Strategy: to tackle a data science problem.

1) Exploratory Analysis: (Data Exploration)

Tools & methods to be used to explore the data.

a) COFOUNDING REASONS:

develop cofounding reasons to build and justify which key features should be used to harness the results.

These selected features (co-founding reasons) will steer an ideal learning curve where there is a short gap between training and testing data. The ideal set will contain minimum feature which will give the mentioned result.

b) perform statical analysis on the data.

* measure MEAN, Median, Mode.
* Take care of Outliers so mean $\bar{X}$ is not bias.

* Make sure the mean $\bar{X}$ lies between IQR.
Inter quantile range is the range between 1st Qtr → 3rd Qtr.
* If the mean does not lie between IQR. Then there are Outliers

* Identify the Outliers, Isolate the data from the outliers and compare the $\bar{X}$, mean and the model before and after the $\bar{X}$, mean is corrected.
* If we have incorrect model due to outliers, we will introduce an in incorrect mean to fill in null value, which will skew the prediction.
* when the model is skewed, the recall and precision of the prediction will be incorrect, since our data is skewed -
* when we are training the data on the offsets will outliers introduced our training and test splits will have contradict values since the existing data was correct but the data which was imputed with an incorrect mean $\bar{X}$, due to outliers will have drifted values leaning towards the outlier. Therefore, the Recall & Precision of the Test prediction will will reduce and the results will contain poor accuracy.