 ASSIGNMENT - 6

# MACHINE LEARNING

## In Q1 to Q5, only one option is correct, Choose the correct option:

1. In which of the following you can say that the model is overfitting?
   * 1. High R-squared value for train-set and High R-squared value for test-set.
     2. Low R-squared value for train-set and High R-squared value for test-set.
     3. High R-squared value for train-set and Low R-squared value for test-set.
     4. None of the above
2. Which among the following is a disadvantage of decision trees?

A) Decision trees are prone to outliers.

B) Decision trees are highly prone to over fitting.

C) Decision trees are not easy to interpret

D) None of the above.

1. Which of the following is an ensemble technique?

A) SVM B) Logistic Regression

C) Random Forest D) Decision tree

1. Suppose you are building a classification model for detection of a fatal disease where detection of the disease is most important. In this case which of the following metrics you would focus on?

A) Accuracy B) Sensitivity

C) Precision D) None of the above.

1. The value of AUC (Area under Curve) value for ROC curve of model A is 0.70 and of model B is

0.85. Which of these two models is doing better job in classification?

A) Model A B) Model B

C) both are performing equal D) Data Insufficient

## In Q6 to Q9, more than one options are correct, Choose all the correct options:

1. Which of the following are the regularization technique in Linear Regression??
   1. Ridge B) R-squared

C) MSE D) Lasso

1. Which of the following is not an example of boosting technique?
   1. Adaboost B) Decision Tree

C) Random Forest D) Xgboost.

1. Which of the techniques are used for regularization of Decision Trees?
   1. Pruning B) L2 regularization

C) Restricting the max depth of the tree D) All of the above

1. Which of the following statements is true regarding the Adaboost technique?
   1. We initialize the probabilities of the distribution as 1/n, where n is the number of data-points

B) A tree in the ensemble focuses more on the data points on which the previous tree was not performing well

* 1. It is example of bagging technique
  2. None of the above

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## Q10 to Q15 are subjective answer type questions, Answer them briefly.

1. Explain how does the adjusted R-squared penalize the presence of unnecessary predictors in the model?

Ans: The adjusted R-squared compensates for the addition of variables and **only increases if the new predictor enhances the model above what would be obtained by probability**. Conversely, it will decrease when a predictor improves the model less than what is predicted by chance.

# Differentiate between Ridge and Lasso Regression.

Ans:

**L2 Ridge Regression**  
It is a [**Regularization Method**](https://en.wikipedia.org/wiki/Regularization_(mathematics)) to reduce [**Overfitting**](https://en.wikipedia.org/wiki/Overfitting).  
We try to use a trend line that overfit the training data, and so, it has much higher variance then the OLS. The main idea of Ridge Regression is to fit a new line that doesn’t fit the training data. In other words, we introduce a certain **Amount on Bias** into the new trend line.

**L1 Lasso Regression**  
It is a [**Regularization Method**](https://en.wikipedia.org/wiki/Regularization_(mathematics)) to reduce [**Overfitting**](https://en.wikipedia.org/wiki/Overfitting).  
It is similar to RIDGE REGRESSION except to a very important difference: the **Penalty Function** now is: lambda\*|slope|.

The result of the Lasso Regression is very similar to the Result given by the Ridge Regression. Both can be used in Logistic Regression, Regression with discrete values and Regression with interaction. The big difference between Rdge and Lassp start to be clear when we **Increase the value on Lambda**. In fact, **Ridge** can only shrink the slope **asynmtotically** close to **zero**, while **Lasso** can shrink the slope **all the way to zero**. The advantage of this is clear when we have lots of parameters in the model.

1. What is VIF? What is the suitable value of a VIF for a feature to be included in a regression modelling?

Ans:

A variance inflation factor (VIF) is **a measure of the amount of multicollinearity in regression analysis**. Multicollinearity exists when there is a correlation between multiple independent variables in a multiple regression model. This can adversely affect the regression results.

Generally, a VIF above 4 or tolerance below 0.25 indicates that multicollinearity might exist, and further investigation is required. When VIF is higher than 10 or tolerance is lower than 0.1, there is significant multicollinearity that needs to be corrected.

1. Why do we need to scale the data before feeding it to the train the model?

Ans:

**To ensure that the gradient descent moves smoothly towards the minima and that the steps for gradient descent are updated at the same rate for all the features**, we scale the data before feeding it to the model.

1. What are the different metrics which are used to check the goodness of fit in linear regression?

Ans:

Three statistics are used in Ordinary Least Squares (OLS) regression to evaluate model fit: **R-squared, the overall F-test, and the Root Mean Square Error (RMSE)**.

1. From the following confusion matrix calculate sensitivity, specificity, precision, recall and accuracy.

|  |  |  |
| --- | --- | --- |
| Actual/Predicted | True | False |
| True | 1000 | 50 |
| False | 250 | 1200 |

Ans: TP = 1000, TN = 1200, FP= 50, FN= 250

**Accuracy** = TP + TN / TP + TN + FP + FN = 1000+1200/1000+50+250+1200 = **0.88**

**Precision** = TP / TP + FP = 1000/1000+50= **0.952**

**Recall** = TP/TP+FN = 1000/1000+250= **0.8**

**Specificity** = TN/TN+FP = 1200/1200+50= **0.96**