

# Stock Price Prediction

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## Introduction

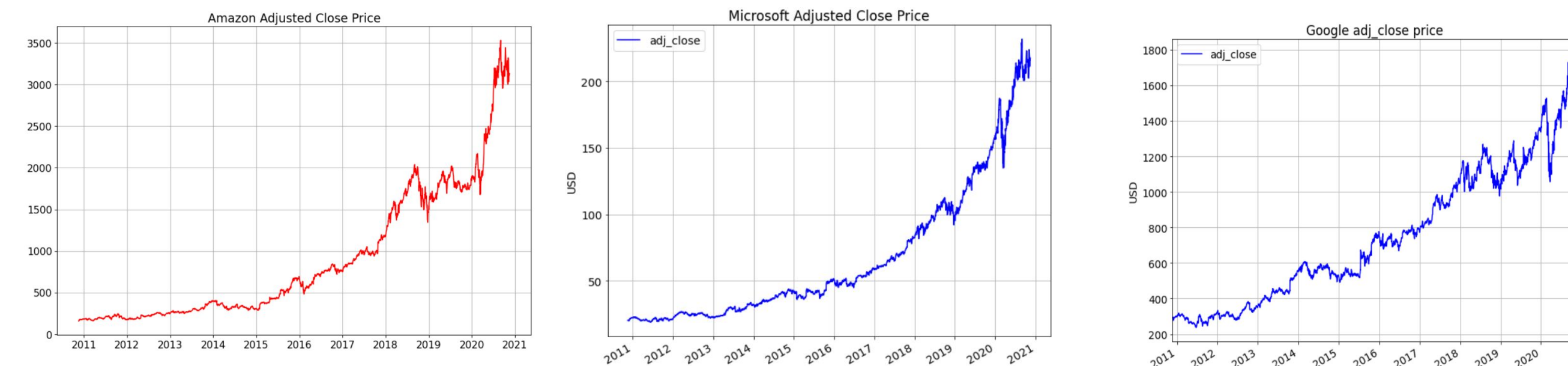
- Time series Analysis is one of the most challenging and interesting application of Machine Learning Algorithms
- Stock price prediction is an enticing prospect for data scientist to evaluate and predict future prices
- Predicting how the stock market will perform is one of the most difficult things to do. There are so many factors involved in the prediction – physical factors vs. psychological, rational and irrational behaviour, etc. All these aspects combine to make share prices volatile and very difficult to predict with a high degree of accuracy.
- We see daily ups and downs in the market and assume a pattern which is highly dependent on sector as well as the company

## Objective

- To predict tech industry stock price for next day and to help investors evaluate their portfolio
- Minimize the prediction error to build an accurate model

## Data Understanding

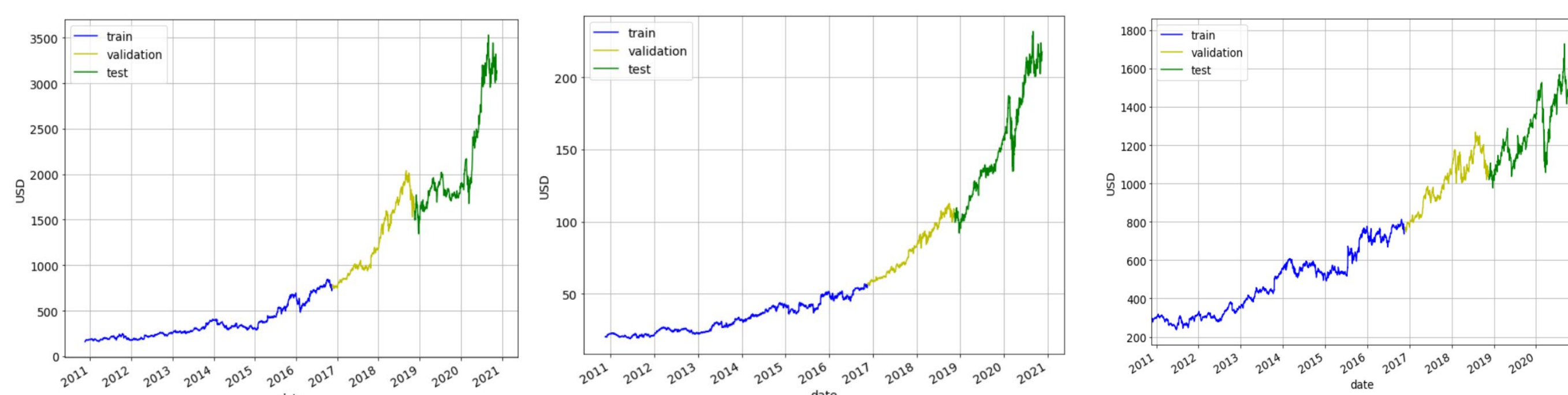
- 10-year daily trading for Amazon, Google and Microsoft was collected from yahoo finance



- The above graphs shows the adjusted closing price for Amazon, Microsoft and Google

## Data Preparation

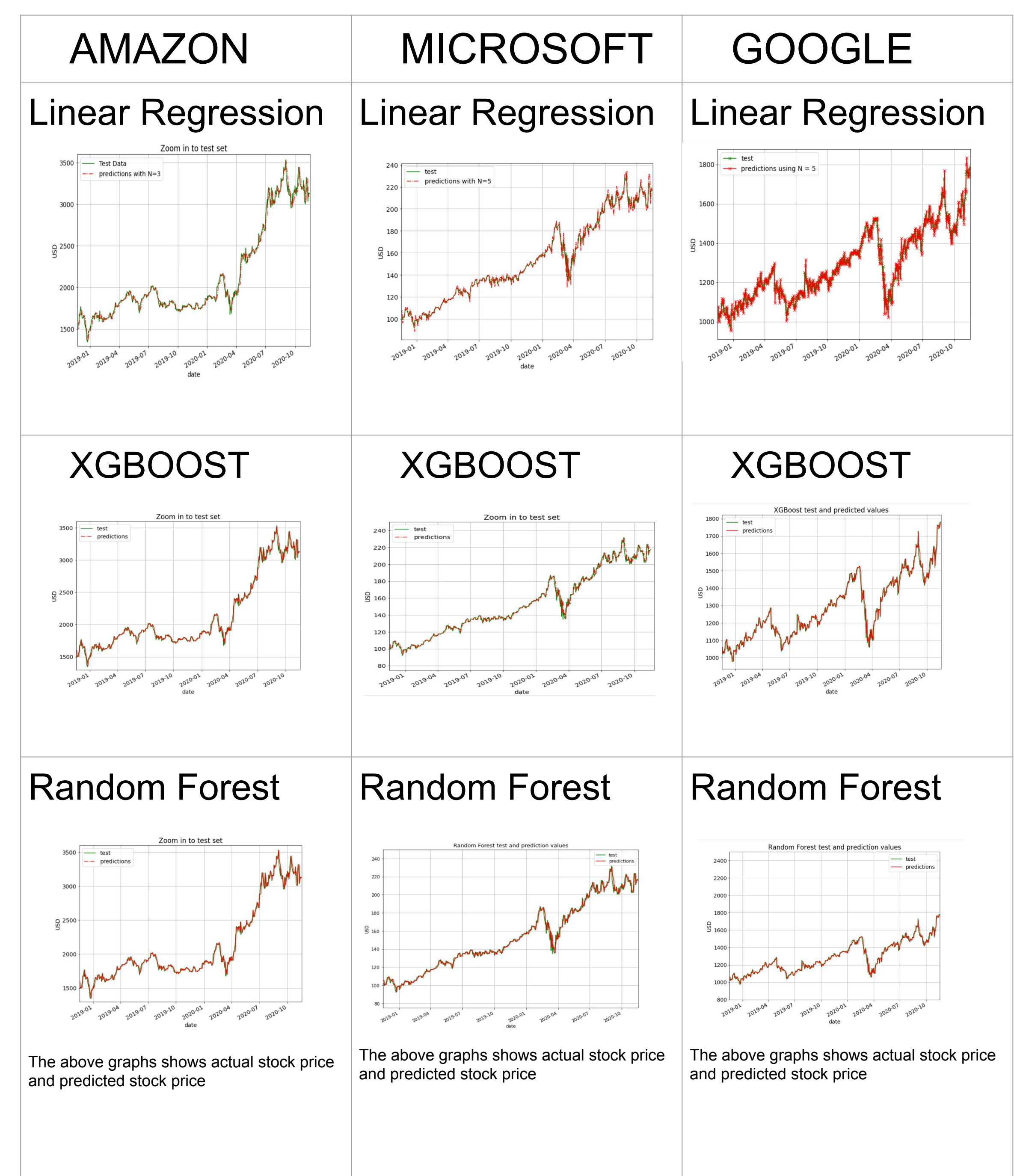
- The data was divided into 3 sets:
- Training Set – 60% of the data
- Validation Set – 20% of the data
- Test Set – 20% of The data
- Since we know stock price are highly volatile and the range of data in training set , test set and validation set is different, We normalized the data using previous 5 days rolling average to have mean 0 and standard deviation 1 for all the data.



## Modeling

- Linear Regression
- XgBoost
- Random Forest
- For the three algorithms we added additional features like lag for previous N days and OHLC difference for previous N days to know the importance of different features and relevance of previous lags.
- Features as input to our model:
  1. Difference between High and low of stock for past n days  
e.g:High-Low\_lag\_1 ,High-Low\_lag\_2,High-Low\_lag\_3
  2. Difference Between Open and close for previous n days  
e.g:Open-Close\_lag\_1,Open-Close\_lag\_2
  3. Difference Between Open and close for previous n days  
e.g:Open-Close\_lag\_1,Open-Close\_lag\_2
  4. Trading volume lag for previous n days  
e.g:Volume\_lag\_1,Volume\_lag\_2,Volume\_lag\_3
- We normalised the training, testing and validation set to have mean 0 and variance 1

## Results & Evaluation



RMSE VALUES	Linear Regression	RF Regressor	XGBoost
GOOG	32.621	26.249	26.227
AMZN	49.343	49.982	49.337
MSFT	3.914	3.474	3.464

## Conclusion

- Using Feature Engineering and testing data on three different models we conclude that Xgboost perform the best.
- Our Investment of \$100 in each stock through the test period (2018-2020) will give us a cumulative return of 94.73%
- S&P500 return is 31.65 % in the same period