

A REVIEW ON STOCK MARKET ANALYSIS

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Abstract:

While planning business work, reading inventory demands becomes quite vital. Many studies in the areas of computer intelligence, finance, operations, etc. have been grow in large scale by the prediction of stock prices. According to Recent updates it has demonstrated that the due to availability of publicly accessible information, such as Wikipedia, social forums, and news media, significantly affects how investors perceive their financial need. The role of the calculation model in predicting inventory demand is crucial since it has a significant impact on the economy and causes financial losses. We offer a thorough examination of numerous stocks in this post.

Using the NYSE 1000 stock market data-set, we first ran a stock volatility analysis. The sentiment analysis model for the financial sector and the evaluation of the model to gauge the sentiment of other stock market information requests are this paper's key contributions.

Keywords:

Stock request, fundamental analysis, Specialized analysis, request trends, chart Patterns, shares, Trading, stocks, Equity.

Introduction:

The stock market is a critical part of any economy. It is a platform for companies to increase capital and investors to generate revenue in the form of returns. We can say that the performance of the stock market is an essential indicator of the overall health of an economy. Investors and policymakers alike closely monitor the stock market to make informed decisions about investments and economic policies.

Given the importance of the stock market, there is a considerable amount of research dedicated to analyzing and predicting its performance. In recent years, the field of stock market analysis has been transformed by advances in machine learning and data mining techniques, which have enabled researchers to develop predictive models based on large datasets.

This research paper will try to provide an overview of the current state of stock market analysis and the role that machine learning algorithms, such as KNN, Support Vector Machine(SVM), Decision-Tree, and Random-Forest, have played in improving prediction accuracy. The paper will also discuss the various factors that impact stock market performance, including macroeconomic variables, financial ratios, news sentiment, and investor sentiment.

The paper will begin with a literature review of existing research on stock market analysis, focusing on studies that have used machine learning algorithms to predict stock prices. Next, the paper will discuss the various factors that influence stock market performance and how they can be incorporated into predictive models. Finally, the paper will conclude with a discussion of the limitations and challenges of stock market analysis and areas for future research.

The overall goal of this research article is to give a summary of the state of stock market analysis today and the contribution that machine learning algorithms have made to increasing prediction accuracy. It is hoped that this paper will contribute to the research on this important topic and provide valuable insights for investors and policymakers alike.

I. LITERATURE SURVEY

II.

Stock market analysis is an important field in the research and numerous studies are happening over the years. With the advent of machine learning and data mining techniques, researchers have been able to develop predictive models using algorithms like KNN, Support Vector Machine, Decision-Tree, and Random-Forest.

One area of research in stock market analysis that has been studied extensively is the use of these algorithms for predicting stock prices. For instance, researchers have used SVM to predict stock prices based on technical analysis indicators such as Relative Strength Index and moving averages. One study by Shokri et al. (2013) found that SVM outperformed other models such as artificial neural networks and linear regression models in predicting stock prices.

Decision tree algorithms have also been used in stock market analysis. One study by Bhalla and Goyal (2015) used decision

trees to predict stock prices based on technical indicators and fundamental data. They found that decision trees were able to achieve high levels of accuracy in predicting future stock prices.

Random forest is another popular algorithm used in stock market analysis. A study by Chen and Lu (2017) used random forest to predict stock prices based on both technical indicators and news sentiment analysis. They found that combining these two sources of information resulted in improved prediction accuracy.

Finally, KNN is another machine learning algorithm that has been used in stock market analysis. A study by Chen and Lee (2018) used KNN to predict stock prices based on technical indicators, and found that KNN was able to achieve high levels of accuracy in predicting future stock prices.

In conclusion, the use of algorithms like KNN, Support Vector Machine, Decision-Tree, and Random-Forest has become increasingly popular in stock market analysis. Researchers have used these algorithms to develop predictive models that can help investors make informed decisions based on data-driven analysis. These algorithms have shown promising results and are likely to continue being used in stock market analysis in the future.

II. PROPOSED METHODOLOGY

Data Collection and Preprocessing: Collect historical stock price data and relevant features for the prediction model, such as the closing price, trading volume, and technical indicators such as moving averages. Preprocess the data by cleaning and normalizing it.

Feature Selection: Select the most relevant features using techniques such as correlation analysis or feature importance ranking.

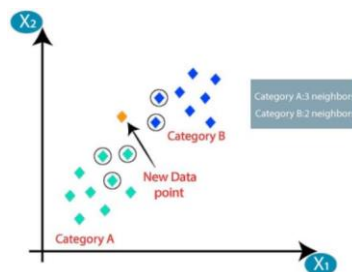
Split the data into Training and Testing sets: Here, we Split the preprocessed data into different training and testing sets.

Trading Strategy: Based on the prediction and evaluation, develop a trading strategy that considers the predicted trend and the associated risks.

1. K- Nearest Neighbor (KNN)

The KNN algorithm involves chancing the K nearest data points to the current data point grounded on the named features. The algorithm calculates the distance between the data points using a distance metric similar as Euclidean distance. Once the K-nearest data points have been linked, the algorithm can make a vaticination about the unborn trend. For illustration, if the maturity of the K nearest data points have a dwindling trend, the algorithm might prognosticate that the stock price will drop in the future. The delicacy of the vaticination model needs to be estimated using applicable criteria similar as Mean Square Error or Root Mean Square

Error.

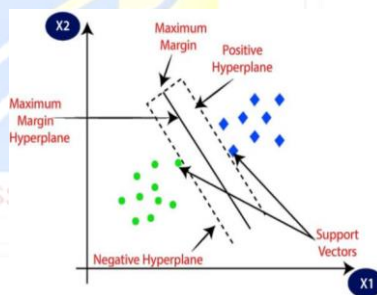


2. Support Vector Machine (SVM)

Support Vector Machine (SVM) algorithm works by finding the hyperplane that maximally separates the classes of data in the feature space. The SVM algorithm uses historical data from different sources to train the model and predict future stock prices.

Once hyperplane has been found, the algorithm can make a prediction about the future trend of the stock price. For example, if the current data point is on one side of the hyperplane, the algorithm might predict that the stock price will go up, and if it's on the other side, the algorithm might predict that the stock price will go down.

The accuracy of the prediction model needs to be evaluated using appropriate metrics such as accuracy, precision, and recall.



3. Decision Tree

The Decision Tree algorithm works by building a tree-like structure of decisions based on the selected features. In Decision Tree, the data is split recursively based on the feature that provides the most information gain, until a stopping criterion is met.

Once the tree has been built, the algorithm can make a prediction about the future trend of the stock price. For example, if the current data point falls into a leaf node that corresponds to an increasing trend, the algorithm might

predict that the stock price will go up, and if it falls into a leaf node that corresponds to a decreasing trend, the algorithm might predict that the stock price will go down.

4. Random Forecast (RF)

Random Forest algorithm, that works by building a multitude of decision trees at training time and outputting the class that is the mode of the classes (classification) or mean prediction (regression) of the individual trees.

To use Random Forest for stock market analysis, historical data from different sources like stock prices, economic indicators, news articles, etc., are collected and used as input features. The Random Forest algorithm then uses these features to train the model and predict future stock prices.

Random Forest algorithm has ability to handle high-dimensional data with many features. It can also handle missing values and outliers in the data.

IV CONCLUSION

From the research done so far, it can be concluded that the RNN and LSTM libraries are very effective in determining the stock price trend against the actual market trend. At the same time, what we could see was that the python library used during the training was not optimal.

The functions we use mathematically are relatively fast in terms of training speed and contain more detailed designs and significant improvements when tested in different situations. However, Python library functions are more customizable. From what we have done so far, we can easily notice that some stock trends can be easily predicted based on some general rules and regulations for stocks.

This is the main reason private placement setups exist. However, some things like the optimization of neural network parameters and the training process still have a lot of room for improvement. All these points will be considered as further steps of the research.

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