

MACHINE LEARNING

In Q1 to Q11, only one option is correct, choose the correct option:

Least Square Error B) Maximum Lik	
C) Logarithmic Loss	D) Both A and B
Which of the following statement is true aA Linear regression is sensitive to outliersC) Can't say	bout outliers in linear regression? B) linear regression is not sensitive to outliers D) none of these
3. A line falls from left to right if a slope is _ A) Positive B Negative C) Zero	
4. Which of the following will have symmetry (A) Regression B) Correlation C) Both of	ric relation between dependent variable and independent variable? Them D) None of these
5. Which of the following is the reason for or A) High bias and high variance B) Low Low bias and high variance	
6. If output involves label then that model is A) Descriptive model B) Predictive mo	dal
C) Reinforcement learning	D) All of the above
7. Lasso and Ridge regression techniques below.A) Cross validation B) Removing outC) SMOTE	
8. To overcome with imbalance dataset which A) Cross validation B) Regularization C) Kernel	
 The AUC Receiver Operator Characteristic classification problems. It uses to m TRR and FPR B) Sensitivity and precision C) Sensitivity and Specificity 	
	(AUCROC) curve for the better model area under the curve should
11. Pick the feature extraction from below: A) P) Apply PCA to project high dimensional C) Removing stop words D) Forward selection	_
In Q12, more than one options are correct, choose	all the correct options:
12. Which of the following is true about Norm Regression?A) We don't have to choose the learning raB) It becomes slow when number of featurC) We need to iterate.D) It does not make use of dependent varia	res is very large.



MACHINE LEARNING

Q13 and Q15 are subjective answer type questions, Answer them briefly.

- 13. Explain the term regularization?
- 14. Which particular algorithms are used for regularization?
- 15. Explain the term error present in linear regression equation?

Answer 13. Regularization is a technique used in machine learning and statistical modelling to prevent overfitting by adding a penalty to the loss function. Overfitting occurs when a model learns the noise and details in the training data to the extent that it negatively impacts its performance on new, unseen data. A model that overfits is too complex and captures patterns that do not generalize to other datasets. Regularization is an essential technique in machine learning that helps to balance the trade- off between fitting the training data well and maintaining the model's ability to generalize to new data.

Answer 14. Regularization is a technique used in machine learning to prevent overfitting by adding a penalty to the loss function. Several algorithms and methods implement regularization. The most common types of regularization include:

- 1. **Lasso Regularization:** Lasso stands for Least Absolute Shrinkage and Selection Operator. It adds the absolute value of the coefficients as a penalty term to the loss function.
- 2. **Ridge Regularization:** Ridge regularization add the squared value of the coefficients as a penalty term to the loss function.
- 3. **Elastic Net Regularization:** Elastic Net combines both L1 and L2 regularization penalties.

- 4. **Dropout Regularization:** Commonly used in neural networks, dropout involves randomly setting a fraction of the neurons to zero during training.
- 5. **Early Stopping:** This technique involves monitoring the model's performance on a validation set during training and stopping the training process when performance starts to degrade.

Answer 15. In the context of linear regression, the term "error" (often denotes as (\epsilon) or (e)) refers to the difference between the observed values and the values predicted by the linear regression model. Linear regression aims to model the relationship between one or more independent variables (predictors) and a dependent variable (response) by fitting a linear equation to the observed data.

Linear Regression Model:

The basic form of a linear regression equation is:

```
[
Y=\beta_0 + \beta_2*_2 + \Idots + \beta_n*n + \epsilon
]
```

Where:

- (y) is the dependent variable (the output we are trying to predict).
- (\beta_0) is the intercept of the regression line (the value of (y) when all (x) values are zero).
- (\beta_1, \beta_2, \Idots, \beta_n) are the coefficients of the independent variables (x_1, x_2, \Idots, x_n).
- (\epsilon) is the error term.