

## MACHINE LEARNING

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

1.	Which of the following methods do we use to fi A) Least Square Error B) Maximum Likeliho C) Logarithmic Loss	ind the best fit line for data in Linear Regression?  od  D) Both A and B
2.	Which of the following statement is true about (A) Linear regression is sensitive to outliers B) li C) Can't say	_
3.	A line falls from left to right if a slope is  A) Positive  B) Negative C) Zero  D) Use	
4.	Which of the following will have symmetric rel A) Regression B) Correlation C) Both of them	ation between dependent variable and independent variable?  D) None of these
5.	Which of the following is the reason for over fit A) High bias and high variance B) Low bias at C) Low bias and high variance	<del>-</del>
6.	If output involves label then that model is called A) Descriptive model  B) Predictive modal C) Reinforcement learning	d as:  D) All of the above
7.	Lasso and Ridge regression techniques belong t A) Cross validation B) Removing outliers C) SMOTE	D) Regularization
8.	To overcome with imbalance dataset which tech A) Cross validation B) Regularization C) Kernel	hnique can be used?  D) SMOTE
9.	The AUC Receiver Operator Characteristic (AU classification problems. It uses to make g A) TPR and FPR B) Sensitivity and pred C) Sensitivity and Specificity	
10	<ul><li>In AUC Receiver Operator Characteristic (AUC be less.</li><li>A) True</li><li>B) False</li></ul>	CROC) curve for the better model area under the curve should
11	Pick the feature extraction from below: A) ConB) Apply PCA to project high dimensional data C) Removing stop words D) Forward selection	
In Q12	, more than one options are correct, choose all th	e correct options:
12	Regression?  A) We don't have to choose the learning rate.  B) It becomes slow when number of features is C) We need to iterate.	quation used to compute the coefficient of the Linear very large.
	D) It does not make use of dependent variable.	



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

13. Explain the term regularization?

Regularisation is the name given to a set of machine learning approaches designed to avoid overfitting and enhance a model's ability. When a model performs well on training data but fails to generalise to new data, it is said to be overfit. Regularisation encourages the model to have fewer parameter values or more straightforward representations by including a penalty term to the objective function. Complex or severe parameter values that could result in overfitting are discouraged by this penalty. During training, the regularisation term is often added to the loss function as a function of the model's parameters.

14. Which particular algorithms are used for regularization?

Different machine learning algorithms can use regularisation to avoid overfitting. The following are a few examples of frequently used regularisation techniques-based algorithms:

- 1. Ridge Regression (Linear Regression with L2 Regularization):
  The OLS objective function of linear regression is modified by the L2 penalty factor introduced by ridge regression. It promotes small model coefficients, which minimises overfitting.
- Lasso Regression (Linear Regression with L1 Regularisation): Lasso regression modifies the OLS objective f
  unction by include an L1 penalty component. By bringing some of the coefficients to absolutely zero, it promo
  tes sparsity by performing feature selection.
- 3. Elastic Net Regression: In linear regression, elastic net combines L1 and L2 regularisation penalties. It strikes a balance between L1 regularization's capacity to increase sparsity and L2 regularization's capacity to reduce parameter sizes.
- Logistic Regression with L1 or L2 Regularisation: To avoid overfitting in binary classification situations, logistic regression can also include L1 or L2 regularisation.
   In order to get the best coefficients, regularised logistic regression penalises high parameter values.
- 5. Regularised Support Vector Machines (SVM): Regularisation can be incorporated into SVM algorithms by using the C parameter. The tradeoff between increasing the margin and reducing the classification error is controlled by the C parameter. A higher C value decreases regularisation and raises the possibility of overfitting.
- 6. Neural Networks with Regularization Techniques: To avoid overfitting, regularisation techniques can be used with neural networks. L1 and L2 regularisation (weight decay), dropout regularisation, and early stopping are frequently used regularisation techniques for neural networks.
- 15. Explain the term error present in linear regression equation?

The difference between the actual observed values and the anticipated values derived from the linear regression equation is referred to as the "error" in linear regression. The linear regression equation is used to model the relationship between a dependent variable and one or more independent variables.