## **Assignment-based Subjective Questions**

1. From your analysis of the categorical variables from the dataset, what could you infer about their effect on the dependent variable? (3 marks)

Ans. In the bike sharing dataset, let's consider the effect of the categorical variable 'weathersit' on the target variable 'cnt'. While performing EDA, I visualized the relationship between the categorical variables and the target variable. It was seen that during the weather situation 1 (Clear, few clouds, partly cloudy, a high number of bike rentals were made, with the median being 50,000 approximately. Similarly, certain inferences could be made 'season' and 'yr' as well. Also, during model building on inclusion of categorical features such as yr, season etc we saw a significant growth in the value of R-squared and adjusted R-squared. This implies that the categorical features were helpful in explaining a greater proportion of variance in the dataset.

2. Why is it important to use drop first=True during dummy variable creation? (2 mark)

Ans. drop first=True is important to use, as it helps in reducing the extra column created during dummy variable creation. Hence it reduces the correlations created among dummy variables.

- 3. Looking at the pair-plot among the numerical variables, which one has the highest correlation with the target variable? (1 mark)
  - Ans. The numerical variable 'registered' has the highest correlation with the target variable 'cnt', if we consider all the features. But after data preparation, when we drop registered due to multicollinearity the numerical variable 'atemp' has the highest correlation with the target variable 'cnt'.
- 4. 4. How did you validate the assumptions of Linear Regression after building the model on the training set? (3 marks)

Ans.

- There should be a linear and additive relationship between dependent (response) variable and independent (predictor) variable(s). A linear relationship suggests that a change in response Y due to one unit change in X¹ is constant, regardless of the value of X¹. An additive relationship suggests that the effect of X¹ on Y is independent of other variables.
- There should be no correlation between the residual (error) terms. Absence of this phenomenon is known as Autocorrelation.
- The independent variables should not be correlated. Absence of this phenomenon is known as multicollinearity.
- The error terms must have constant variance. This phenomenon is known as homoskedasticity. The
- presence of non-constant variance is referred to heteroskedasticity.
- The error terms must be normally distributed.
- 5. Based on the final model, which are the top 3 features contributing significantly towards explaining the demand of the shared bikes? (2 marks)

- We can see that temperature variable is having the highest coefficient 0.4914, which means if the temperature increases by one unit the number of bike rentals increases by 0.4914 units.
- We also see there are some variables with negative coefficients. A negative coefficient suggests that as the independent variable increases, the dependent variable tends to decrease.
- We have spring, mist cloudy, light snow variables with negative coefficient.
  The coefficient value signifies how much of the dependent variable changes
  given a one-unit shift in the independent variable while holding other variables
  in the model constant.

## **General Subjective Questions**

Explain the linear regression algorithm in detail. (4 marks)
 Ans.

Linear regression is one of the very basic forms of machine learning where we train a model to predict the behaviour of your data based on some variables. In the case of linear regression as you can see the name suggests linear that means the two variables which are on the x-axis and y-axis should be linearly correlated.

Regression analysis is used for three types of applications:

- 1. Finding out the effect of Input variables on Target variable.
- 2. Finding out the change in Target variable with respect to one or more input variable.
- 3. To find out upcoming trends.

Here are the types of regressions:

- 1. Linear Regression
- 2. Multiple Linear Regression
- 3. Logistic Regression
- 4. Polynomial Regression.
- 2. 2. Explain the Anscombe's quartet in detail.

Ans. Anscombe's Quartet can be defined as a group of four data sets which are nearly identical in simple descriptive statistics, but there are some peculiarities in the dataset that fools the regression model if built. They have very different distributions and appear differently when plotted on scatter plots.

## The four data set are

- a) Dataset 1: -fits the linear regression model pretty well
- b) Dataset 2: could not fit linear regression model on the data quite well as the data is non-linear.

- c) Dataset 3: shows the outliers involved in the dataset which cannot be handled by linear regression model
- d) Dataset 4: shows the outliers involved in the dataset which cannot be handled by linear regression model.
- 3. What is Pearson's R?

Ans. Pearson's r is a numerical summary of the strength of the linear association between the variables. If the variables tend to go up and down together, the correlation coefficient will be positive. If the variables tend to go up and down in opposition with low values of one variable associated with high values of the other, the correlation coefficient will be negative.

4. What is scaling? Why is scaling performed? What is the difference between normalized scaling and standardized scaling?

Ans. It is a step of data Pre-Processing which is applied to independent variables to normalize the data within a particular range. It also helps in speeding up the calculations in an algorithm.

Most of the times, collected data set contains features highly varying in magnitudes, units and range. If scaling is not done then algorithm only takes magnitude in account and not units hence incorrect modelling. To solve this issue, we have to do scaling to bring all the variables to the same level of magnitude.

Normalisation typically means rescales the values into a range of

[0,1]. Standardization typically means rescales data to have a mean of 0 and a standard deviation of 1 (unit variance).

5. You might have observed that sometimes the value of VIF is infinite. Why does this happen? (3 marks)

Ans. If there is perfect correlation, then VIF = infinity. This shows a perfect correlation between two independent variables. In the case of perfect correlation, we get R2 =1, which lead to 1/(1-R2) infinity. To solve this problem, we need to drop one of the variables from the dataset which is causing this perfect multicollinearity. An infinite VIF value indicates that the corresponding variable may be expressed exactly by a linear combination of other variables (which show an infinite VIF as well).

6. What is a Q-Q plot? Explain the use and importance of a Q-Q plot in linear regression. (3 marks)

Ans. Q-Q Plots (Quantile-Quantile plots) are plots of two quantiles against each other. A quantile is a fraction where certain values fall below that quantile. For example, the median is a quantile where 50% of the data fall below that point and 50% lie above it. The purpose of Q Q plots is to find out if two sets of data come from the same distribution. A 45-degree angle is plotted on the Q Q plot; if the two data sets come from a common distribution, the points will fall on that reference line.

If the two distributions being compared are similar, the points in the Q-Q plot will approximately lie on the line y=x. If the distributions are linearly related, the points in the Q-Q plot will approximately lie on a line, but not necessarily on the line y=x. Q-Q plots can also be used as a graphical means of estimating parameters in a location-scale family of distributions.

A Q—Q plot is used to compare the shapes of distributions, providing a graphical view of how properties such as location, scale, and skewness are similar or different in the two distributions.