



EECS 6895 Project Final Presentation

Stock Performance Prediction and Recommendation with Deep Learning Analysis

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Background



Background

- An era of growing economy and financial inflation
- People always eager to earn money from stock
- Widely used big data & AI



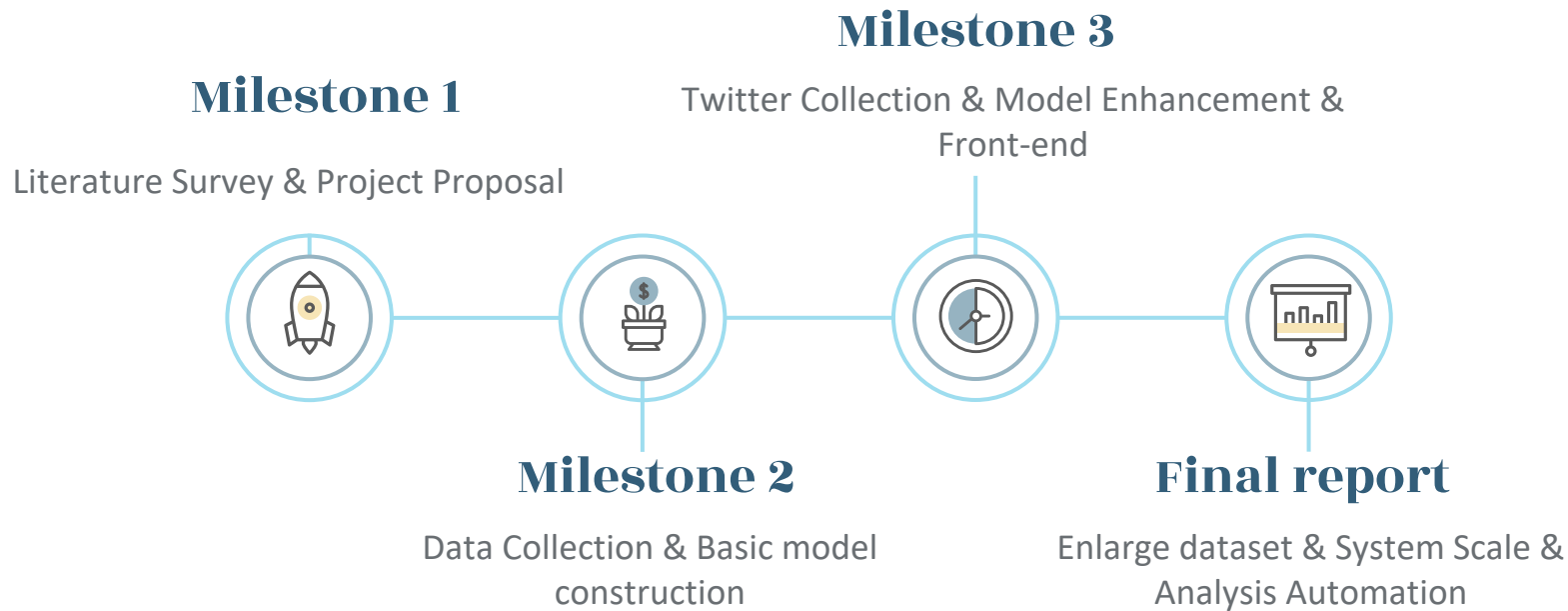
Challenges

- Stock price is influenced by multiple factors
- Stock is usually random walk
- Volume of data is significantly large

GOALS

The goal of our system is to comprehensively analyze stock and predict the stock price from both short-term and long-term view for a certain number of companies. With the help of the cutting-edge deep learning models, it could serve as a tool for investors to make their investment decisions.

Timeline Progress



Literature Survey

Financial analysis

Technical analysis: use pure historical data to predict stock market

Fundamental analysis: combine with financial factors like tweets and news

Time-series Deep Learning model

LSTM: -dependency reflection
-prediction accuracy
-smaller model update cycle

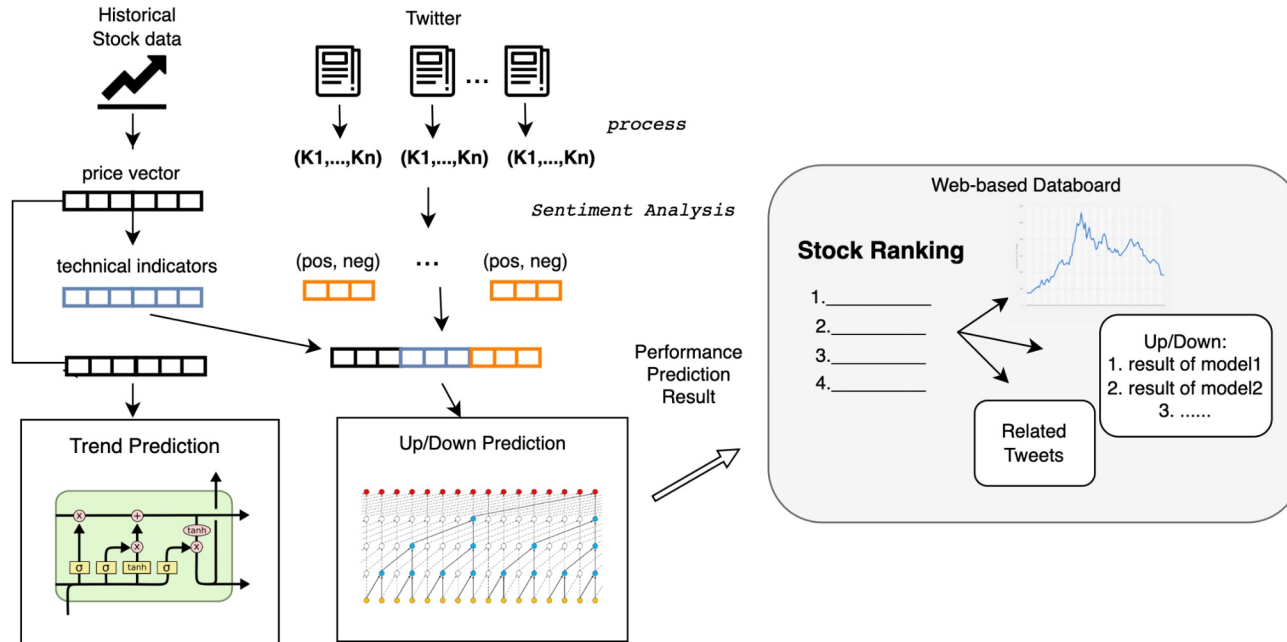
TCN: up-to-date genetic architecture

Sentiment in Behavior Economics

The nature of the market trend is impacted by sentiment analysis in the stock market.

Twitter: rich information about the companies

System Design



Supporting Dataset

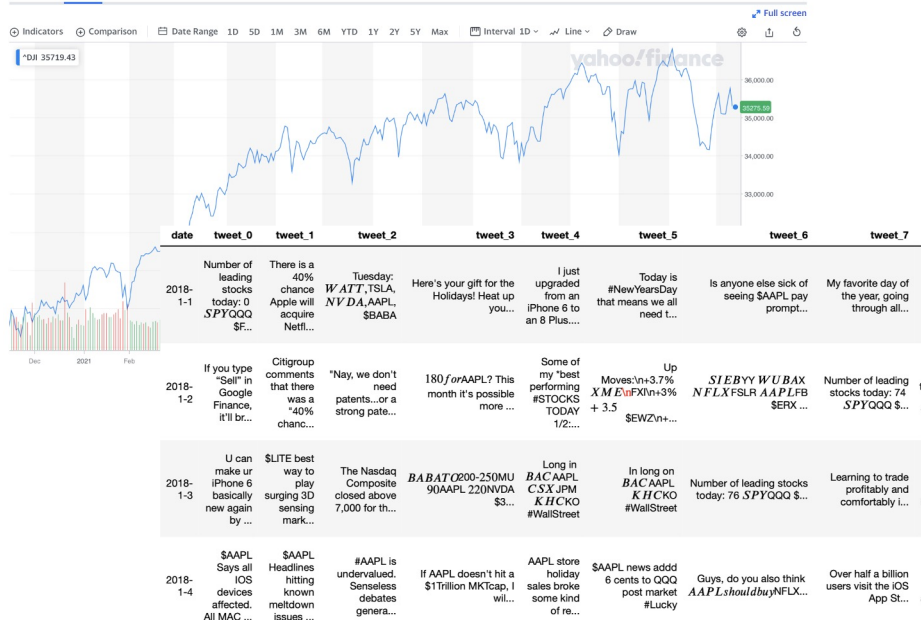
Dow Jones Industrial Average (^DJII)

☆ Add to watchlist

35,273.30 +31.71 (+0.09%)

As of 11:17AM EST. Market open.

Summary Chart Conversations Historical Data Options Components



Yahoo! Finance

Historical stock data (2018-2022):
price & volumes

Calculate technical indicators

Twitter:

sntwitter.TwitterSearchScraper API
(2018-2022): 60 tweets per day

Extract Sentiment

Methodology - Long-term Trend Prediction

Goal: Predict next 15 day stock price (For a long-term general trend)

Model: LSTM (3-layer with dropout)
Predict on: Previous 30-day open price

Roll-over prediction: involve new prediction in next-day prediction

Performance:

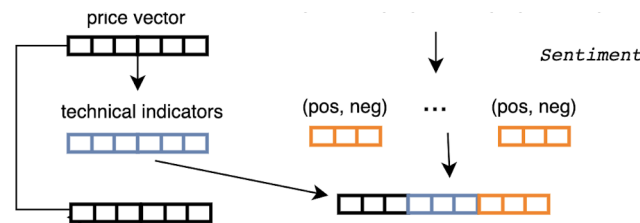
- Observe Prediction Trend vs. Actual Trend



Methodology - Next-day Up/Down Prediction

Data encoder: history driven + knowledge driven

- **stock price vector:** open, close, ...
- **+ technical indicators:** Exponential Moving Average (EMA), Average Direction Index (ADX) ...
- **+ sentiment factors**
 - VADAR model
 - Compound score ≥ 0.05 --> positive
 - Compound score ≤ -0.05 --> negative
 - Positive proportion = $\#pos / (\#pos + \#neg)$
 - Negative proportion = $\#neg / (\#pos + \#neg)$



Methodology - Next-day Up/Down Prediction

Goal: Predict whether the stock will go up or go down in the next day (For a short-term quick judgement)

Model: Self-made ensemble model

Model involvement: traditional machine learning models (ex. SVM), LSTM, TCN model

Voting decision: voted based on models with acc>75%;

Class definition:

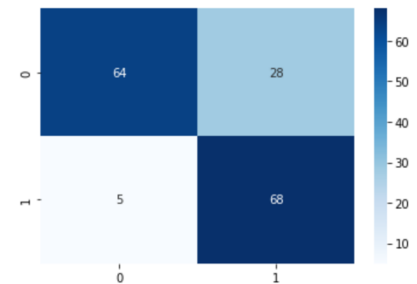
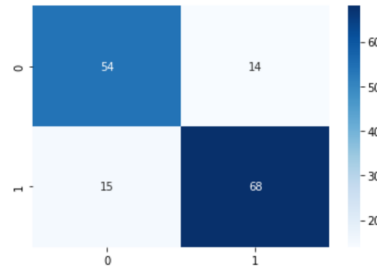
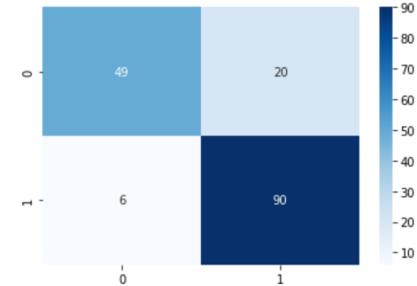
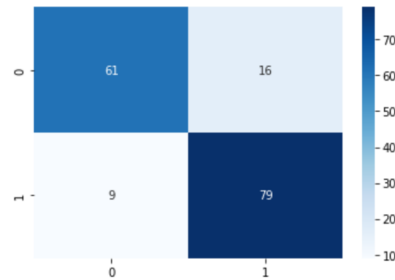
$$y_i = \begin{cases} 1, & p_{i+n} \leq p_i \times (1 + \alpha) \\ 0, & p_{i+n} > p_i \times (1 + \alpha) \end{cases}$$

Performance: Different stock model has different prediction power due to stock's predictability

- Accuracy score: ~80% accuracy (at worst >60%)
- Confusion Matrix

Methodology - Next-day Up/Down Prediction

Model	Accuracy
KNN	0.7784
SVM	0.8352
Random Forest	0.8466
Logistic Regression	0.8125
Naive Bayes	0.7613
LSTM	0.8182
TCN	0.8636



Collective Prediction

Historical-driven trend prediction

- Period prediction => General trend for a 15-day window
- Investment plan

Up/Down class prediction

- Next-day prediction => Immediate up or down tomorrow
- Investment decision today

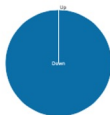
Stock	Current Price	Predicted Trend in 15 days	Up/Down in next day
AMZN	2297	22.6787	Up
SONY	84.43000030517578	16.097	Down
INTC	44.4900016784668	4.1981	Down
GOOG	2310.3798828125	4.1624	Up
CSCO	49.1500015258789	1.0209	Up
IBM	135.47000122070312	0.28200000000000003	Up
AAPL	156.00999450683594	-1.1186	Up
MSFT	274.80499267578125	-1.6162	Up
^GSPC	4128.169921875	-2.1418	Up
FB	207.33999633789062	-4.8091	Down
DELL	46.75	-7.2161	Down
TCEHY	44.400001525878906	-8.289399999999999	Down
HPQ	37.349998474121094	-9.4594	Down

(The analysis is based on date 2022-05-06)

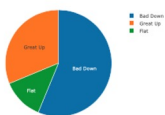
Here, we provide you with a comprehensive stock analysis and predictions with focus on famous international companies. We applied the state-of-art machine learning techniques to predict the future trend and next-day up/down situation. The analysis is based on both fundamental analysis and some event-driven factors by the means of twitter collection. The overview page lists the conclusive analysis and visualizes the general market situation. The company stock details list the detailed situation for each company, including which models we used, which twitter content is related to, etc. Currently, we provide analysis for S&P 500 and twelve famous technical companies: Google (GOOG), Meta (FB), Amazon (AMZN), Microsoft (MSFT), Apple (AAPL), IBM (IBM), DELL, DELLW, SONY (SONY), Intel (INTC), Tencent (TCEHY), HP (HPI), Cisco (CSCO)

Stock	Current Price	Predicted Trend in 15 days	Up/Down in next day
AMZN	2460	22.6787	Down
SONY	87.0599957589375	16.007	Down
INTC	45.599998474121094	4.1981	Down
GOOG	2404.4009121109375	4.1824	Down
CSCO	51.1500015258789	1.0209	Down
IBM	136.46000971386722	0.28500000000000003	Down
AAPL	163.85000610351562	-1.1186	Down
MSFT	285.5400085449219	-1.6162	Down
"GSPC	4270.43217578125	-2.1418	Down
FB	218.7400054931641	-8.8091	Down
DELL	48.6500015258789	-7.2161	Down
TECHY	46.11000061035156	-2.893999999999999	Down
HPQ	38.9300030517578	-9.4594	Down

Distribution of Up/Down in next-day



Distribution of Trend Up/Down in future 15 days



GOOG

Performance

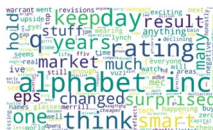
12/31/2018 → 05/05/2022

☐ Include Rangeslider

Up/Down Prediction under different models

Model	Accuracy	Up/Down prediction
TCN	0.8060606060606059	Down
LSTM	0.8	Up
Random Forest	0.8	Down
Logistic Regression	0.7939393939393941	Down
SVM	0.8060606060606059	Down
Naive Bayes	0.7151515151515152	Down
KNN	0.7676767676767676	Down

Related Topics



Current analysis is for 12 technical companies:

Google, Meta, Amazon, Apple, Microsoft, IBM, Dell, Intel,
Tencent, Cisco, Sony, HP

Two pages: Overview & Company stock details

Installation: (GitHub will be public soon)

clone <https://github.com/amayaqing/stock-prediction-with-deep-learning-analysis.git>

```
$ python3 app.py
```

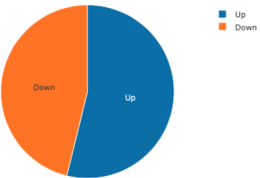
Automatically update to today's prediction when opening the dashboard

Overview

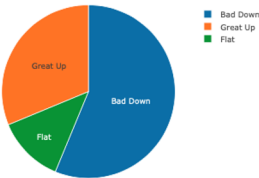
Stock Analysis and Prediction Overview

Stock	Current Price	Predicted Trend in 15 days	Up/Down in next day
AMZN	2297	22.6787	Up
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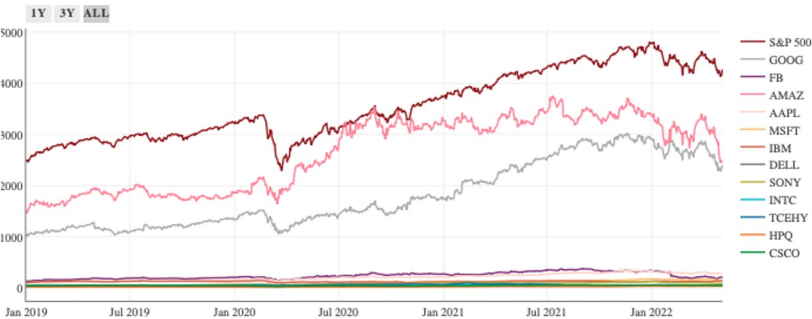
Distribution of Up/Down in next-day



Distribution of Trend Up/Down in future 15 days



The Whole Picture



TCEHY

Up/Down Prediction under different models

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Logistic Regression	0.7939393939393941	Down
SVM	0.8060606060606059	Down
Naive Bayes	0.7151515151515152	Down
KNN		

Prediction Price



Company Stock Details

Select the company stock interested

Related Tweets



System is easily scalable!



**Six
Company stock**

["GOOG", "FB", "AMZN",
"MSFT", "AAPL", "IBM"]



**Twelve company
stocks**

["GOOG", "FB", "AMZN", "MSFT",
"AAPL", "IBM", 'DELL', 'SONY', 'INTC',
'TCEHY', 'HPQ', 'CSCO']



**Future:
More...**

This system serves as a
start

Conclusion

Conclusion

- Build a system that comprehensively analyze and predict the stock price from both short-term and long-term view
- Historical-driven & Knowledge driven
- Easily scalable system --> serve for future extension

Discussion and Possible Future enhancement

- More twitter could be better
- More twitter involvement if any better way
- More element-driven, ex. company correlation, financial report.....

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THANKS!

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