

Stock News Prediction

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Outline

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- Data
- Methodology
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- Conclusions
 - For a live dashboard an XGBOOST model is the best fit for implementation.
 - Add news sentiment to stock predictions.
 - Stock volatility adds noise to prediction models. To mitigate this, lower the test size.
- Further Work
- Questions

Business Definition

Business Problem: Stock prices are a very important metric because they represent the value of publicly traded companies in the eyes of society. There is also a strong financial incentive to have a higher stock value for current shareholders because stocks can be liquidated into currency. Stock prices fluctuate based on the supply and demand of stocks associated with a particular company. If stock price can be predicted it could bring these business values:

- Increased confidence of return when investing in a company's stock
- Increase the confidence of current shareholders regarding stock's direction which can influence their decision to buy/sell your stock
- Determine key metrics that drive a stock price up or down
- Forecasting stock crashes will lead to the minimization of capital loss

Data

- The data for this dataset comes from <https://fmpcloud.io/>
- The data included in the dashboard are
 - Annual Revenue
 - Earnings Per Share
 - Insider Trades
 - Recent News
 - Stock Price Over Time
 - Recent News + Sentiment Scores

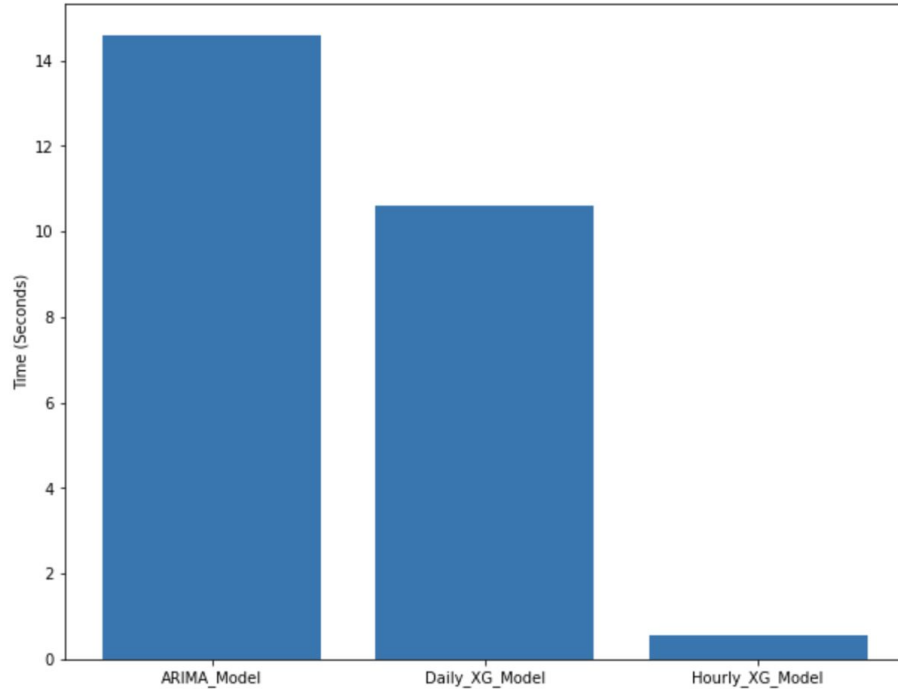
Methodology

- General Model Daily Prediction
 - XGBOOST
 - Contains 3 day previous values for:
 - High Price
 - Close Price
 - Change Over Time Decimal
 - Volume
 - Previous day news sentiment

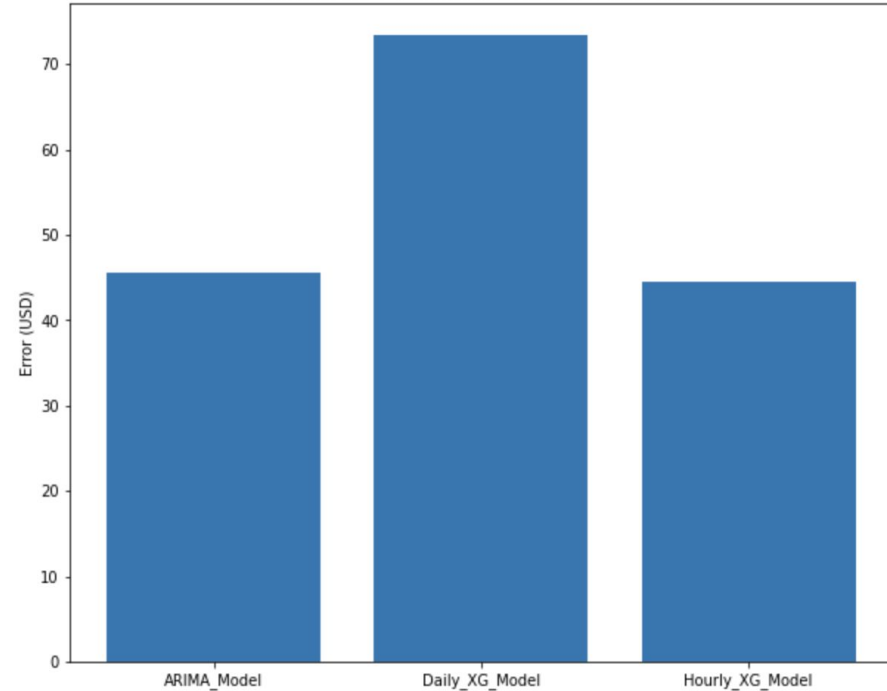
- General Model Hourly Prediction
 - XGBOOST
 - Contains 3 hourly previous values for:
 - High Price
 - Close Price
 - Volume
 - Low Price

Results: Model Comparison

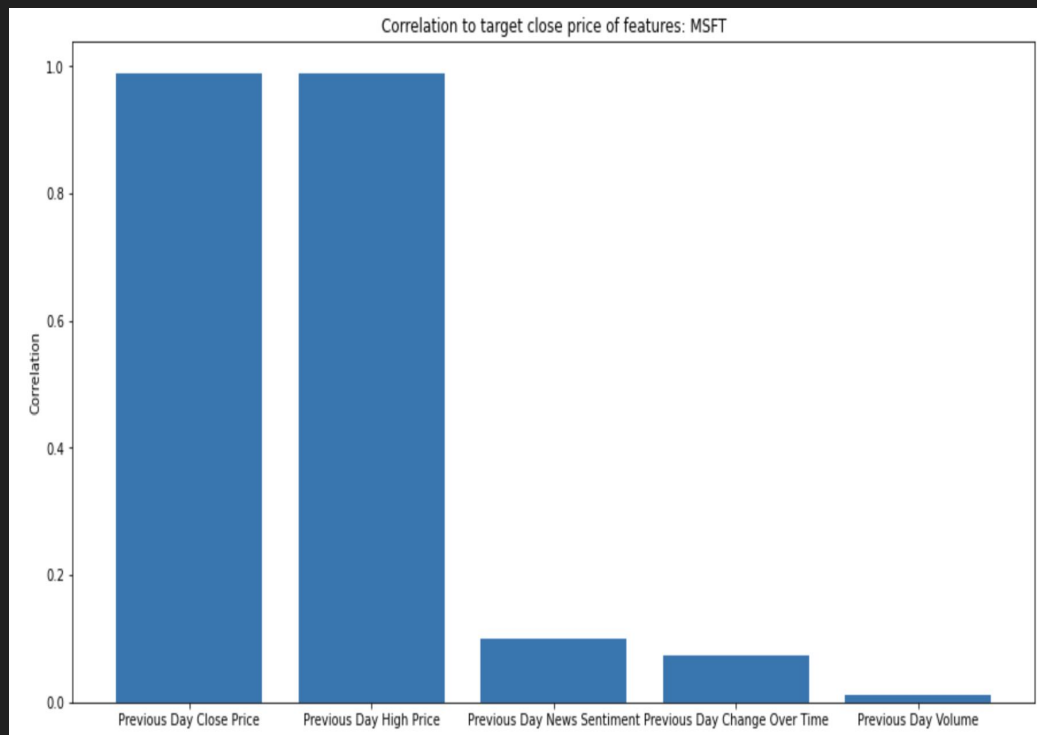
Times of Stock Prediction Models



Mean Absolute Error of Stock Prediction Models : AMZN



Results: News Sentiment



	importance	feature_name
0	0.472796	Close_1day
1	0.293361	Close_3day
2	0.148589	Close_2day
3	0.049592	High_1day
4	0.012810	High_2day
5	0.004726	Cot_3day
6	0.004395	High_3day
7	0.003072	Vol_3day
8	0.002869	Cot_1day
9	0.002842	Vol_1day
10	0.001985	Vol_2day
11	0.001796	News_Sentiment_1day
12	0.001170	Cot_2day

Results: Overfitting

60/40 SPLIT, Hourly Prediction For TSLA, Next Hour Prediction: [918.05443811] +/- 66.64954603161402



75/25 SPLIT, Hourly Prediction For TSLA, Next Hour Prediction: [939.22095202] +/- 24.102219011398986



Conclusions

- For a live dashboard an XGBOOST model is the best fit for implementation.
 - Faster than ARIMA/NN
- Add news sentiment to stock predictions.
- Stock volatility adds noise to prediction models. To mitigate this, lower the test size.

Further Work

- Provide with more time, I would :
- Investigate if/how insider training affects model performance
- Implement analyst ratings to the prediction model
- Investigate actively raising stocks to determine causers of such movement

Questions?

About the author: Strongly self-motivated data science professional with 3 years of programming experience. Recent graduate of a full time data science bootcamp and professional experience as a language analyst for 3 years. Passionate about data, STEM, business intelligence and am eager to learn and grow.

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