Novel Approaches to Sentiment Analysis for Stock Prediction

Team 1

Agrim Rawat 2020102037

Jewel Benny 2020102057

KNV Karthikeya 2020102003

Mudit Gaur 2020112008

Abstract—Stock market forecasts are a suitable fit for a machine learning system due to their quantitative nature. A supervised learning model can combine technical information and qualitative sentiment from news to predict the course of stock movement. Information is encoded into real vectors of fixed length. We test numerous models to encode qualitative sentiment information into characteristics and to generate a final forecast about the direction of a particular stock given encoded news and technical aspects, which may be either up or down.

I. Introduction

In the world of finance, stock market forecasting has long been a contentious and important topic. According to some theorists, the stock market is fundamentally unexpected because stock prices, which they contend reflect all available information, support the efficient-market hypothesis. Others have tried using fundamental research, technical analysis, and, more recently, machine learning, to forecast the market.

II. BASIC CONCEPTS

A. Support Vector Machine

Support vector machine is a supervised learning algorithm that is used to draw a decision boundary between n-dimensional data to generate classes. The algorithm takes the farthest points of the data to draw the decision boundary, which is a hyperplane. These data points are known as support vectors, hence the name.

B. Bi-directional Long Short Term Memory

Neural networks form a major portion of the deep learning algorithms and are able to mimic the structure of the human brain. It comprises of building blocks known as perceptrons which are connected to other perceptrons with associated weights and thresholds within the network. It comprises of an input layer, output layer, and hidden layers in between them. BLSTM is a type of a recurrent neural network that contains feed-back and feed-forward loops.

C. Random Forests

It is an ensemble learning method that builds multiple decision trees and outputs the class that recieves the most votes or the average of values returned by the regression operations. They are an improvement over a single decision tree to reduce the tendency of decision trees to overfit the training data.

III. METHODOLOGY

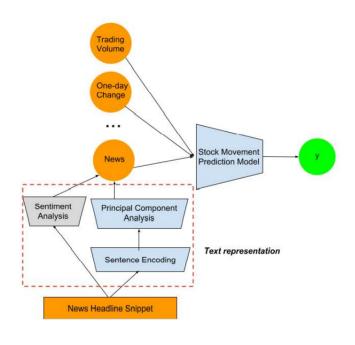


Fig. 1. Model for stock prediction

A. Data Collection

We use stock data collected from Yahoo Finance. This includes opening, closing, highest lowest price of stock and the volume of trading done for the stock over the period of 1 year. The timeframe used was 1^{st} November 2021 to 1^{st} November 2022.

B. Sentimental Analysis

As the headlines would effect the mind of a trader, this may lead to them to buy or dump the stock which would drive the price of stock up or down. Since, it is well known that many buyers are easily influenced by the actions of others and in this case the news headlines play a major role on their perception of the stock of the company.

Keeping in mind the above, to improve our predictions we are including sentimental analysis. We are using the data

headlines and snippets of news articles related to the company which is scraped from google news and finding their respective sentiment scores.

C. Prediction Models

We have predicted the trend of the stocks with and without sentiment analysis for three well-known companies, Meta, Apple and Amazon.

- 1) Without Sentiment Analysis:
- SVM
- Random Forest
- Neural Network
- 2) With Sentiment Analysis: BLSTM

IV. RESULTS

A. Sentiment Analysis

The timeline for this is 1 year on the following:

- META
- AAPL
- AMZN
- MSFT

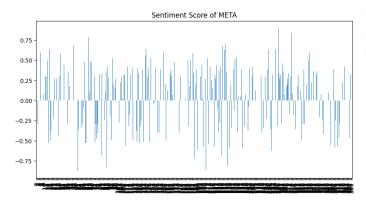


Fig. 2. Sentiment Analysis for META

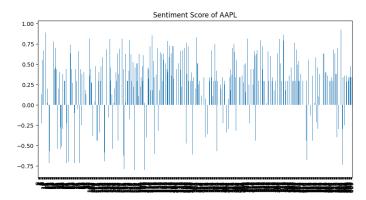


Fig. 3. Sentiment Analysis for AAPL

B. Prediction without Sentiment Analysis

1) SVM:

- Ticker Used is META (Meta).
- Accuracy = 62.745098039215684%
- Mean Absolute Error = 0.7450980392156863

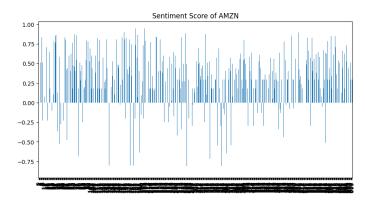


Fig. 4. Sentiment Analysis for AMZN

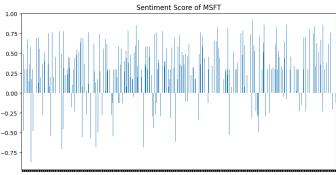


Fig. 5. Sentiment Analysis for MSFT

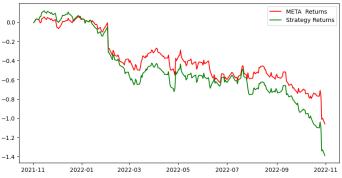


Fig. 6. Actual and Predicted returns of META using SVM

2) Random Forest:

- Ticker Used is AAPL (Apple).
- Accuracy = 99.28980978391753%
- Mean Absolute Error = 1.7931902601707042
- 3) Neural Network:
- Ticker Used is AMZN (Amazon).
- Accuracy (with tolerance 10) = 92.97658862876254%
- Mean Absolute Error = 5.015725789660196

C. Prediction with Sentiment Analysis

Bi-LSTM:

- Accuracy = 89.65028355387523%
- Mean Absolute Error = 18.17380680884666

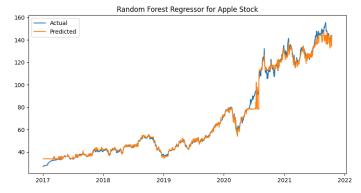


Fig. 7. Actual and Predicted returns of AAPL using Random Forest

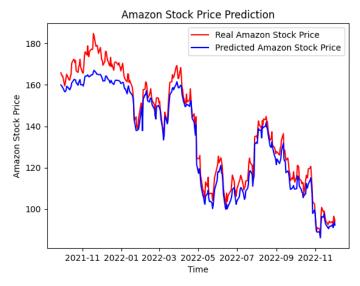


Fig. 8. Actual and Predicted returns of AMZN using Neural Network

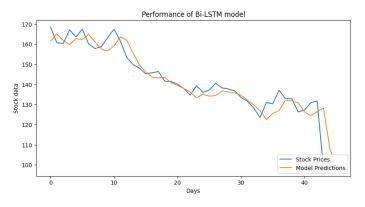


Fig. 9. Actual and Predicted returns of META using BLSTM

V. CONCLUSION

The BLSTM model with sentiment analysis gives better prediction quality than the rest since it not only predicts on the basis of the raw stock prices but also on the basis of the news sentiment which directly effects the buyers. The news headlines play a major role on their perception of the stock of

Model	Accuracy	Mean Absolute Error
Without Sentiment Analysis		
SVM	62.75%	0.7451
Random Forest	99.23%	0.7931
Neural Network	92.98%	5.0157
With Sentiment Analysis		
BLSTM	89.65%	18.1738

the company.

- The stock predictions are very sensitive to the trading data provided.
- The random forest and NN-LSTM models seem to overfit the input data.
- The model for a specific company ticker is not transferable to a different company since the stock price behaviour changes from company to company.
- The sentiment-based Bi-LSTM seems to give better all round results even though the MAE is high.

Github Repo: https://github.com/grokebloke/Stonks-Team1

ACKNOWLEDGMENT

We would like to thank our Professor Mr. Anoop Namboodri for giving us the opportunity to do this project. This project was a profound learning experience for all of us and we would like to thank him for allowing us to choose our topics too. We would also like to thank our TA Sanjai Kumaran for helping and guiding us whenever necessary and providing us with valuable feedback.

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