

ASSIGNMENT – 39

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 find the best fit line for data in Linear Regression?

- A) Least Square Error B) Maximum Likelihood
- C) Logarithmic Loss D) Both A and B

ANS- A) Least Square Error

2. Which of the following statement is true about outliers in linear regression?

- A) Linear regression is sensitive to outliers B) linear regression is not sensitive to outliers
- C) Can't say D) none of these

ANS- A) Linear regression is sensitive to outliers

3. A line falls from left to right if a slope is _____?

- A) Positive B) Negative
- C) Zero D) Undefined

ANS- B) Negative

4. Which of the following will have symmetric relation between dependent variable and independent variable?

- A) Regression B) Correlation
- C) Both of them D) None of these

ANS- B) Correlation

5. Which of the following is the reason for over fitting condition?

- A) High bias and high variance B) Low bias and low variance
- C) Low bias and high variance D) none of these

ANS- C) Low bias and high variance

6. If output involves label then that model is called as:

- A) Descriptive model B) Predictive modal
- C) Reinforcement learning D) All of the above

ANS- B) Predictive modal

7. Lasso and Ridge regression techniques belong to _____?

- A) Cross validation B) Removing outliers
- C) SMOTE D) Regularization

ANS- D) Regularization

8. To overcome with imbalance dataset which technique can be used?

- A) Cross validation B) Regularization
- C) Kernel D) SMOTE

ANS- D) SMOTE

9. The AUC Receiver Operator Characteristic (AUCROC) curve is an evaluation metric for binary classification problems. It uses _____ to make graph?

- A) TPR and FPR
- B) Sensitivity and precision
- C) Sensitivity and Specificity
- D) Recall and precision

ANS- C) Sensitivity and Specificity

10. In AUC Receiver Operator Characteristic (AUCROC) curve for the better model area under the curve should be less.

- A) True
- B) False

ANS- B) False

11. Pick the feature extraction from below:

- A) Construction bag of words from a email
- B) Apply PCA to project high dimensional data
- C) Removing stop words
- D) Forward selection

ANS- A) Construction bag of words from a email

In Q12, more than one options are correct, choose all the correct options:

12. Which of the following is true about Normal Equation used to compute the coefficient of the Linear Regression?

- A)
- D) It does not make use of dependent variable.

ANS- 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.

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

13. Explain the term regularization?

ANS-

Regularization refers to techniques that are used to calibrate machine learning models in order to minimize the adjusted loss function and prevent overfitting or underfitting.

This is a form of regression, that constrains/ regularizes or shrinks the coefficient estimates towards zero. In other words, this technique discourages learning a more complex or flexible model, so as to avoid the risk of overfitting.

A simple relation for linear regression looks like this. Here Y represents the learned relation and β represents the coefficient estimates for different variables or predictors(X).

The fitting procedure involves a loss function, known as residual sum of squares or RSS. The coefficients are chosen, such that they minimize this loss function.

14. Which particular algorithms are used for regularization?

ANS-

Ridge Regression (L2 Regularization)

Ridge regression is also called L2 norm or regularization.

When using this technique, we add the sum of weight's square to a loss function and thus create a new loss function which is denoted thus:

$$\text{Loss} = \sum_{j=1}^m \left(Y_i - W_0 - \sum_{i=1}^n W_i X_{ji} \right)^2 + \lambda \sum_{i=1}^n W_i^2$$

As seen above, the original loss function is modified by adding normalized weights. Here normalized weights are in the form of squares.

You may have noticed parameters λ along with normalized weights. λ is the parameter that needs to be tuned using a cross-validation dataset. When you use $\lambda=0$, it returns the residual sum of square as loss function which you chose initially. For a very high value of λ , loss will ignore core loss function and minimize weight's square and will end up taking the parameters' value as zero.

Now the parameters are learned using a modified loss function. To minimize the above function, parameters need to be as small as possible. Thus, L2 norm prevents weights from rising too high.

Lasso Regression (L1 Regularization)

Also called lasso regression and denoted as below:

$$\text{Loss} = \sum_{j=1}^m \left(Y_i - W_0 - \sum_{i=1}^n W_i X_{ji} \right)^2 + \lambda \sum_{i=1}^n |W_i|$$

This technique is different from ridge regression as it uses absolute weight values for normalization. λ is again a tuning parameter and behaves in the same as it does when using ridge regression.

As loss function only considers absolute weights, optimization algorithms penalize higher weight values.

In ridge regression, loss function along with the optimization algorithm brings parameters near to zero but not actually zero, while lasso eliminates less important features and sets respective weight values to zero. Thus, lasso also performs feature selection along with regularization.

15. Explain the term error present in linear regression equation?

ANS-

The error term is the stuff that isn't explained by the model.

For a very simple example, suppose you are predicting the weight of adult human males based on their height. Well, height is certainly related to weight - taller people tend to be heavier - but the model won't be perfect because there is a range of weights at each height. The error is the difference between the predicted value and the actual value.