

MACHINE LEARNING

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



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

13. Explain the term regularization?

Answer: Regularization is a technique used in machine learning to prevent overfitting and improve generalization performance. Regularization introduce additional constraints or penalties to model's objective

Function, which helps control the complexity of the model and prevent it from becoming too sensitive to the training data. The goal is to find a balance between fitting the training data well and maintaining good generalization, reducing overfitting and improving the model's ability to generalize to unseen data.

Several types of regularization techniques used some are:

- 1.Lasso
- 2. Ridge
- 3.elastic net
- 4.Dropout
- 5. Early stopping
- 14. Which particular algorithms are used for regularization?

Answer: Several algorithms and regularization techniques can be used depending on the specific problem and the nature of data. Some commonly used algorithms are:

- 1. Linear Regression: This models to prevent overfitting and control magnitude of the coefficients.
- 2. **Logistic regression**: Logistic Regression model can benefit from linear regression to avoid overfitting and improve generalization.
- 3. **Support vector Machines:** SVM models can be regularized using techniques such as Ridge regularization to control the complexity of the decision boundary.
- 4.**Decision Trees**: Decision trees can be regularized using techniques like pruning, where branches of tree are removed based on certain criteria to prevent overfitting.
- 5.**Random forests:** Random forests, which are an ensemble of decision trees can benefit from regularization techniques like feature subsampling and bootstrap bagging to improve generalization.
- 15. Explain the term error present in linear regression equation?

Answer: In linear regression, the term "error" refers to the difference between the actual observed values and the predicted values obtained from the linear regression equation. It represents the unexplained or residual variation in the dependent variable that the linear regression model cannot account for.

The linear regression equation is typically represented as:

$$Y=B_0+B_1X_1+B_2X_2....+\epsilon$$

Where:

- -y is the dependent variable or the target variable being predicted.
- -X₁X_{2...} are independent variable or features used to predict the dependent variable.
- $-B_0B_1$ Are the coefficients or weights associated with each independent variable.
- -ε represents the error term or residual.