#### **Assignment-based Subjective Questions**

**Question 1**. From your analysis of the categorical variables from the dataset, what could you infer about their effect on the dependent variable? (Do not edit)

Total Marks: 3 marks (Do not edit)

Answer: <Your answer for Question 1 goes below this line> (Do not edit)

Based on our analysis, categorical variables showed significant effects on bike rentals:

- 1. Season: Highest demand in Summer and Fall, lowest in Winter
- 2. Year: 2019 showed ~23% higher demand than 2018, indicating growing adoption
- 3. Weather: Clear weather (category 1) showed 30% higher rentals compared to adverse weather conditions
- 4. Working Day: Working days showed more consistent rental patterns compared to holidays/weekends

These variables help explain the seasonal and temporal patterns in bike rental demand.

**Question 2.** Why is it important to use **drop\_first=True** during dummy variable creation? (Do not edit)

Total Marks: 2 marks (Do not edit)

**Answer:** <Your answer for Question 2 goes below this line> (Do not edit)

Using drop first=True is important for two reasons:

- 1. Avoids the dummy variable trap (perfect multicollinearity) by removing one category as the reference level
- 2. Prevents redundant information since n-1 dummy variables can fully represent n categories
- 3. Improves model stability and interpretation as coefficients represent the effect relative to the reference category

**Question 3.** Looking at the pair-plot among the numerical variables, which one has the highest correlation with the target variable? (Do not edit)

Total Marks: 1 mark (Do not edit)

**Answer:** <Your answer for Question 3 goes below this line> (Do not edit)

Temperature ('temp') showed the highest correlation with the target variable 'cnt' with a correlation coefficient of 0.62. This indicates that temperature is the strongest numerical predictor of bike rental demand.

**Question 4.** How did you validate the assumptions of Linear Regression after building the model on the training set? (Do not edit)

Total Marks: 3 marks (Do not edit)

**Answer:** <Your answer for Question 4 goes below this line> (Do not edit)

We validated the linear regression assumptions through:

- 1. Linearity: Residual plot showed no systematic patterns, confirming linear relationships
- 2. Normality: Q-Q plot demonstrated residuals following approximately normal distribution
- 3. Homoscedasticity: Plot of residuals vs predicted values showed relatively constant variance
- 4. Independence: Time-based plot of residuals showed no significant autocorrelation

5. Multicollinearity: VIF analysis on predictor variables ensured no severe multicollinearity

**Question 5.** Based on the final model, which are the top 3 features contributing significantly towards explaining the demand of the shared bikes? (Do not edit)

Total Marks: 2 marks (Do not edit)

**Answer:** <Your answer for Question 5 goes below this line> (Do not edit)

The top 3 features with highest absolute coefficients were:

- 1. Temperature (0.5 coefficient): Strongest positive impact on demand
- 2. Year\_2019 (0.23 coefficient): Significant year-over-year growth
- 3. Clear Weather (0.19 coefficient): Strong positive impact during good weather conditions These features showed the most substantial and statistically significant effects on bike rental demand.

# **General Subjective Questions**

Question 6. Explain the linear regression algorithm in detail. (Do not edit)

Total Marks: 4 marks (Do not edit)

**Answer:** Please write your answer below this line. (Do not edit)

<Your answer for Question 6 goes here>

Linear regression models the relationship between a dependent variable (Y) and one or more independent variables (X) using the equation:

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + ... + \beta_n X_n + \varepsilon$$

#### Where:

- $\beta_0$  is the intercept
- $\beta_i$  are coefficients representing the change in Y for one unit change in  $X_i$
- $\epsilon$  is the error term

#### The algorithm:

- 1. Uses Ordinary Least Squares (OLS) to minimize the sum of squared residuals
- 2. Finds optimal coefficients that best fit the training data
- 3. Makes predictions using the linear combination of features and their coefficients

Question 7. Explain the Anscombe's quartet in detail. (Do not edit)

Total Marks: 3 marks (Do not edit)

Answer: Please write your answer below this line. (Do not edit)

<Your answer for Question 7 goes here>

Anscombe's quartet consists of four datasets that have nearly identical statistical properties (mean, variance, correlation, linear regression line) but look very different when plotted. Key points:

- 1. Demonstrates importance of visualizing data before analysis
- 2. Shows limitations of summary statistics alone
- 3. Illustrates why checking model assumptions is crucial
- 4. Each dataset has:
  - Same mean of x and y
  - Same variance of x and y
  - Same correlation coefficient
  - Same linear regression line

But they represent very different relationships, including linear, nonlinear, and outlier cases.

Question 8. What is Pearson's R? (Do not edit)

Total Marks: 3 marks (Do not edit)

Answer: Please write your answer below this line. (Do not edit)

<Your answer for Question 8 goes here>

# Pearson's R (correlation coefficient) measures the strength and direction of linear relationship between two variables:

- 1. Ranges from -1 to +1
  - +1 indicates perfect positive correlation
  - -1 indicates perfect negative correlation
  - 0 indicates no linear correlation
- 2. Properties:
  - Scale-invariant
  - Symmetric
  - Only measures linear relationships
- 3. Formula:  $r = cov(X,Y)/(\sigma x^*\sigma y)$

Where cov is covariance and  $\sigma$  is standard deviation

**Question 9.** What is scaling? Why is scaling performed? What is the difference between normalized scaling and standardized scaling? (Do not edit)

Total Marks: 3 marks (Do not edit)

Answer: Please write your answer below this line. (Do not edit)

<Your answer for Question 9 goes here>

# Scaling transforms features to a similar range:

- 1. Why Scaling is Important:
  - Prevents features with larger ranges from dominating
  - Improves convergence of gradient descent
  - Makes features comparable

- 2. Normalized Scaling (Min-Max):
  - Scales features to [0,1] range
  - Formula: X\_norm = (X X\_min)/(X\_max X\_min)
  - Preserves zero values
  - Better with non-normal distributions
- 3. Standardized Scaling (Z-score):
  - Transforms to mean=0, std=1
  - Formula:  $X_std = (X \mu)/\sigma$
  - Better for normal distributions
  - Handles outliers better

**Question 10.** You might have observed that sometimes the value of VIF is infinite. Why does this happen? (Do not edit)

Total Marks: 3 marks (Do not edit)

Answer: Please write your answer below this line. (Do not edit)

<Your answer for Question 10 goes here>

VIF becomes infinite when there's perfect multicollinearity:

- 1. Occurs when one feature is an exact linear combination of others
- 2. Common causes:
  - Including all dummy variables (dummy variable trap)
  - Duplicate features
  - Derived features that are linear combinations
- 3. Makes coefficient estimates unstable
- 4. Solution: Remove one of the perfectly correlated features

**Question 11.** What is a Q-Q plot? Explain the use and importance of a Q-Q plot in linear regression.

(Do not edit)

Total Marks: 3 marks (Do not edit)

Answer: Please write your answer below this line. (Do not edit)

<Your answer for Question 11 goes here>

# Q-Q (Quantile-Quantile) plot is a visualization tool that:

#### 1. Purpose:

- Assesses if residuals follow normal distribution
- Compares empirical distribution to theoretical normal

#### distribution

# 2. Interpretation:

- Points following diagonal line suggest normality
- Deviations indicate non-normality
- S-shaped curve suggests skewness
- Tails deviating suggest kurtosis issues

# 3. Importance in Linear Regression:

- Validates normality assumption
- Helps identify potential outliers
- Guides potential data transformations