

MACHINE LEARNING

MCQ

ANSWER 1= A) Least Square Error

ANSWER 2= A) Linear regression is sensitive to outliers

ANSWER 3= B) Negative

ANSWER 4= B) Correlation

ANSWER 5= C) Low bias and high variance

ANSWER 6= B) Predictive model

ANSWER 7= A) Cross validation

ANSWER 8= D) SMOTE

ANSWER 9= A) TPR and FPR

ANSWER 10=B) False

ANSWER 11=B) Apply PCA to project high dimensional data

ANSWER 12=A,B,AND C A) We don't have to choose the learning rate

B) It becomes slow when number of features is very large

C) We need to iterate

SUBJECTIVE ANSWER 13 TO 15

ANSWER 13= Regularisation, meaning in the machine learning context, refers to minimising or shrinking the coefficient estimates towards zero to avoid underfitting or overfitting the machine learning model.

There are three main regularisation techniques: L2(Ridge) regularisation, L1(Lasso) regularisation, and Dropout regularisation.

1) Ridge Regularization (L2 Regularization)

Ridge Regularization is also known as L2 regularisation or ridge regression. It works by adding a penalty or complexity term in the cost function of the standard least squares model.

2) Lasso Regularization (L1 Regularization)

Lasso regularisation, also known as L1 regularisation or lasso regression, also works by minimising the weights by including them as a penalty in the cost function.

3) Dropout Regularisation

Besides ridge and lasso regression, there are other less popular regularisation techniques, such as dropout regularisation. Dropout regularisation is used in neural network and Elastic Net regularisation, and it combines the penalty function of L1 and L2 regularisation.

ANSWER 14=Regularization refers to a range of strategies for regularising learning from specific characteristics in classical algorithms or neurons in neural network algorithms. It normalises and moderates weights associated with a feature or a neuron so that algorithms aren't reliant on a small number of features or neurons to predict the outcome.

Ridge regression – Its purpose is to overcome problems such as data overfitting and multicollinearity in data. When there is considerable collinearity (the existence of near-linear connections among the independent variables) among the feature variables, a typical linear or polynomial regression model will fail.

ANSWER 15=The error term, also known as the residual, in a regression represents the difference between the observed value of the dependent variable and the predicted value of the dependent variable. It measures the unexplained variation in the dependent variable that is not accounted for by the independent variables in the model. In other words, it represents the noise or randomness in the data that cannot be explained by the model.