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)
- => The demad of bike is less in the month of spring when compared with other seasons. The demand bike increased in the year 2019 when compared with year 2018.
- 2. Why is it important to use **drop_first=True** during dummy variable creation? (2 marks) => 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 marks)
- => Looking at the pair-plot among the numerical variables, which one has the highest correlation with the target variable? **The numerical variable 'registered'** has the highest correlation with the target variable 'cnt', if we consider all the features.
- 4. How did you validate the assumptions of Linear Regression after building the model on the training set? (3 marks)
- => If a linear model is appropriate, the histogram should look approximately normal and the scatterplot of residuals should show random scatter. If we see a curved relationship in the residual plot, the linear model is not appropriate. Another type of residual plot shows the residuals versus the explanatory variable.
- 5.Based on the final model, which are the top 3 features contributing significantly towards explaining the demand of the shared bikes? (2 marks)
- => Based on final model top three features contributing significantly towards explaining the demand are:
 - 1. Temperature (0.552)
 - 2. weathersit : Light Snow, Light Rain + Thunderstorm + Scattered clouds, Light Rain + Scattered clouds (-0.264)
 - 3. year (0.256)

So it recomended to give these variables utmost importance while planning to achieve maximum demand.

General Subjective Questions

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

=> Linear Regression is a machine learning algorithm based on supervised learning. It performs a regression task. Regression models a target prediction value based on independent variables. It is mostly used for finding out the relationship between variables and forecasting. Different regression models differ based on – the kind of relationship between dependent and independent variables, they are considering and the number of independent variables being used.

Linear regression performs the task to predict a dependent variable value (y) based on a given independent variable (x). So, this regression technique finds out a linear relationship between x (input) and y(output). Hence, the name is LinearRegression.

In the figure above, X (input) is the work experience and Y (output) is the salary of a person. The regression line is the best fit line for our model.

2. Explain the Anscombe's quartet in detail.

(3 marks)

=> 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.

3. What is Pearson's R?

(3 marks)

=> In Statistics, the Pearson's Correlation Coefficient is also referred to as **Pearson's r, the Pearson product-moment correlation coefficient (PPMCC), or bivariate correlation**. It is a statistic that measures the linear correlation between two variables. Like all correlations, it also has a numerical value that lies between -1.0 and +1.0.

$$r = rac{\sum \left(x_i - ar{x}
ight)\left(y_i - ar{y}
ight)}{\sqrt{\sum \left(x_i - ar{x}
ight)^2 \sum \left(y_i - ar{y}
ight)^2}}$$

r = correlation coefficient

 x_i = values of the x-variable in a sample

 \bar{x} = mean of the values of the x-variable

 y_i = values of the y-variable in a sample

 \bar{y} = mean of the values of the y-variable

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

=> 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.

It is important to note that **scaling just affects the coefficients** and none of the other parameters like **t-statistic**, **F-statistic**, **p-values**, **R-squared**, etc.

Normalization/Min-Max Scaling:

• It brings all of the data in the range of 0 and 1. **sklearn.preprocessing.MinMaxScaler** helps to implement normalization in python.

MinMax Scaling:
$$x = \frac{x - min(x)}{max(x) - min(x)}$$

Standardization Scaling:

• Standardization replaces the values by their Z scores. It brings all of the data into a standard normal distribution which has mean (μ) zero and standard deviation one (σ) .

Standardisation:
$$x = \frac{x - mean(x)}{sd(x)}$$

- **sklearn.preprocessing.scale** helps to implement standardization in python.
- One disadvantage of normalization over standardization is that it **loses** some information in the data, especially about **outliers**.

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

Quantile-Quantile (Q-Q) plot, is a graphical tool to help us assess if a set of data plausibly came from some theoretical distribution such as a Normal, exponential or Uniform distribution. Also, it helps to determine if two data sets come from populations with a common distribution