

Capstone Project – 2

Supervised Machine Learning -Regression

Yes Bank Stock Closing Price Prediction

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Will Be Discussing On:

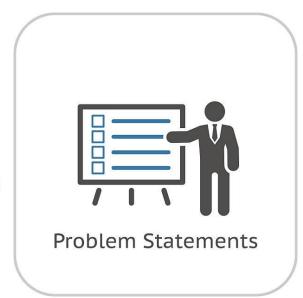
- 1. Problem Statement
- 2. Introduction
- 3. Data Cleaning
- 4. Exploratory Data Analysis (EDA)
- 5. Transforming Data
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- 7. Fitting Different Model
- 8. Cross Validation & Hyperparameter Tuning
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1. Problem Statement:

- Prediction of Yes Bank stock closing price.
- Getting accuracy score of several machine learning model.
- Will be discovering important factors that govern hotely vear to book a hotel room.





2. Introduction:

- 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.



3. Data Cleaning

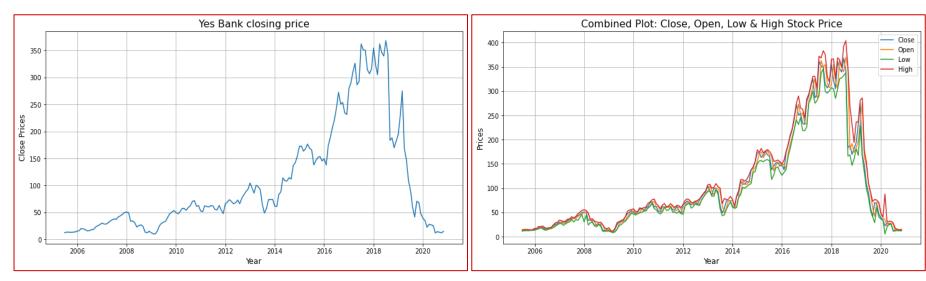
- Null Values Treatment
- Duplicated Values Treatment
- Date Format Change

(i.e from Jul-05 to 2005-07-01)





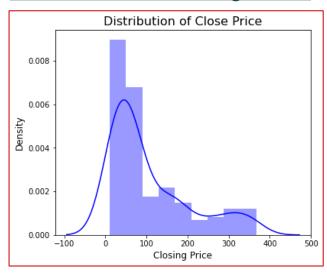
4. Exploratory Data Analysis (EDA)



- Sudden fall in stock after 2018 which justify the effect of fraud case involving Rana Kapoor.
- From combined plot, it shows strong correlation between each feature.

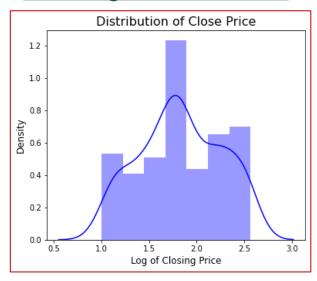


• Distribution of Closing Price:



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

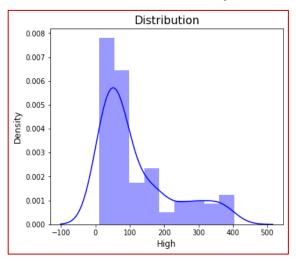
After Log Transformation

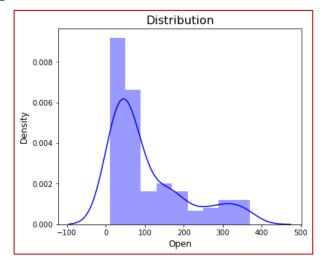


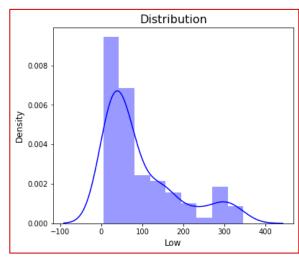
Distribution of closing price is normal distribution.



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



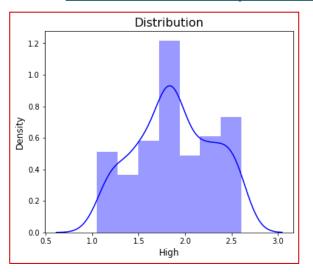


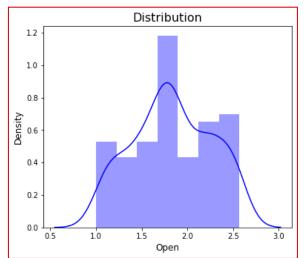


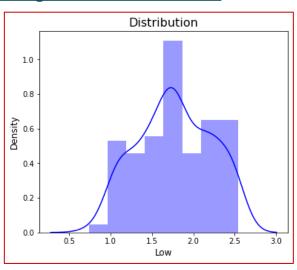
- Distribution of opening price, high price and low price are also right skewed.
- Log transformation applied to make this distribution normal.



• <u>Distribution of Open, High & Low Price of a stock after Log Transformation:</u>



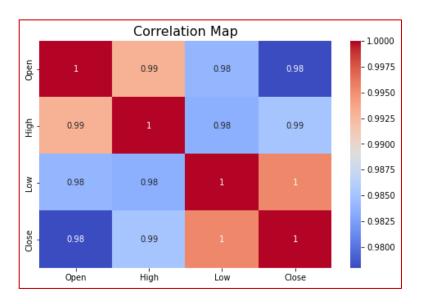




• Distribution of opening price, high price and low price are now normal distribution.



• Correlation:



• All the features are strongly correlated with each other.



5. Transformation of Data

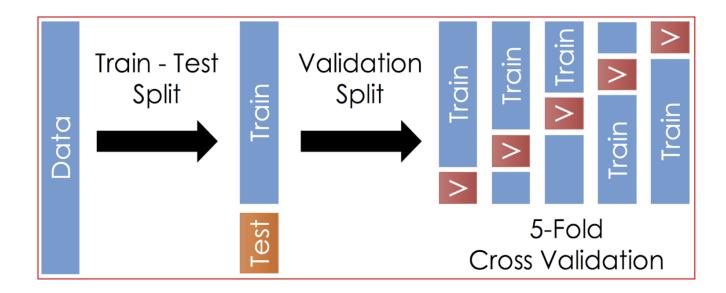
- To scale data into a uniform format that would allow us to utilize the data in a better way.
- For performing fitting and applying different algorithms to it.
- The basic goal was to enforce a level of consistency or uniformity to dataset.



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6. 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.





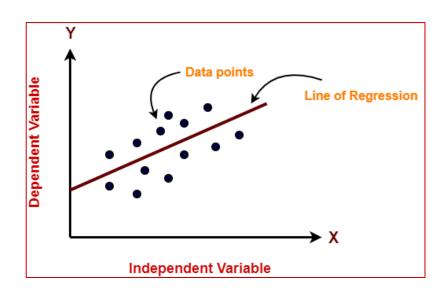
7. Fitting Different Model

7.1 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.

Evaluation Metrics: Linear Regression							
MSE RMSE MAE MAPE R2							
0.032	0.178	0.151	0.095	0.823			





7.2 <u>Lasso Regression</u>

- 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						
0.032	0.179	0.152	0.096	0.820		



7.3 Ridge Regression

- Ridge regression is a model tuning method that is used to analyses 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.

Evaluation Metrics: Ridge Regression						
MSE RMSE MAE MAPE R2						
0.032 0.178		0.151	0.095	0.823		



7.2 Elastic Net

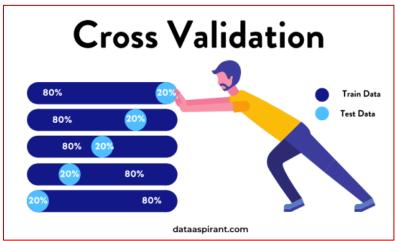
- Elastic net is a popular type of regularized linear regression that combines two popular penalties, specifically the L1 and L2 penalty functions.
- Elastic Net is an extension of linear regression that adds regularization penalties to the loss function during training.

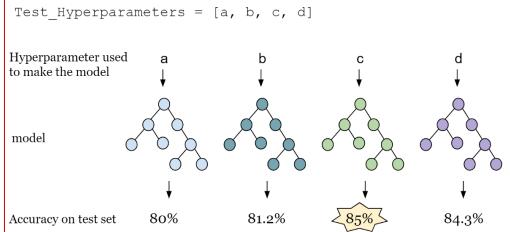
Evaluation Metrics: Elastic Net							
MSE	MSE RMSE MAE MAPE R2						
0.036	0.191	0.157	0.102	0.820			

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8. 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							
0.032 0.180 0.153 0.097 0.819							

Evaluation Metrics:							
CV & tuning on Ridge Regression							
MSE RMSE MAE MAPE R2							
0.033	0.180	0.153	0.097	0.817			

Evaluation Metrics:							
CV & tuning on Elastic Net							
MSE RMSE MAE MAPE R2							
0.032	0.180	0.153	0.097	0.819			



9. Evaluation Metrics Comparison

	Model	MSE	RMSE	MAE	MAPE	R2
0	LinearRegression	0.032	0.178	0.151	0.095	0.823
1	Ridge	0.032	0.178	0.151	0.095	0.823
2	Lasso	0.032	0.179	0.152	0.096	0.820
3	ElasticNet	0.036	0.191	0.157	0.102	0.796



10. Conclusion

- 1. We got a maximum accuracy score of 82%.
- 2. Linear, lasso and ridge regression show almost same R squared values.
- 3. Whereas elastic net model shows lowest R squared value and high MSE, RMSE, MAE & MAPE values.
- 4. Close, Open and high price of stock are strongly correlated with each other.