

## 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)

Answer:

While doing analysis on Categorical columns using the boxplot and bar plot. Below are the few points we can infer from the visualization –

- ❖ Fall season seems to have attracted more booking. And, in each season the booking count has increased drastically from 2018 to 2019.
- ❖ Most of the bookings have been done during the month of may, june, july, aug, sep and oct. Trend increased starting of the year till mid of the year and then it started decreasing as we approached the end of year.
- ❖ Clear weather attracted more booking which seems obvious.
- ❖ Thu, Fri, Sat and Sun have more number of bookings as compared to the start of the week.
- ❖ When it's not holiday, booking seems to be less in number which seems reasonable as on holidays, people may want to spend time at home and enjoy with family.
- ❖ Booking seemed to be almost equal either on working day or non-working day.
- ❖ 2019 attracted more number of booking from the previous year, which shows good progress in terms of business.

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

Answer:

`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.

Syntax -

`drop_first`: bool, default False, which implies whether to get k-1 dummies out of k categorical levels by removing the first level.

Let's say we have 3 types of values in Categorical column and we want to create dummy variable for that column. If one variable is not A and B, then it is obvious C. So, we do not need 3rd variable to identify the C.

3. Looking at the pair-plot among the numerical variables, which one has the highest correlation with the target variable? (1 mark)

Answer:

'temp' Variable has the highest Correlation with the Target Variable.

4. How did you validate the assumptions of Linear Regression after building the model on the training set? (3 marks)

Answer:

We have validated the assumption of Linear Regression Model based on below 5 Assumptions -

- ❖ **Normality of error terms**
  - Error terms should be normally distributed

❖ **Multicollinearity check**

- There should be insignificant multicollinearity among variables.

❖ **Linear relationship validation**

- Linearity should be visible among variables

❖ **Homoscedasticity**

- There should be no visible pattern in residual values.

❖ **Independence of residuals**

- No auto-correlation

5. **Based on the final model, which are the top 3 features contributing significantly towards explaining the demand of the shared bikes? (2 marks)**

**Answer:**

Below are the top 3 features contributing significantly towards explaining the demand of the shared bikes –

1. temp
2. winter
3. sep

## **General Subjective Questions**

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

**(4 marks)**

**Answer:**

Linear regression may be defined as the statistical model that analyses the linear relationship between a dependent variable with given set of independent variables. Linear relationship between variables means that when the value of one or more independent variables will change (increase or decrease), the value of dependent variable will also change accordingly (increase or decrease).

Mathematically the relationship can be represented with the help of following equation –

$$Y = mX + c$$

Here, Y is the dependent variable we are trying to predict.

X is the independent variable we are using to make predictions.

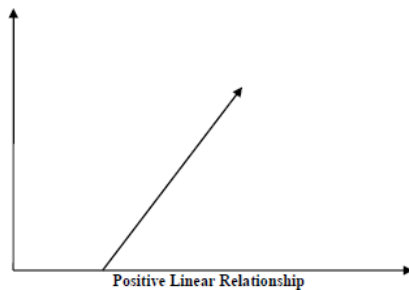
m is the slope of the regression line which represents the effect X has on Y

c is a constant, known as the Y-intercept. If X = 0, Y would be equal to c.

Furthermore, the linear relationship can be positive or negative in nature as explained below–

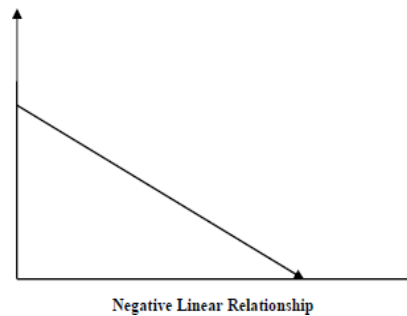
➤ **Positive Linear Relationship:**

- A linear relationship will be called positive if both independent and dependent variable increases. It can be understood with the help of following graph –



➤ **Negative Linear relationship:**

- A linear relationship will be called positive if independent increases and dependent variable decreases. It can be understood with the help of following graph –



**Linear regression is of the following two types –**

- **Simple Linear Regression:** SLR is used when the dependent variable is predicted using only one independent variable.
- **Multiple Linear Regression:** MLR is used when the dependent variable is predicted using multiple independent variables.

The equation for MLR will be:

$$y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_p x_{ip} + \epsilon$$

**Where, for i=n observations:**

$y_i$  = Dependent variable

$x_i$  = Explanatory variables

$\beta_0$  = y-intercept (constant term)

$\beta_p$  = Slope coefficients for each explanatory variable

$\epsilon$  = The model's error term (also known as the residuals)

**Assumptions -**

The following are some assumptions about dataset that is made by Linear Regression model –

✓ **Multi-collinearity –**

- Linear regression model assumes that there is very little or no multi-collinearity in the data. Basically, multi-collinearity occurs when the independent variables or features have dependency in them.

✓ **Auto-correlation –**

- Another assumption Linear regression model assumes is that there is very little or no auto-correlation in the data. Basically, auto-correlation occurs when there is dependency between residual errors.

✓ **Relationship between variables –**

- Linear regression model assumes that the relationship between response and feature variables must be linear.

✓ **Normality of error terms –**

- Error terms should be normally distributed

✓ **Homoscedasticity –**

- There should be no visible pattern in residual values.

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

**(3 marks)Answer:**

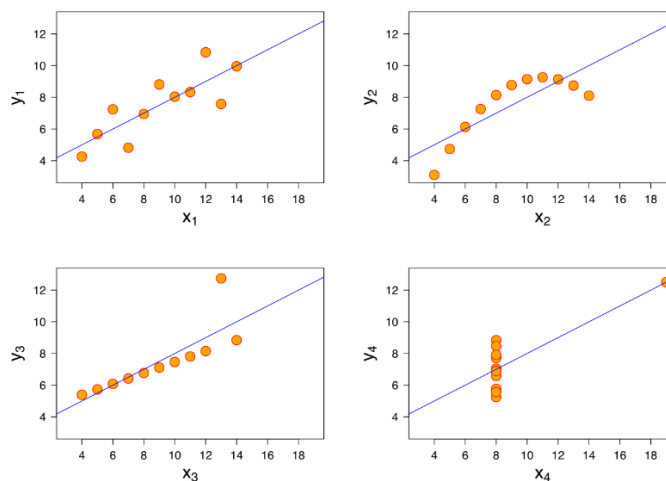
**Anscombe's Quartet** was developed by statistician Francis Anscombe. It comprises four Datasets, each containing eleven (x, y) pairs. The essential thing to note about these datasets is that they share the same descriptive statistics. But things change completely, and I must emphasize COMPLETELY, when they are graphed. Each graph tells a different story irrespective of their similar summary statistics.

The summary statistics show that the means and the variances were identical for x and y across the groups:

	I		II		III		IV	
	x	y	x	y	x	y	x	y
	10	8,04	10	9,14	10	7,46	8	6,58
	8	6,95	8	8,14	8	6,77	8	5,76
	13	7,58	13	8,74	13	12,74	8	7,71
	9	8,81	9	8,77	9	7,11	8	8,84
	11	8,33	11	9,26	11	7,81	8	8,47
	14	9,96	14	8,1	14	8,84	8	7,04
	6	7,24	6	6,13	6	6,08	8	5,25
	4	4,26	4	3,1	4	5,39	19	12,5
	12	10,84	12	9,13	12	8,15	8	5,56
	7	4,82	7	7,26	7	6,42	8	7,91
	5	5,68	5	4,74	5	5,73	8	6,89
SUM	99,00	82,51	99,00	82,51	99,00	82,50	99,00	82,51
AVG	9,00	7,50	9,00	7,50	9,00	7,50	9,00	7,50
STDEV	3,32	2,03	3,32	2,03	3,32	2,03	3,32	2,03

- Mean of x is 9 and mean of y is 7.50 for each dataset.
- Similarly, the variance of x is 11 and variance of y is 4.13 for each dataset
- The correlation coefficient (how strong a relationship is between two variables) between x and y is 0.816 for each dataset

When we plot these four datasets on an x/y coordinate plane, we can observe that they show the same regression lines as well but each dataset is telling a different story:



- Dataset I appear to have clean and well-fitting linear models.
- Dataset II is not distributed normally.
- In Dataset III the distribution is linear, but the calculated regression is thrown off by an outlier.
- Dataset IV shows that one outlier is enough to produce a high correlation coefficient.

This quartet emphasizes the importance of visualization in Data Analysis. Looking at the data reveals a lot of the structure and a clear picture of the dataset.

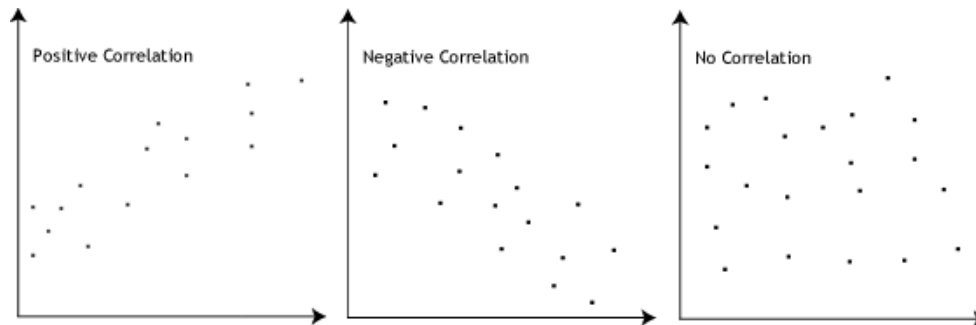
### 3. What is Pearson's R?

(3 marks)

**Answer:**

**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.

The Pearson correlation coefficient,  $r$ , can take a range of values from +1 to -1. A value of 0 indicates that there is no association between the two variables. A value greater than 0 indicates a positive association; that is, as the value of one variable increases, so does the value of the other variable. A value less than 0 indicates a negative association; that is, as the value of one variable increases, the value of the other variable decreases. This is shown in the diagram below:



The Equation for calculating Pearson's  $r$  goes as below:

$$r_{xy} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2} \sqrt{\sum_{i=1}^n (y_i - \bar{y})^2}}$$

$r$  = 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

$r = 1$  means the data is perfectly linear with a positive slope.

$r = -1$  means the data is perfectly linear with a negative slope.

$r = 0$  means there is no linear association.

**4. What is scaling? Why is scaling performed? What is the difference between normalized scaling and standardized scaling? (3 marks)**

**Answer:**

Feature Scaling is a technique to standardize the independent features present in the data in a fixed range. It is performed during the data pre-processing to handle highly varying magnitudes or values or units. If feature scaling is not done, then a machine learning algorithm tends to weigh greater values, higher and consider smaller values as the lower values, regardless of the unit of the values.

Example: If an algorithm is not using feature scaling method then it can consider the value 3000 meter to be greater than 5 km but that's actually not true and in this case, the algorithm will give wrong predictions. So, we use Feature Scaling to bring all values to same magnitudes and thus, tackle this issue.

S.NO.	Normalized scaling	Standardized scaling
1.	Minimum and maximum value of features are used for scaling	Mean and standard deviation is used for scaling.
2.	It is used when features are of different scales.	It is used when we want to ensure zero mean and unit standard deviation.
3.	Scales values between [0, 1] or [-1, 1].	It is not bounded to a certain range.
4.	It is really affected by outliers.	It is much less affected by outliers.
5.	Scikit-Learn provides a transformer called MinMaxScaler for Normalization.	Scikit-Learn provides a transformer called StandardScaler for standardization.

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

**Answer:**

If there is perfect correlation, then  $VIF = \infty$ . A large value of VIF indicates that there is a correlation between the variables. If the VIF is 4, this means that the variance of the model coefficient is inflated by a factor of 4 due to the presence of multicollinearity.

When the value of VIF is infinite it shows a perfect correlation between two independent variables. In the case of perfect correlation, we get  $R\text{-squared } (R^2) = 1$ , which leads to  $1 / (1 - R^2)$  infinity. To solve this we need to drop one of the variables from the dataset which is causing this perfect multicollinearity.

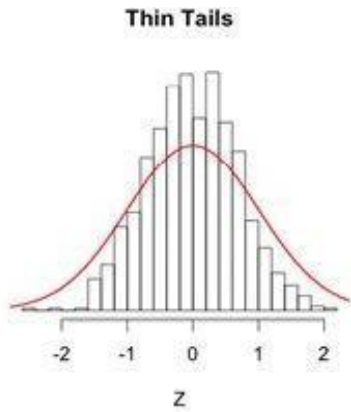
**6. What is a Q-Q plot? Explain the use and importance of a Q-Q plot in linear regression. (3 marks)**

**Answer:**

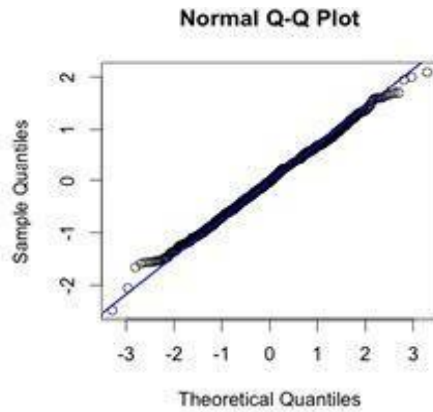
A q-q plot is a plot of the quantiles of the first data set against the quantiles of the second data set. It is used to compare the shapes of distributions. A Q-Q plot is a scatterplot created by plotting two sets of quantiles against one another. If both sets of quantiles came from the same distribution, we should see the points forming a line that's roughly straight. The q-q plot is used to answer the following questions:

- ❖ Do two data sets come from populations with a common distribution?
- ❖ Do two data sets have common location and scale?
- ❖ Do two data sets have similar distributional shapes?
- ❖ Do two data sets have similar tail behavior?

If all the points plotted on the graph perfectly lies on a straight line then we can clearly say that this distribution is normally distribution because it is evenly aligned with the standard normal variate which is the simple concept of Q-Q plot.



Normally distributed data



Normal Q- Q Plot













