

## 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?

**Ans.** Here are some of the inferences I made from my analysis of categorical variables from the dataset on the dependent variable (Count)

- Fall has the highest median, which is expected as weather conditions are most optimal to ride bike followed by summer.
- Median bike rents are increasing year on as year 2019 has a higher median than 2018, it might be due the fact that bike rentals are getting popular and people are becoming more aware about environment.
- Overall spread in the month plot is reflection of season plot as fall months have higher median.
- People rent more on non-holidays compared to holidays, so reason might be they prefer to spend time with family and use personal vehicle instead of bike rentals.
- Overall median across all days is same, but spread for Saturday and Wednesday is bigger. Maybe evident that those who have plans for Saturday might not rent bikes as it a non working day.
- Working and non-working days have almost the same median, although spread is bigger for non-working days, as people might have plans and do not want to rent bikes because of that.
- Clear weather is most optimal for bike renting, as temperature is optimal, humidity is less and temperature is less.

### 2. Why is it important to use drop\_first=True during dummy variable creation?

**Ans.** A variable with  $n$  levels can be represented by  $n - 1$  dummy variables. So, if we remove the first column, then also we can represent the data. If the value of variable from 2 to  $n$  is 0, it means that the value of first variable is 1.

Example: 'Relationship' with three levels, namely, 'Single', 'In a Relationship' and 'Married'. I would create a dummy table like the following:

Relationship status	Single	In a Relationship	Married
Single	1	0	0
In a Relationship	0	1	0
Married	0	0	1

But I can clearly see that there is no need to define **three** different levels. If I drop a level 'Single', I would still be able to explain the 3 levels.

Let us drop the dummy variable single from the column and see what the table looks like.

Relationship status	In a Relationship	Married
Single	0	0
In a Relationship	1	0
Married	0	1

If both the dummy variables namely 'In a Relationship' and 'Married' are equal to 0, that means the person is single. If 'In a relationship' is 1 and 'Married' is 0 that means that the person is in a relationship, and finally, if 'In a Relationship' is 0 and 'Married' is 1, that means that the person is married.

### 3. Looking at the pair-plot among the numerical variables, which one has the highest correlation with the target variable?

**Ans.** 'temp' has the highest correlation coefficient of 0.63.

### 4. How did you validate the assumptions of Linear Regression after building the model on the training set?

**Ans.** By plotting the residuals distribution, it came out to be a normal distribution with a mean of 0.

**5. Based on the final model, which are the top 3 features contributing significantly towards explaining the demand of the shared bikes?**

**Ans.** The following are the top 3 features contributing significantly towards explaining the demands of the shared bike:

- a temp (0.412)
- yr (0.236)
- weathersit Light rain (-0.275)

## General Subjective Questions

### 1. Explain the linear regression algorithm in detail.

**Ans.** A linear regression algorithm tries to explain the relationship between independent and dependent variable using a straight line. It is applicable to numerical variables only. Following steps are performed while doing linear regression:

- The dataset is divided into features (independent) and target (dependent) datasets.
- A linear model is fitted using the training dataset. Internally the api's from python uses gradient descent algorithm to find the coefficients of the best fit line. The gradient descent algorithm works by minimising the cost function. A typical example of cost function is residual sum of squares.
- In case of multiple features, the predicted variable is a hyperplane instead of line. The predicted variable takes the following form:

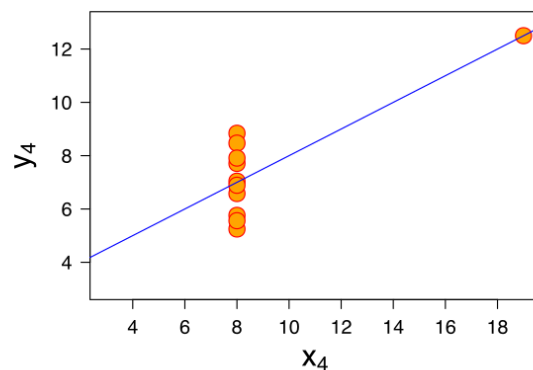
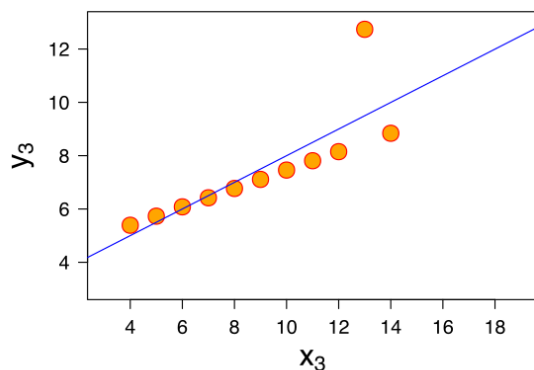
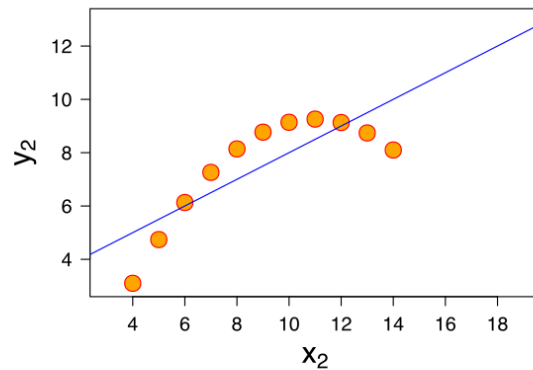
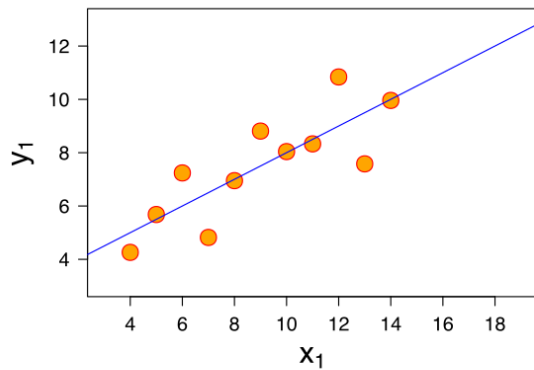
$$y = b_0 + b_1x_1 + b_2x_2 + b_3x_3 + \dots + b_nx_n$$

- The predicted variable is then compared with test data and assumptions are checked.

### 2. Explain the Anscombe's quartet in detail.

**Ans.** Anscombe's quartet comprises of four data sets that have nearly identical simple descriptive statistics but have quite different distribution when visualized graphically. The simple statistics consist of mean, sample variance of x and y, correlation coefficient, linear regression line and R-Square value.

Anscombe's Quartet shows that multiple data sets with many similar statistical properties can still be vastly different from one another when graphed. The graphs are shown below:



- First plot (top left) appears to be simple linear relationship
- The second plot (top right) is not distributed normally and correlation coefficient is irrelevant as it shows a non-linear relationship
- The third plot (bottom left) is linear but has different regression line. This is happening because of the outliers present in the data
- The fourth plot (bottom right) does not show linear relationship however due to outliers the statistics got adjusted.

In a nutshell, it is a better practice to visualize data and remove outliers before analysing it.

### 3. What is Pearson's R?

**Ans.** Pearsons are measures the strength of association of two variables. It is the covariance of two variables divided by product of their standard deviation. It has a value from +1 to -1.

- A value of 1 means a total positive linear correlation. It means that if one variable increase, then other will also increase.

- A value of 0 means no correlation
- A value of -1 means a total negative correlation. It means that if one variable increase, then other will decrease.

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

**Ans.** Scaling of a variable is performed to keep a variable in certain range. Scaling is a pre-processing step in linear regression analysis. The reason we scale a variable is to make the computation of gradient descent faster. The steps the step size of gradient descent are generally low for accuracy, if the data has some small values (values in the range 0-1) and some big variable (value in range of 0-1000) then the time taken by the gradient descent algorithm will be huge.

Normalized Scaling	Standardized Scaling
Called min max scaling, scales the variable such that the range is 0-1	Values are centered around mean with a unit standard deviation
Good for non-gaussian distribution	Good for gaussian distribution
Value is bounded between 0 and 1	Value is not bounded
Outliers are also scaled	Does not affect outliers

**5. You might have observed that sometimes the value of VIF is infinite. Why does this happen?**

**Ans.** The formula of VIF is:

$$VIF = 1 / (1 - R^2)$$

Basically, if R square is 1 then VIF become infinite. It means that there is perfect correlation between the features.

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

**Ans.** A Q-Q plot is a scatter plot of two sets of quantities against each other. Its purpose is to check if the two sets of data came from the same distribution. It is a visual check of data. If the data is from same source then the plot will appear as a line.