

# Capstone Project -2

## Yes bank Stock Closing Price Prediction

Done By:

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# Problem Statement

- Perform regression analysis using multiple models to predict the closing price of the stock and compare the evaluation metrics for all of them to find the best model.
- Prediction of Yes Bank stock closing price.
- Getting accuracy score of several machine learning model.

# Introduction

- Data set - data Yes Bank Stock Prices - contains observations regarding open,close, high and low prices of the yes bank stock from July 2005 - November 2020.
- Date: Monthly observation of stock prices since its inception.
- Open: The price of a stock when stock exchange market open for the day.
- Close: The price of a stock when stock exchange market closed for the day.
- High: The maximum price of a stock attained during given period of time.
- Low: The minimum price of a stock attained during given period of time.

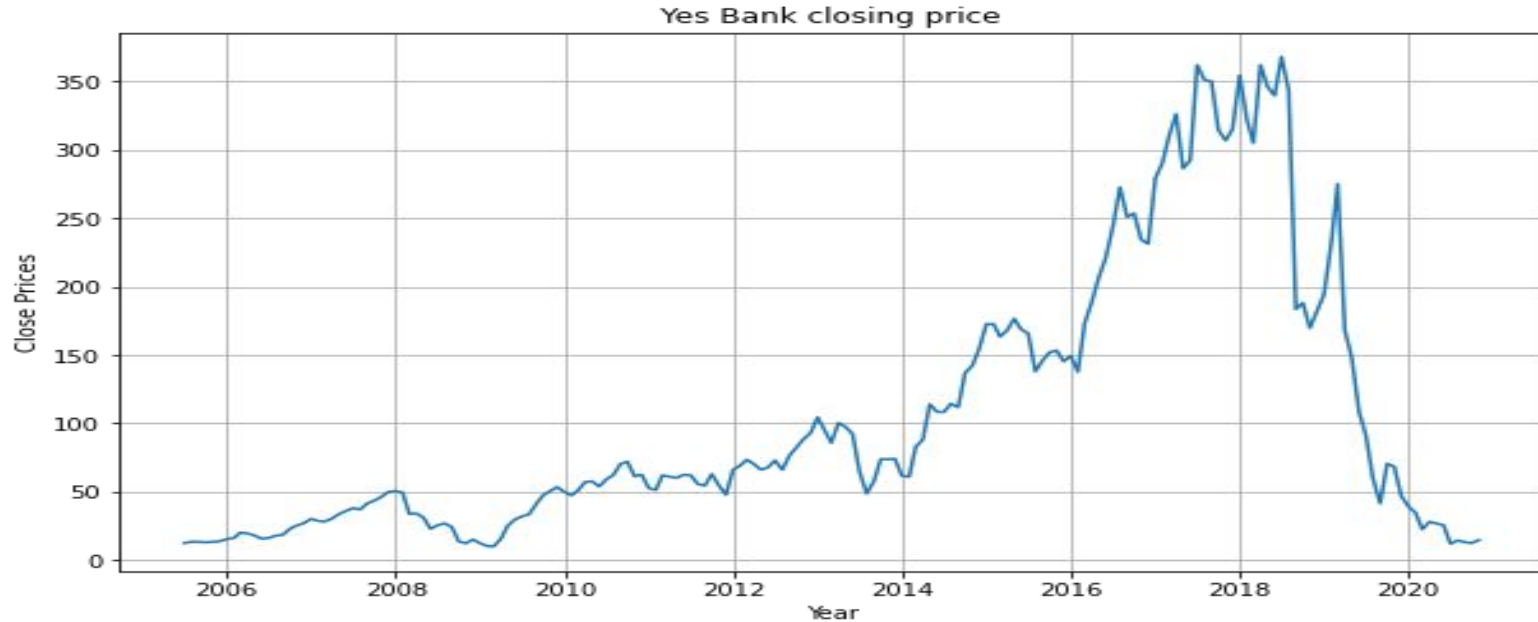
# Data Cleaning



- Null Values Treatment
- Duplicated Values Treatment
- Date Format Change (i.e from Jul-05 to 2005-07-01)
- Checking outliers
- So after successfully cleaning the dataset we have 185 columns and 5 rows

# Exploratory Data Analysis

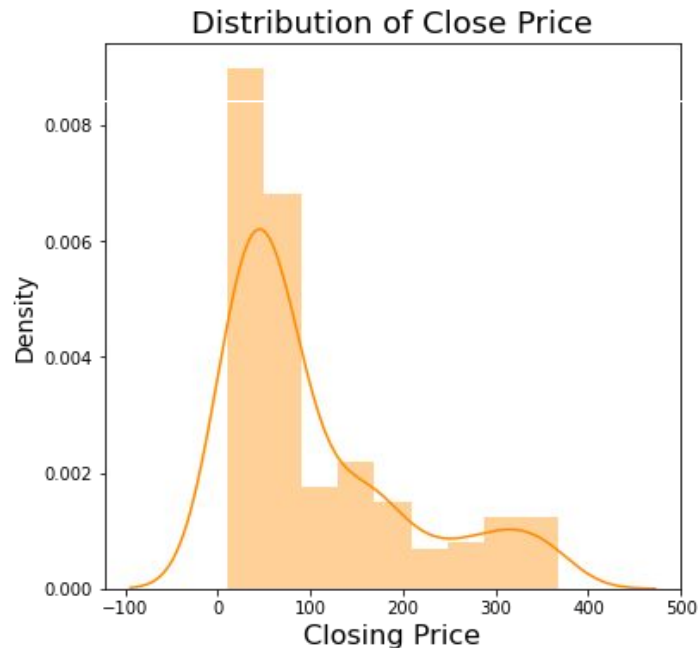
## Visualising The Data



Here, it's clearly visible from the above plot that the stock prices saw a significant rise from year 2006 to 2018. However, since 2018 the stock prices saw a major downfall and that may be due to the fraud case.

# EDA (Continued)

- **Distribution of Closing Price :**



- Distribution of closing price is right skewed.
- We need this distribution to be normal distribution for training algorithm.

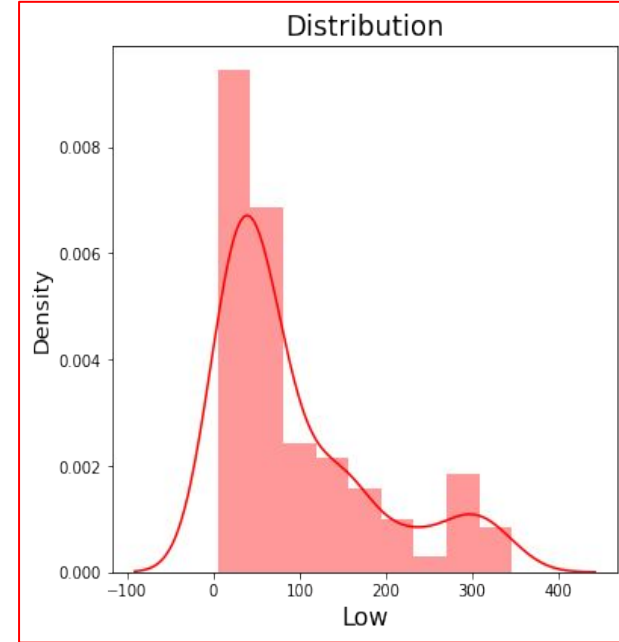
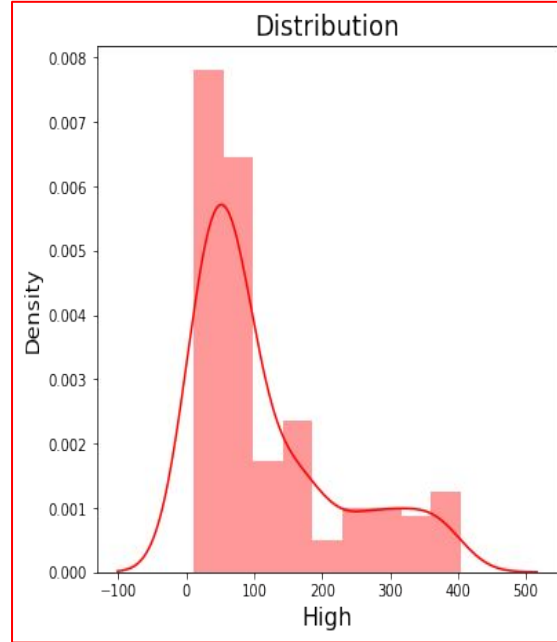
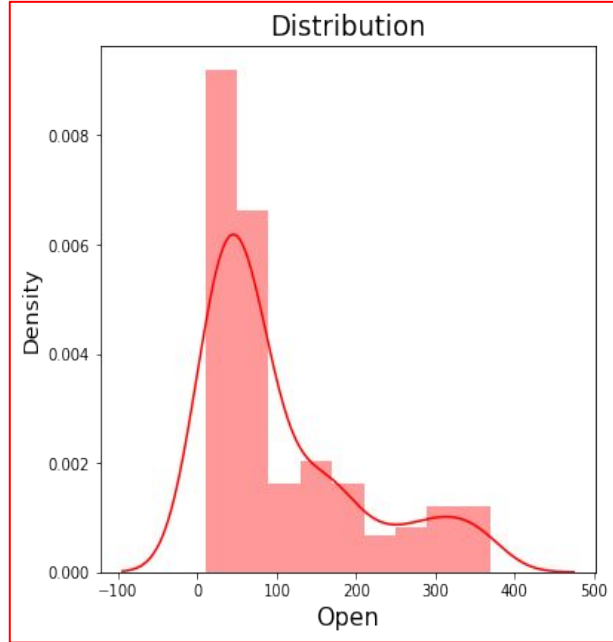
- **After Log Transformation :**



- Distribution of closing price is normal distribution.

# EDA (Continued)

## • Distribution of Open, High & Low Price of a stock :

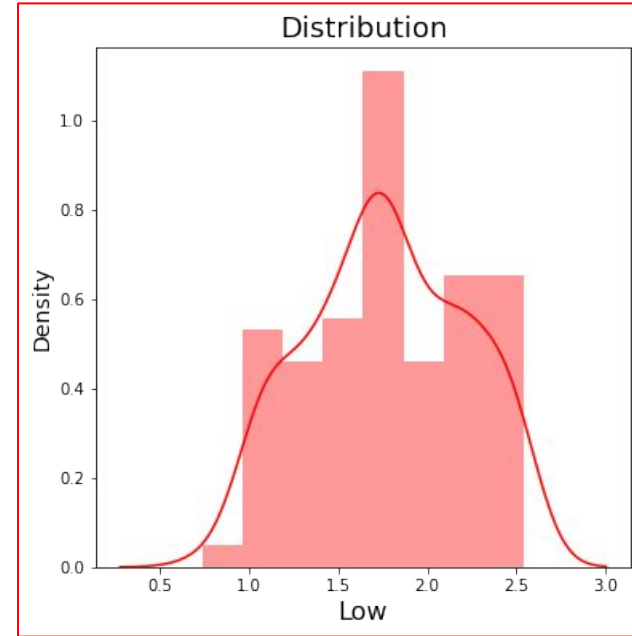
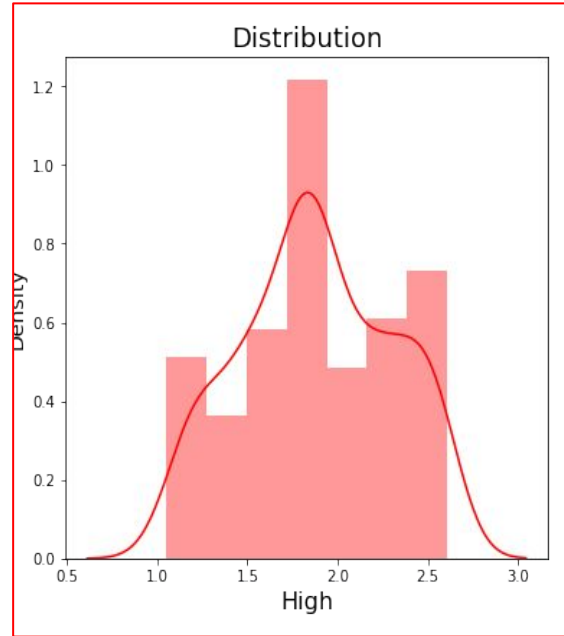
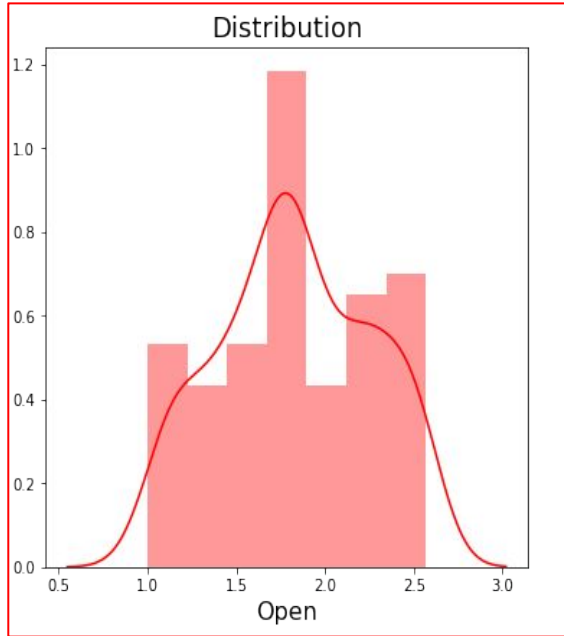


- Distribution of opening price, high price and low price are also right skewed.
- Log transformation applied to make this distribution normal.



# EDA (Continued)

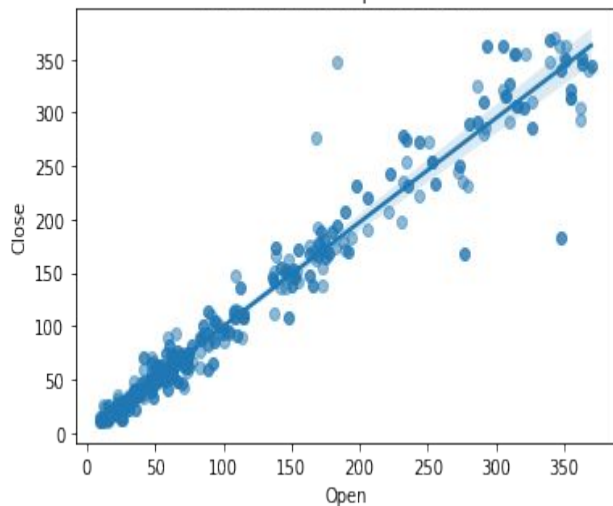
- Distribution of Open, High & Low Price of a stock after Log Transformation :**



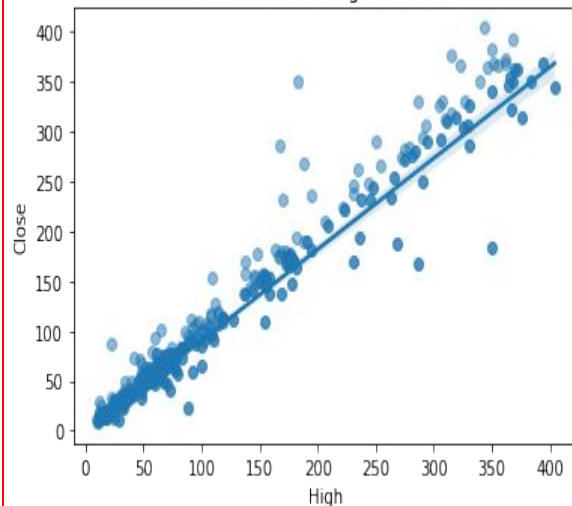
- Distribution of opening price, high price and low price are now normal distribution.

# Bivariate Analysis Plots

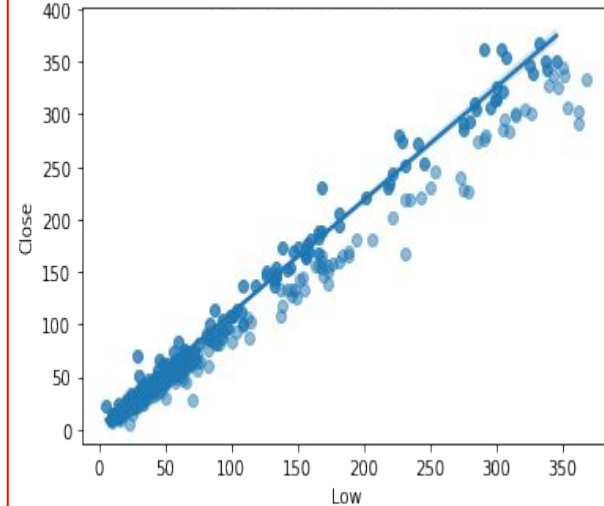
Scatter Plot Of Open Or Close



Scatter Plot Of High Or Close

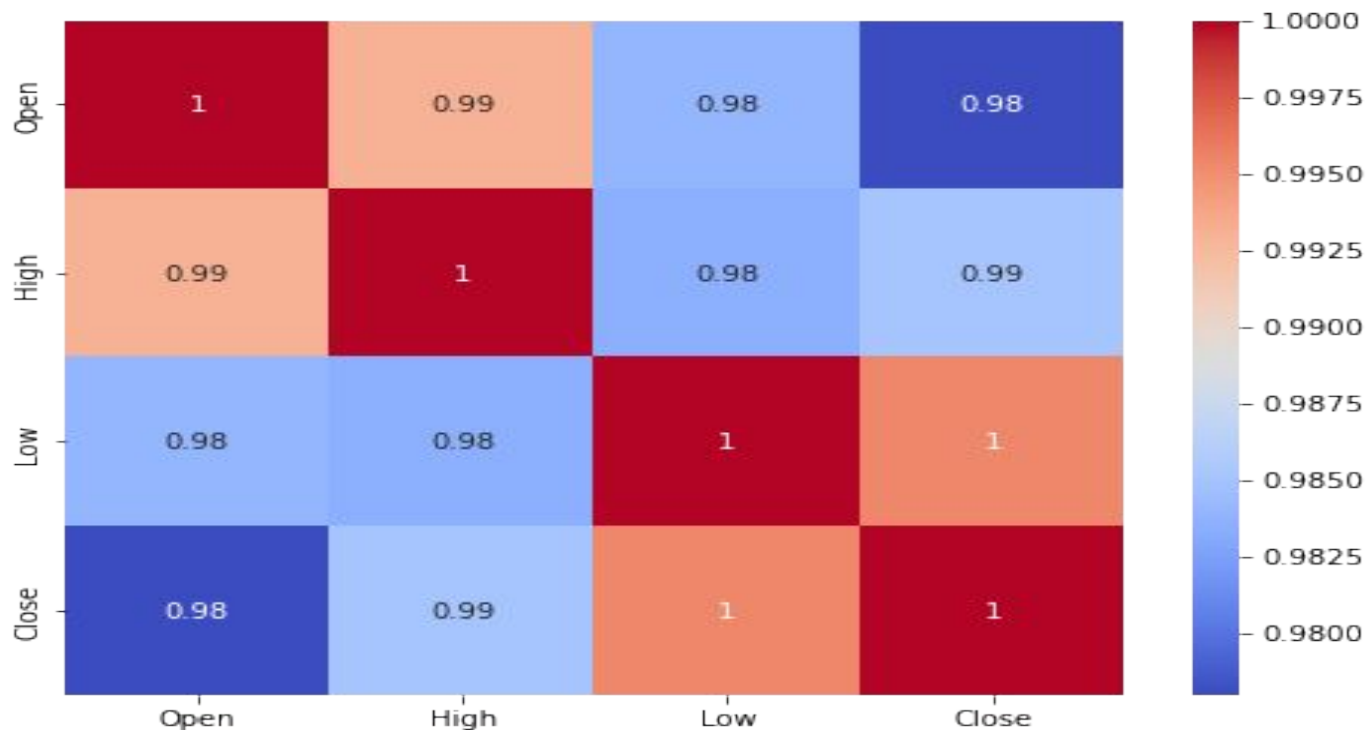


Scatter Plot Of Low Or Close



# EDA (Continued)

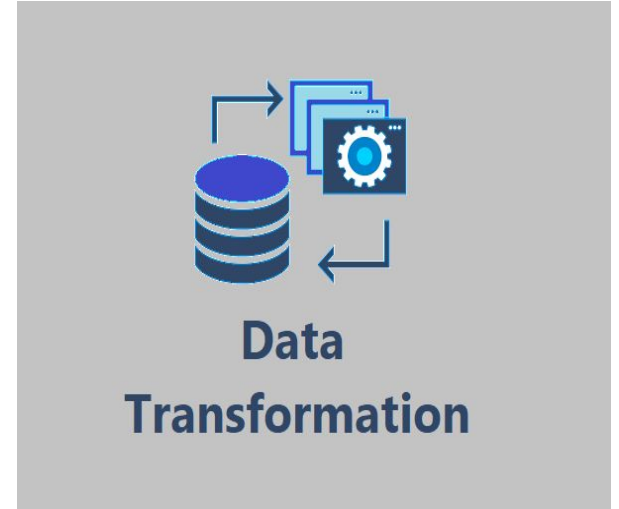
- Correlation :



- All the features are strongly correlated with each other.

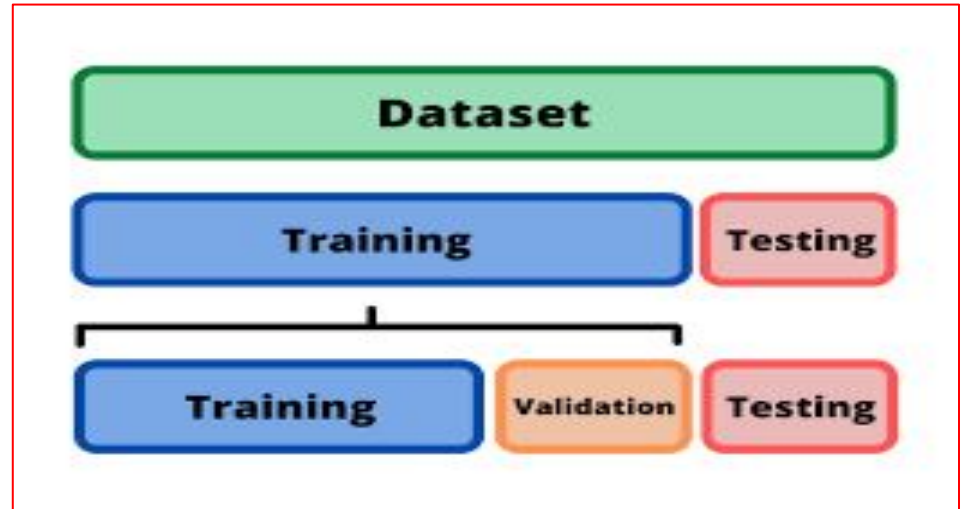
# Transformation of Data

- To scale data into a uniform format that would allow us to utilize the data in a better way.
- For performing fitting and applying different algorithms to it.
- The basic goal was to enforce a level of consistency or uniformity to dataset.



# Splitting Data

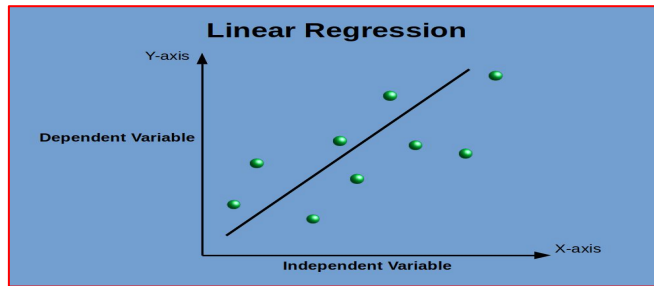
- Data splits into training dataset and testing dataset.
- Training dataset is for making algorithm learn and train model.
- Test dataset is for testing the performance of train model.
- Here 80% of data taken as training dataset & remaining 20% of dataset used for testing purpose.



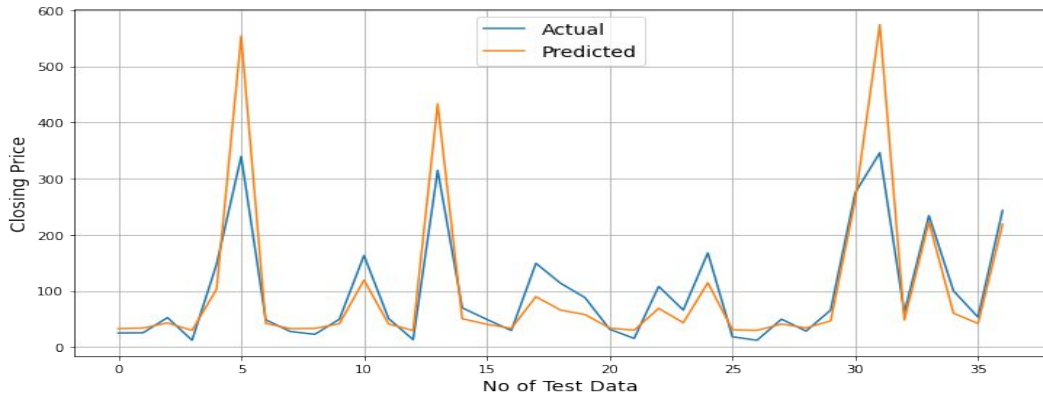
# Fitting Different Model

## Linear Regression

- Linear regression is one of the easiest and most popular Machine Learning algorithms.
- It is a statistical method that is used for predictive analysis.
- Linear regression algorithm shows a linear relationship between a dependent and independent variable; hence it is called as linear regression.



Actual Vs Price Close Price

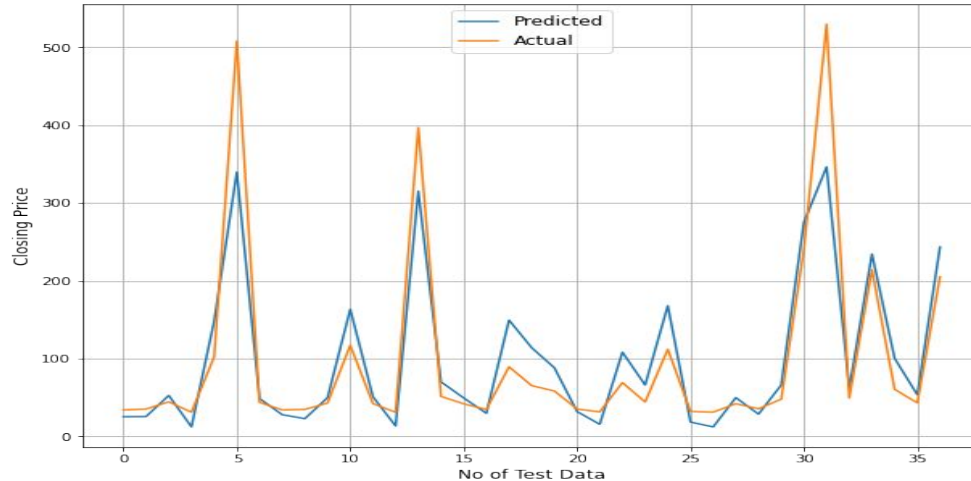


### Evaluation Metrics: Linear Regression

| MSE    | RMSE   | MAE    | MAPE   | R2     |
|--------|--------|--------|--------|--------|
| 0.0316 | 0.1777 | 0.1513 | 0.0954 | 0.8226 |

# Lasso Regression

- Lasso: Least Absolute Shrinkage and Selection operator
- It is a regression analysis method that performs both variable selection and regularization in order to enhance the prediction accuracy and interpretability of the resulting statistical model.
- This method performs L1 regularization.



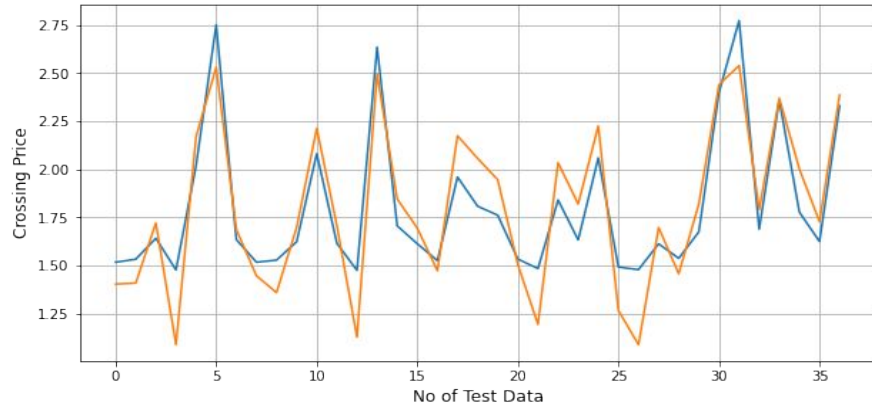
## Evaluation Metrics: Lasso Regression

| MSE    | RMSE   | MAE    | MAPE   | R2     |
|--------|--------|--------|--------|--------|
| 0.0326 | 0.1806 | 0.1535 | 0.0978 | 0.8168 |

# Ridge Regression

- Ridge regression is a model tuning method that is used to analyse any data that suffers from Multicollinearity.
- When the issue of multicollinearity occurs, least-squares are unbiased, and variances are large, this results in predicted values to be far away from the actual values.
- This method performs L2 regularization.

Actual Vs Predicted Value



## Evaluation Metrics: Ridge Regression

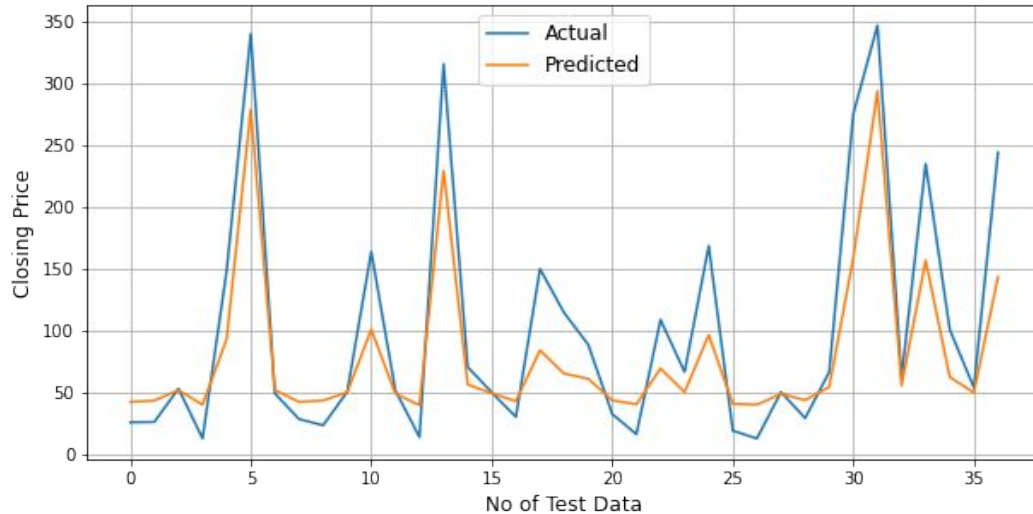
| MSE    | RMSE   | MAE    | MAPE   | R2     |
|--------|--------|--------|--------|--------|
| 0.0321 | 0.1791 | 0.1522 | 0.0959 | 0.8197 |



# Elastic Net

- Elastic net is a popular type of regularized linear regression that combines two popular penalties, specifically the L1 and L2 penalty functions.
- Elastic Net is an extension of linear regression that adds regularization penalties to the loss function during training.

Actual Vs. Predicted Close Price: Elastic Net

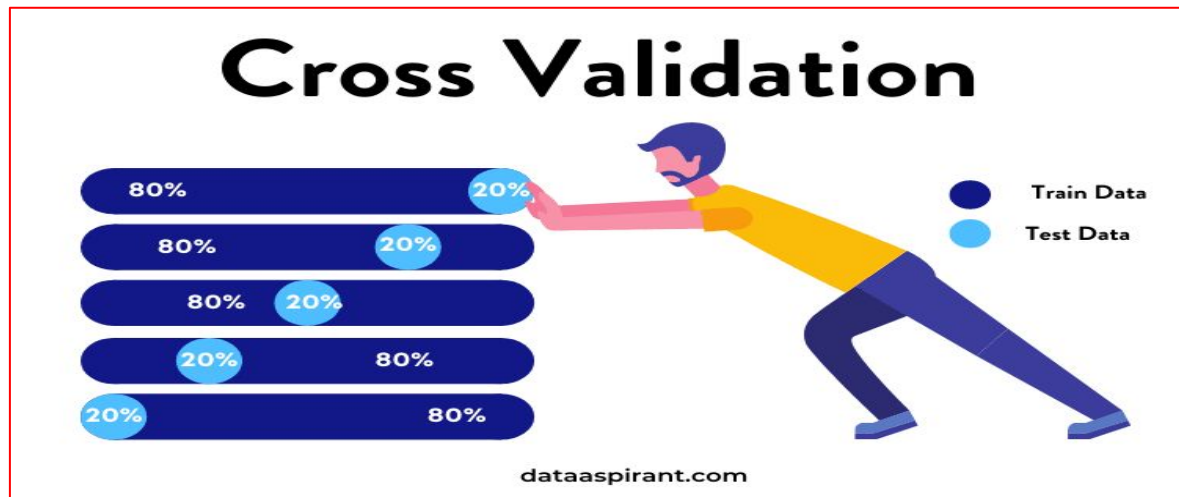


## Evaluation Metrics: Elastic Net

| MSE   | RMSE  | MAE   | MAPE  | R2    |
|-------|-------|-------|-------|-------|
| 0.051 | 0.226 | 0.182 | 0.118 | 0.714 |

# Cross Validation & Hyperparameter Tuning

- It is a resampling procedure used to evaluate machine learning models on a limited data sample.
- Basically, Cross Validation is a technique using which Model is evaluated on the dataset on which it is not trained that is it can be a test data or can be another set as per availability or feasibility.
- Tuning the hyperparameters of respective algorithms is necessary for getting better accuracy and to avoid overfitting.



- Cross Validation & Hyperparameter tuning on Lasso Regression

- Cross Validation & Hyperparameter tuning on Ridge Regression

- Cross Validation & Hyperparameter tuning on Elastic Net

### Evaluation Metrics :- CV & tuning on Lasso Regression

| MSE   | RMSE   | MAE    | MAPE   | R2     |
|-------|--------|--------|--------|--------|
| 0.032 | 0.1789 | 0.1522 | 0.0961 | 0.8202 |

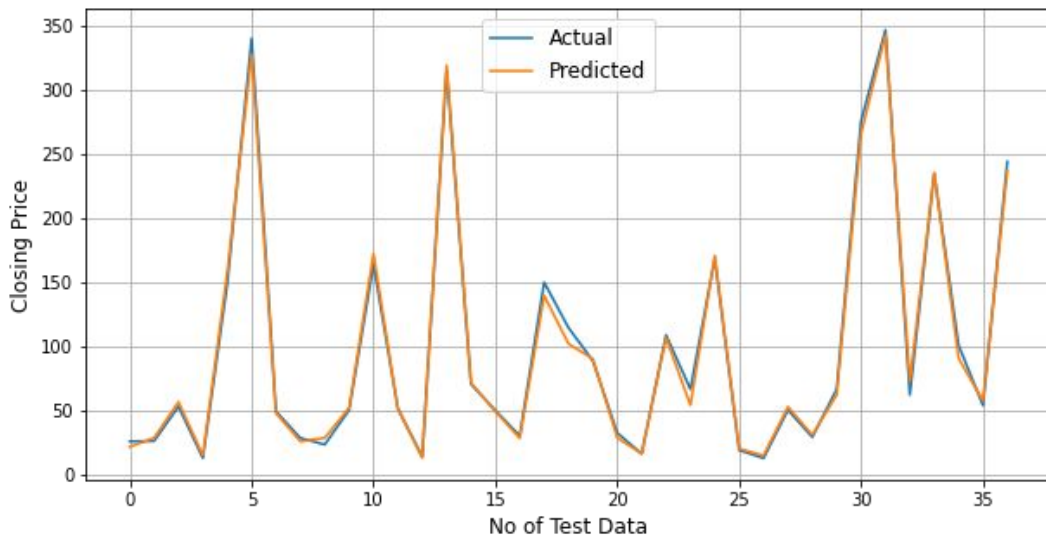
| MSE    | RMSE   | MAE    | MAPE   | R2     |
|--------|--------|--------|--------|--------|
| 0.0327 | 0.1808 | 0.1534 | 0.0971 | 0.8164 |

| MSE   | RMSE   | MAE    | MAPE   | R2     |
|-------|--------|--------|--------|--------|
| 0.032 | 0.1789 | 0.1522 | 0.0961 | 0.8202 |

# XGBoost Regressor

XGBoost stands for “Extreme Gradient Boosting”.XGBoost is an optimized distributed gradient boosting library designed to be highly efficient, flexible and portable. It implements Machine Learning algorithms under the Gradient Boosting framework. It provides a parallel tree boosting to solve many data science problems in a fast and accurate way.

Actual Vs. Predicted Close Price: XG Boost



## Evaluation Metrics: XGBoost Regression

MSE

RMSE

MAE

MAPE

R2

0.0016

0.0394

0.0303

0.0196

0.9913

# Conclusion

- The popularity of stock closing is growing extremely rapidly day by day which encourage researcher to find new methods if any fraud happens.
- This technique is used for prediction is not only helpful to researchers to predict future stock closing prices or any fraud happen or not but also helps investors or any person who dealing with the stock market in order to prediction of model with good accuracy.
- In this work we use linear regression technique, lasso regression, ridge regression and elastic net regression technique. these four models gives us the following results
- High, low, open are directly correlate with the closing price of stocks
- Target variable(dependent variable) strongly dependent on independent variables
- Xgboost regression is best model for yes bank stock closing price data this model use for further prediction

**THANK YOU**