime		L	]
A) K-Nearest Neighbors (KNN)	B) Support Vector Machines (SVM)		
C) Naive Bayes	D) Rocchio algorithm	г	ı
26. Which distance measure is commonly used in S	••••••	L	]
A) Euclidean distance	B) Manhattan distance		
C) Minkowski distance	D) Kernel distance		

#### **R23** Course Code: 23CS0902 27. What is the name of the classification algorithm that uses Manhattan distance [ 1 A) K-Nearest Neighbors (KNN) B) Support Vector Machines (SVM) C) Decision Trees D) Random Forest 28. Which classification algorithm uses Minkowski distance [ ] A) K-Nearest Neighbors (KNN) B) Support Vector Machines (SVM) D) Random Forest C) Decision Trees 29. What is the primary disadvantage of using distance-based classification algorithms ] ſ A) They are sensitive to outliers B) They are computationally expensive C) They are not suitable for high-dimensional data **D**) All of the above 30. What is the primary advantage of using K-Nearest Neighbor (KNN) classifiers ] [ A) They can handle high-dimensional data B) They are robust to outliers C) They are simple to implement D) They can handle non-linear relationship 31. Which of the following is a characteristic of K-Nearest Neighbor (KNN) classifiers [] A) They are parametric models **B)** They are non-parametric models C) They are suitable for high-dimensional data D) They are robust to outliers 32. How does the larger value of K affect the performance of K-Nearest Neighbor (KNN) classifiers ſ ] **B**) leads to underfitting A) leads to overfitting C) leads to befitting D) none of the above 33. Which of the following is a limitation of K-Nearest Neighbor (KNN) classifiers 1 [ A) They are not suitable for high-dimensional data B) They are sensitive to outliers C) They are computationally expensive **D)** All of the above 34. What is the name of the technique used to select the optimal value of K in K-Nearest Neighbor (KNN) classifiers ſ 1 A) Cross-validation B) Bootstrapping C) Grid search D) Random search 35. What is the primary goal of K-Nearest Neighbor (KNN) regression 1 ſ A) To classify data into categories **B)** To predict continuous outcomes C) To cluster data into groups D) To identify patterns in data 36. Which of the following is a characteristic of K-Nearest Neighbor (KNN) regression ] ſ A) It is a parametric model **B)** It is a non-parametric model C) It is suitable for high-dimensional data D) It is robust to outlier 37. How does the value of K affect the performance of K-Nearest Neighbor (KNN) regression[ ] A) A small value of K leads to overfitting B) A large value of K leads to underfitting C) A small value of K leads to underfitting D) A large value of K leads to overfitting 38 What is the name of the metric that measures the proportion of variance in the dependent variable that is predictable from the independent variable(s) in a regression model 1 ſ A) R-squared B) Coefficient of Determination C) Mean Squared Error (MSE) D) Mean Absolute Error (MAE) 39. Which of the following metrics is sensitive to outliers in a regression model 1 ſ A) Mean Squared Error (MSE) B) Mean Absolute Error (MAE) C) Root Mean Squared Error (RMSE) D) All of the above 40. What is the name of the technique used to evaluate the performance of a regression model on unseen data [ ] A) Cross-validation B) Bootstrapping C) Holdout method D) All of the above

### Decision Trees and Bayes probability

1. What is a decision tree?	[	]
A) type of neural network B) tree-like m	odel used for classification & regression	n
C) type of clustering algorithm D) type	e of dimensionality reduction technique	
2. Which of the following is a characteristic	c of decision trees? [	]
A) They are parametric models <b>B</b> ) They	y are non-parametric models	
C) They are suitable for high-dimensional dat		
3. What is the primary advantage of using		1
A) They can handle non-linear relationships	B) They are robust to outliers	-
C) A & B	D) None of the above	
4. Which of the following is a type of decisi	·	]
A) ID3 B) C4.5 C) CAR		1
5. What is the name of the process used to	,	]
A) <b>Pruning</b> B) Trimming C) Redu		1
6. Which of the following is a benefit of usi		1
-	B) They are suitable for large datasets	1
· · · ·	D) All of the above	
7. What is the primary disadvantage of usi	,	1
	B) They are sensitive to outliers	1
	D) All of the above	
8. Which of the following is a type of split u	,	1
A) Binary split B) Multiway split		-
9. What is the name of the measure used to		
tree?	r	
	C) Information gain <b>D</b> ) <b>All of the abov</b>	<b>1</b>
10. What is the primary purpose of impuri	r in the sures in decision tree construction	л л
A) To evaluate the performance of the model	D) To calcot the best culit for a r	ا دامہ
A) To evaluate the performance of the model		oue
C) To prune the tree	D) To handle missing values	1
11. Which is the better method for fast co	-	]
	C) Variance D) All of the above	1
12. What is Gini impurity?	$\mathbf{P}$ $\mathbf{A}$ measure of the entropy in a	
A) A measure of the variance in a dataset $C$	, <b>1</b> ,	lataset
C) A measure of the probability of misclas		
D) A measure of the proportion of samples in	e e	
13. Which impurity measure is based on th		J
A) Gini impurity <b>B) Entropy</b> C) Varia	-	
14. What is the primary advantage of using	g impurity measures in decision tree	-
construction?	l	]
		a node
C) They help to handle missing values	<b>B</b> ) They help to select the best split for	
D) They help to improve the interpretability of	of the model	
<ul><li>D) They help to improve the interpretability of</li><li>15. What is the primary goal of regression</li></ul>	of the model based on decision trees? [	]
<ul><li>D) They help to improve the interpretability of 15. What is the primary goal of regression</li><li>A) To classify data into categories B) To p</li></ul>	of the model based on decision trees? [ redict continuous outcomes	]
<ul> <li>D) They help to improve the interpretability of 15. What is the primary goal of regression</li> <li>A) To classify data into categories B) To p</li> <li>C) To cluster data into groups D) To it</li> </ul>	of the model <b>based on decision trees?</b> [ redict continuous outcomes dentify patterns in data	]
<ul><li>D) They help to improve the interpretability of 15. What is the primary goal of regression</li><li>A) To classify data into categories B) To p</li></ul>	of the model <b>based on decision trees?</b> [ redict continuous outcomes dentify patterns in data	]

18. Which is not a common meth	of the target variable <b>and used to improve the performance</b>	e of regression	
rees?		[	]
A) Pruning <b>B</b> ).Clustering	C) Bagging D) Boosting	r >	
19. What is a random forest?		[	]
A) A single decision tree	B) An ensemble of decision trees		
C) A type of neural network	D) A clustering algorithm		
<b>20. Which of the following is a b</b>		[	]
A) Improved accuracy	B) Reduced overfitting		
C) Increased disorder	D) Extrapolate well		
1. How do random forests hand	6	[	]
A) By ignoring them	B) By imputing them		
C) By using surrogate splits	D) By using a combination of the		
	ge of using random forests for class	ification and	
egression?		[	]
A) They can handle high-dimens		0	
C) Computationally expensive	D) Wrong continuous natur	re prediction	-
3. What is Bayes' theorem used	•	L	]
· · · ·	B) To predict continuous outcome		
C) To cluster data into groups	D) To update probabilities based	l on new eviden	
4 Which of the following is a ty			]
		Neither A nor B	-
	tion made by the Naive Bayes classi	fier?	]
A) Features are independent	B) Features are dependent		
C) Features are normally distribute		ly distribute	
	enefit of using Bayes classifiers?	L	]
	onal data B) They are robust to outlie		
C) They can handle missing values		plement	-
7. What is the Bayes decision ru		L	]
A) Choose the class with the higher			
B) Choose the class with the high			
C) Choose the class with the lowes			
D) Choose the class with the lower		о г	-
	pe of probability used in Bayes' the		J
A) Prior probability B) Likelihoo		None of the above	e _
	e .	L	J
	ationships B) They are robust to outlie		
A) They can handle non-linear rela	- · ·		
<ul> <li>A) They can handle non-linear relation</li> <li>C) They can handle missing values</li> </ul>	s D) They are based on sound math	ematical princi	ple
<ul> <li>A) They can handle non-linear relation</li> <li>C) They can handle missing values</li> <li><b>0. Which of the following is an</b></li> </ul>	s D) They are based on sound math application of Bayes classifiers?	-	ple ]
<ul> <li>A) They can handle non-linear relations</li> <li>C) They can handle missing values</li> <li>C) Which of the following is an analysis of the following</li></ul>	<b>b D</b> ) They are based on sound math application of Bayes classifiers? mage classification <b>C</b> ) <b>A &amp; B</b>	D) None	ple ]
<ul> <li>A) They can handle non-linear related.</li> <li>C) They can handle missing values</li> <li>O. Which of the following is an atom (A) Text classification (B) In (A) Text of the following is a line (A) (B) (A) (B) (A) (B) (B) (B) (B) (B) (B) (B) (B) (B) (B</li></ul>	<b>b D</b> ) They are based on sound math application of Bayes classifiers? mage classification <b>C</b> ) <b>A &amp; B</b> <b>mitation of Bayes classifiers?</b>	[ D) None [	ple ] ]
<b>80. Which of the following is an</b> A) Text classificationB) In <b>81. Which of the following is a line</b>	<ul> <li>b) They are based on sound math application of Bayes classifiers?</li> <li>mage classification C) A &amp; B</li> <li>mitation of Bayes classifiers?</li> <li>f features B) They are sensitive to ou</li> </ul>	[ D) None [	ple ] ]

		_	_
<b>33.</b> What is the posterior probability	•	[	]
A) The probability of the hypothesis be	0		
B) The probability of the data given the			
C) The probability of the hypothesis	after observing the data		
D) The probability of the data			-
6	i i	[	]
A) Updating prior beliefs with new d	-		
B) Using prior beliefs to predict future			
C) Ignoring prior beliefs and relying so	-		
D) Using posterior beliefs to determine		г	1
<b>35.What is a key advantage of Bayes</b> A) It provides a fixed and certain answ		[	]
B) It allows for the incorporation of			
C) It is only suitable for simple problem			
D) It ignores uncertainty in the data			
	idered in terms of classification perform	anca? [	· 1
•	B) An optimal classifier	ance. [	
, 11	D) A random classifier		
37. Which of the following is a prope		[	]
A) It minimizes the expected risk B		L	1
· · · · · · · · · · · · · · · · · · ·	D) It relies solely on likelihood		
	the Bayes classifier is considered optima	d? [	1
A) It uses a complex decision boundary	•	-	
C) It ignores uncertainty in the data	D) It relies on a specific distr		n
39. Which of the following statement	· •	ſ	]
A) It is only suitable for binary classifi	•	-	-
B) It requires a large amount of trainin	g data		
C) It provides the lowest possible err	for rate		
D) It is computationally expensive			
40. What is the Bayes error rate?		[	]
A) The minimum possible error rate	for a classifier		
B) The maximum possible error rate for	or a classifier		
C) The average error rate for a classifie	er		
D) The error rate for a specific classifi	or		

D) The error rate for a specific classifier