

Final Project (24.3)

Stock Predictor – Price & Media

Agenda and Team Members

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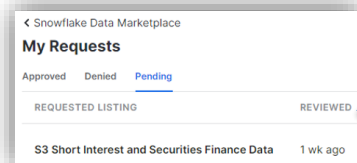
Final Project - Stocks

Project Scope

- Predict future stock price and how it may be influenced by social media noise – such as what had happened recently to GameStop and AMC.

Site Hosts

- Stock Predictor (ccc-gh.github.io)
<https://ccc-gh.github.io/FinalProject-Stock/>
- GitHub Code Repository
<https://github.com/CCC-GH/FinalProject-Stock>



Key Data Sources

Used:

- [yfinance](#) - # Yahoo! Finance market data
`$ pip install yfinance`
`import yfinance`
- [pandas-datareader](#) – Tweet data feed
`$ pip install pandas-datareader`
`import pandas_datareader`
- [Reddit](#) – user-generated bulletins
<https://www.reddit.com/r/datasets/>
- [Kaggle](#) – WallStreetBest posts
<https://www.kaggle.com/datasets>

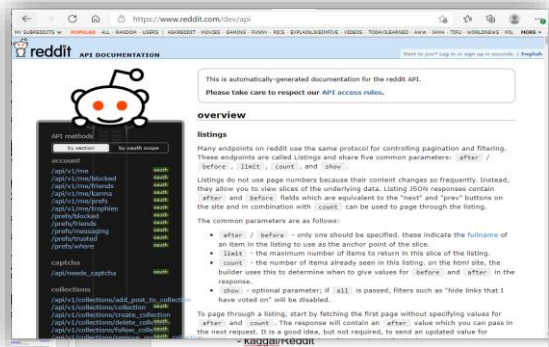


Reviewed:

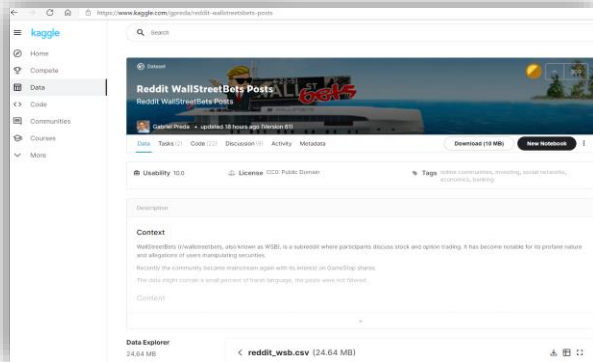
- [snowflake](#) – great data cloud connectivity store/retrieve, but no access given to stock-shorting data.

Social Media and Cloud Data

Reddit – <https://www.reddit.com/r/datasets/>

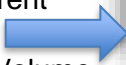


Kaggle –
WallStreetBest posts



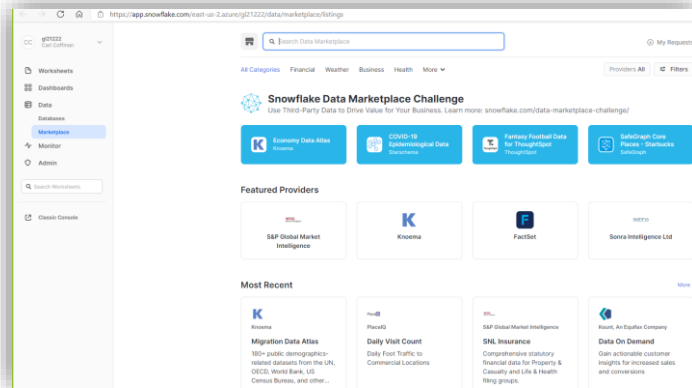
Data Challenges

- No significant, historical warehouse for social media data.
- Finding “free” historical data source for a particular stock’s short activity – current (.info) and basic-historical is made available; Open, High, Low, Close, Volume



Date	Open	High	Low	Close	Adj Close	Volume
2020-01-02 00:00:00	158.78	160.73	158.33	160.62	158.936	22622100
2020-01-03 00:00:00	158.32	159.95	158.06	158.62	156.957	21116200
2020-01-06 00:00:00	157.08	159.1	156.51	159.03	157.363	28813700
2020-01-07 00:00:00	159.32	159.67	157.32	157.58	155.928	21634100
2020-01-08 00:00:00	158.93	160.8	157.95	160.09	158.412	27746500

[snowflake](#) – Cloud Data/warehouse & Market Data



Libraries Used

from sklearn:

- `linear_model import LinearRegression`
- `tree import DecisionTreeRegressor`
- `model_selection import train_test_split`
- `metrics import mean_squared_error`

***ARIMA** - A popular and widely used statistical method for **time series forecasting**.

- `from statsmodels.tsa.arima_model import ARIMA`

Other:

- `import pandas as pd`
- `import numpy as np`
- `import matplotlib.pyplot as plt`
- `from pandas.plotting import lag_plot`
- `from pandas.tseries.holiday import USFederalHolidayCalendar`
- `from pandas.tseries.offsets import CustomBusinessDay`
- `from pandas_datareader.data import DataReader`
- `import yfinance as yf`
- `import datetime`

Linear Regression - Time Series Forecasting

***ARIMA** - AutoRegressive Integrated Moving Average:

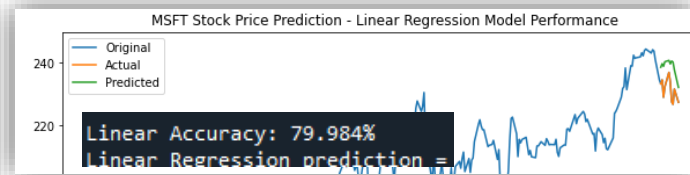
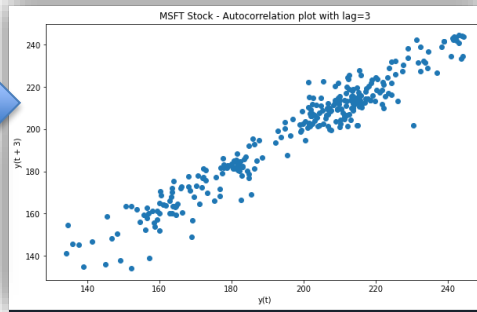
- **AR:<Auto Regressive>** uses the dependent relationship between an observation and some predefined number of lagged observations (also known as “time lag” or “lag”)
- **I:<Integrated >** model employs differencing of raw observations (e.g. it subtracts an observation from an observation at the previous time step) in order to make the time-series stationary
- **MA: < Moving Average >** model exploits the relationship between the residual error and the observations

Model parameters - expect as input parameters 3 arguments (p,d,q)

- **p** is the number of lag observations (*using 4 lags*)
- **d** is the degree of differencing (*not-stationary, days, so placed “1” in model*)
- **q** is the size/width of the moving average window

ARIMA / Linear Regression

ARIMA (or Box-Jenkins method) is a good model to be applied to this type of data (there is auto-correlation in the data).



ARIMA Test - 75% data split test model to train/fit model *10.2 MSE

```
train_data, test_data = df[0:int(len(df)*0.75)], df[int(len(df)*0.75):]
training_data = train_data['Close'].values
```

ARIMA MSE - *10.2 MSE

**note: avg squared value across all the test set predictions*

Testing Mean Squared Error is 10.164232398775123

order: p=4 lags, d=1 (not stationary)

```
time_point = range(n_test_observation)
model = ARIMA(history, order=(4,1,0))
model_fit = model_fit(diagnostics)
```

```
futureDF['Predict'] = model_fit.forecast(steps=futureDays)[0]
```

ARIMA Model Results

Dep. Variable:	D.y	No. Observations:	295
Model:	ARIMA(4, 1, 0)	Log Likelihood	-851.658
Method:	css-mle	S.D. of innovations	4.340
Date:	Tue, 09 Mar 2021	AIC	1715.316
Time:	17:48:27	BIC	1737.438
Sample:	1	HQIC	1724.174

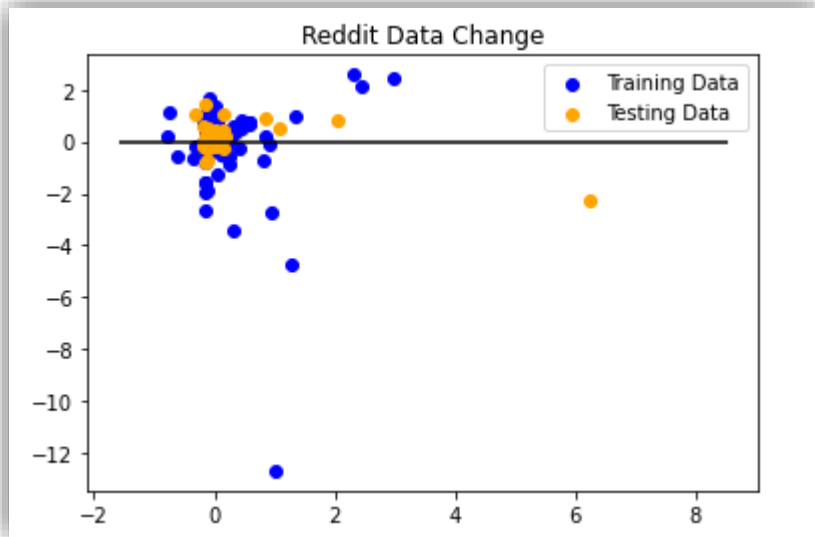
	coef	std err	z	P> z	[0.025	0.975]
const	0.2450	0.210	1.168	0.243	-0.166	0.656
ar.L1.D.y	-0.3073	0.058	-5.275	0.000	-0.421	-0.193
ar.L2.D.y	0.0593	0.061	0.976	0.329	-0.060	0.178
ar.L3.D.y	0.0632	0.061	1.036	0.300	-0.056	0.183
ar.L4.D.y	-0.0200	0.058	-0.342	0.733	-0.134	0.095

Roots

	Real	Imaginary	Modulus	Frequency
AR.1	-1.5637	-1.2121j	1.9785	-0.3951
AR.2	-1.5637	+1.2121j	1.9785	0.3951
AR.3	3.1479	-1.7014j	3.5783	-0.0789
AR.4	3.1479	+1.7014j	3.5783	0.0789

Linear Accuracy: 81.415%

Reddit Data



MSE: 0.15319122649933242, R2: 0.8150623337916859

```
X = post_data[['post_count_change', 'avg_score_change', 'Volume_change', 'Count_Score_300']]  
y = post_data['Adj Close_change'].values.reshape(-1, 1)
```

Stock Predictor Web Page

← → ↻ 🏠 🔒 <https://ccc-gh.github.io/FinalProject-Stock/>

Stock Predictor

Query Stocks

Input Ticker Symbol

Run Prediction

Stock Information

Company:

Current Price (

Open Price:

Average Volum

Volume:

52-Week High:

52-Week Low:

Analyst Rating

Date	Close	Volume	Predict	# Posts	Average Score	Over 3000
3/5/2020	3.96	33,599,500				
12/8/2020	16.94	16,120,200		80	1	0

Potential Next Steps

- Additional historical stock data such as short activity to focus on potential GameStop(GME) or AMC activity.
- If data can't be found, warehouse daily to build missing data – screen-scraping(beautiful-soup) is an option from sites such as [shortsqueeze](#) and/or [Daily Short Sale Volume](#).
- Further investigation of ARIMA model – test/track various P/D/Q configs
- Reduce time grain to hourly, minute for automated trading/prescription reporting (take automatic action on report).
- Build additional charts and dashboards to track multiple stocks.

QUESTIONS?