



Forecasting Stock Price Using Machine Learning

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Abstract: This study aims to investigate how machine learning and deep learning algorithms predict stock values. The project will include collecting and analysing historical stock price data, as well as training a number of machine learning models, including linear regression and random forest, in order to forecast future stock prices. The research will also examine the use of deep learning techniques including neural networks, long-short term memory, and gated recurrent units in order to increase the accuracy of stock price predictions. The results of this experiment may provide financial experts and investors with vital knowledge on the accuracy and effectiveness of different machine learning and deep learning models in predicting stock prices.

Keywords: Machine Learning, Deep Learning, Neural Networks, Long-short term Memory, Gated recurrent unit, linear Regression, Random Forest, Stock Market.

I. INTRODUCTION

Over the years, the stock market has remained an enigma; a mercurial sector in the fiscal landscape that eludes even those with incisive insight. Yet modern advances have made forecasting its prices more possible than ever before. Machine learning algorithms such as regression, LSTM and GRU have been deployed to great effect, parsing voluminous datasets to detect patterns and realize better predictions. As technology serves up ever richer sources of information, these same techniques will continue to be utilized with greater frequency - indeed their potential is considered nigh limitless when it comes to transforming contemporary finance. In this article we shall further explore this concept, tracing how these algorithms are used for forecasting as well as evaluating both their assets and limitations accordingly. Firstly, let us consider the implications of stock market forecasting, as well as its potential applications. Machine learning algorithms could be used to help inform the decisions of traders, investors, and companies with an eye to market trends. For example, the use of such algorithms in corporate taxation could be designed to maximize deduction deductions, identify advantageous investments, and generate data-driven recommendations.

Meanwhile, amongst individual traders, algorithms can be an invaluable tool in portfolio prediction and triage. Indeed, machine learning can be used to identify the most profitable stocks, and reconfigure a portfolio accordingly. Additionally, it could be used to predict the direction of a stock before an earnings report, analyse news stories for their potential impact, and generate informed predictions. In this sense, machine learning can operate much like a weather forecast, apprising its readers of market conditions and enabling them to take anticipatory action.

At the same time, there are certain limitations that machine learning algorithms for forecasting stock markets must contend with. Due to the fact that the stock market is constantly in flux, no single metric or equation can be relied upon to determine veritable patterns; only short-term future guesses. Additionally, there are also moral considerations to take into account, such as the question of whether or not algorithms can ever truly reflect human sentiment or cognitive bias. Does the risk of an algorithm becoming too powerful, far outweigh any benefits it can offer?

Despite such issues, stock market forecasting algorithms are still widely employed and are likely to remain so in the years to come. As developers continue to refine their methods by plugging in new data sources and tweaking their algorithms accordingly, so too will their accuracy increase. However, precisely how useful such techniques will be in the professional sphere remains to be seen; until then, only time will tell

II. LITERATURE SURVEY

The strategy that Ishita Parmar and Navanshu Agarwal offer in their paper The aim of stock market prediction is to forebode the value of a company’s financial stocks in the future. Machine learning technology for stock market forecasting is a relatively innovation [2]. Using historical data to train, this technology generates predictions based on the values of current indexes of the stock market. Machine learning use a range of models to help and validate prediction. The focus of the study is on how to estimate stock values using LSTM and regression-based machine learning. Considerations include openness, closeness, low, high, and volume. In machine learning, the dataset is essential. The dataset should be as accurate as possible because even little changes in the data could have a big impact on the outcomes.

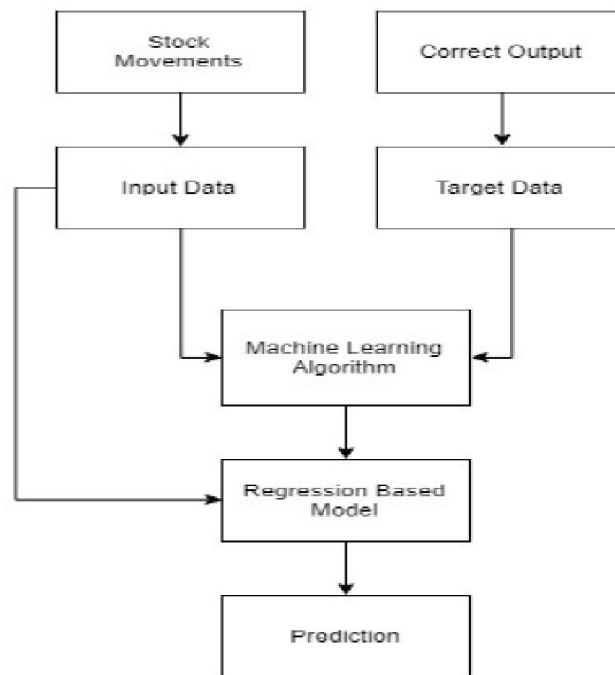


Figure 1: Regression Based Model

This study employs supervised machine learning on data gathered from Yahoo Finance. This dataset consists of the open, close, low, high, and amount variables. The open, close, low, and high bid prices for the stock are all distinct and have nearly identical names. The volume is indicator of how many shares were traded overall within the specified period. The test data are then used to evaluate the model.

The stock market or share market is one of the most intricate and sophisticated ways to transact business, according to Kunal Pahwa and Neha Agarwal. Small ownerships, brokerage companies, and the banking sector all depend on this one entity to make money and distribute risks, making it a very complex model [3]. However, in order to make this erratic business model a little more predictable, this study recommends using machine learning algorithms to forecast future stock prices for exchange by utilizing open-source libraries and existing algorithms. We’ll see if the outcomes of this simple execution meet our needs. The outcome is completely decided by math, which also makes a number of real-world assumptions that might or might not be true at the time of prediction.

The Stock Market is one of the first places where a normal person can trade stocks, make investments, and benefit from companies that use this platform to sell a piece of themselves. This tactic shows its potential as an investment scheme when used cautiously.

In statistics, by looking at the values and characteristics of an issue depicted as a graph, we can identify the dependent and independent variables and try to establish or recognize an existing link between them. This commonly used

strategy, known in statistics as linear regression, is relatively simple to use and has a high level of efficiency. The similar method has been tailored for machine learning, where the characteristics are used to train the classifier, which subsequently predicts the value of the label with a given accuracy that can be verified during training and testing of the classifier.

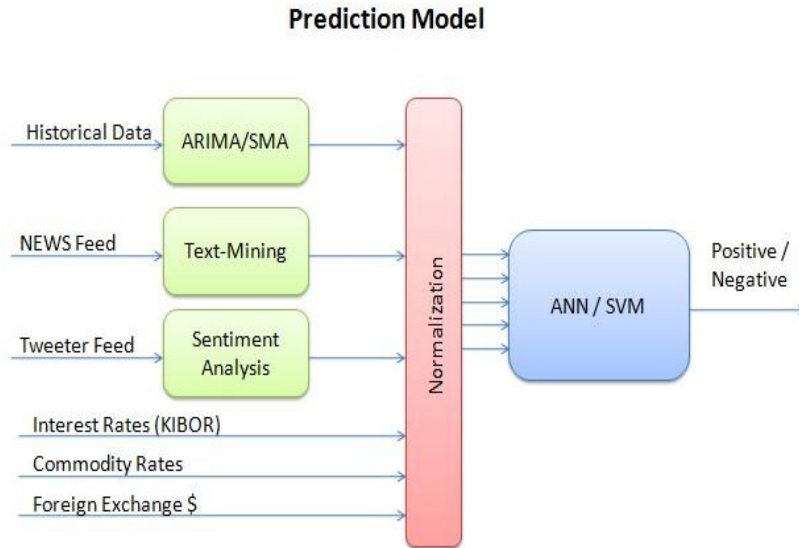


Figure 2: Prediction Model

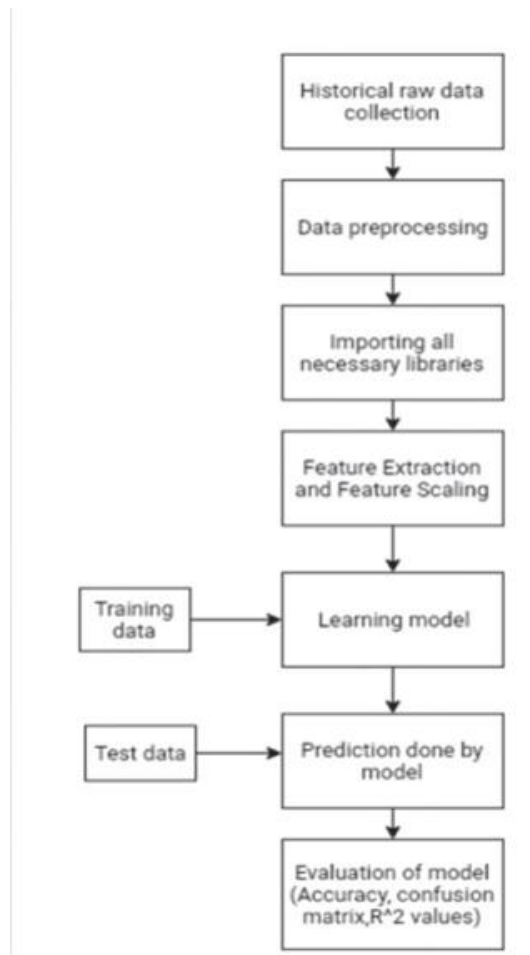
For the past ten years, researchers have been utilizing the neural network, one of the intelligent data mining techniques. Today's economy relies heavily on stock market data prediction and analysis. Linear models (AR, MA, ARIMA, ARMA) and non-linear models (ARCH, GARCH, Neural Network) encompass the various forecasting algorithms. For the purpose of predicting a company's stock price based on available historical prices, we employ four distinct deep learning architectures in this paper: Multilayer Perceptron (MLP), Recurrent Neural Networks (RNN), Long Short-Term Memory (LSTM), and Convolutional Neural Network (CNN). We are utilizing the day-by-day closing prices of two distinct stock markets, the New York Stock Exchange (NYSE) and the Indian National Stock Exchange (NSE). The network was trained with the stock price of one NSE-listed company and predicted the stock price of five NSE-listed and NYSE-listed companies [4]. CNN has outperformed the other models, according to observations. Despite having been trained with NSE data, the network was able to predict for the NYSE. Because the inner dynamics of the two stock markets are similar, this was possible. Neural networks outperform the existing linear model (ARIMA), as evidenced by comparison of the obtained results to the ARIMA model.

Since simple models are incapable of correctly forecasting future asset values due to the abnormality of the financial market, investing in a group of assets has never been straightforward. Machine learning, which is the practice of teaching computers to perform tasks that typically take human intelligence, is the most recent trend in scientific research. The objective of article is to develop a model for forecasting future stock values using Recurrent Neural Networks (RNN), more particularly the Long-Short Term Memory model (LSTM)[5]. The primary objective of this paper is to investigate the forecasting power of machine learning algorithms and the extent to which epochs can enhance our model. These algorithms outperform linear regression algorithms in terms of effectiveness and multicollinearity. The use of machine learning techniques in finance is the topic of numerous studies at the moment. Some of these studies used tree-based models to forecast portfolio returns, while others employed deep learning to predict the future values of financial assets. Additionally, some writers reviewed the AdaBoost algorithm's forecasting of returns.

III. PROPOSED SYSTEM

So many elements have yet to be considered and because forecasting the stock market initially doesn't seem statistical, it appears to be a complex task. But with wise use of using machine learning techniques, it is possible to compare historical data to the present data and teach the computer to draw the proper conclusions from it.

- The input raw data extracted from Yahoo finance.
- For feature extraction of 'date' and 'closing price' of a stock the attributes used.
- Features used to predict the momentum of stock price of company are 'stock momentum', 'index volatility', 'sector momentum'. These features are scaled.
- The dataset is splited into training and test data set.
- The training dataset is used for model training and test dataset is used for prediction. The significance of a feature is determined using the R2 values.
- The values of the test data are predicted and the results are evaluated. The result is given based on accuracy, confusion matrix and time required for the model used in this.
- Collecting stock data from NSE (National Stock Exchange), BSE (Bombay Stock Exchange) and Kaggle.com.



Stock market prediction plays a significant role in the financial industry as it provides investors with insights into the future trends of the stock market. Accurate predictions can help investors make informed decisions, reduce risks, and increase profitability. Furthermore, stock market predictions can also be used by financial institutions and governments to make policy decisions and improve the overall stability of the market.

Machine Learning Algorithms for Stock Market Prediction:

A. Regression

A statistical technique known as regression is used to show that a dependent variable and one or much more independent variables have a connection. In order to create a mathematical equation that predicts the dependent variable, regression analysis aims to estimate the coefficients of the independent variables. Regression enables the identification of the effects of changes in one variable on other variables.

B. Random Forest

In order to produce more precise predictions, the Random Forest ensemble machine learning method builds numerous decision trees and then combines their outputs. Each decision tree in Random Forest is trained using a random subset of both the training data and the feature set. This method aids in lowering overfitting and enhancing the model's capacity for generalisation. Combining all of the decision trees' projections yields the ultimate conclusion.

C. LSTM

The long short-term memory, or LSTM. It is a particular kind of recurrent neural network (RNN) created to get around the difficulties encountered by conventional RNNs in addressing long-term dependencies. A memory cell is used by LSTM to store data over time, and three gates—input, output, and forget gates—are used to regulate the flow of data into and out of the cell. This design, which has been extensively employed in speech recognition, natural language processing, and other sequence-to-sequence applications, enables LSTMs to acquire and retain long-term dependencies.

D. GRU

Recurrent neural networks (RNNs) of the Gated Recurrent Unit (GRU) variety are comparable to LSTM (Long Short-Term Memory) networks. GRU is intended to solve the drawbacks of conventional RNNs in handling long-term dependencies, much like LSTM. GRU is quicker and simpler to train than LSTM since it is more computationally efficient and uses fewer parameters.

A GRU network's basic structure is made up of a number of recurrent hidden units, each of which takes input from the previous time step in addition to the current input. GRU, on the other hand, differs from conventional RNNs in that it has gating mechanisms that regulate the flow of data into the network, enabling it to selectively remember or forget data over time.

An update gate and a reset gate are normally present in a GRU unit. The reset gate defines how much of the new input should be used to update the hidden state, while the update gate indicates how much of the prior hidden state should be kept. An update gate-weighted combination of the prior hidden state and the fresh input makes up a GRU unit's output. Speech recognition, NLP, and other sequence-to-sequence applications have all made extensive use of GRU. It has demonstrated strong performance in a range of applications and has emerged as a well-liked LSTM substitute in many deep learning models.

V. CONCLUSION

The initiative seeks to connect predictions made by machine learning models to retail investors in order to forecast stock market patterns using deep learning and machine learning technologies. With extra analysis, it aids investors in navigating the stock market and assists them in reaching more informed decisions. The results showed that the program offers value in trend prediction. The prediction displays a helpful trend tendency with the actual stock trend when compared to the baseline. The models used in the program will keep getting better by using an evolution algorithm to find better model topology, structure, and hyperparameters. The research showed that the evolution algorithm is effective at reducing the mean squared error when predicting stock prices, which benefits retail buyers by enhancing trend prediction.

The goal is to develop a user-friendly system that allows retail investors to navigate machine model predictions results with helpful benchmarks despite their lack of prior technical expertise. To assist investors in understanding the implications of stock price predictions, such as when to purchase or sell, a better presentation of stock price predictions could be created.

This would truly democratize deep learning and machine learning technologies, which were previously thought to only be in the hands of a very small number of people, and enable investors to make more informed decisions based on the Deep learning models.

REFERENCES

- [1] Adil MOGHAR, Mhamed HAMICHE, “Stock Market Prediction Using LSTM Recurrent Neural Network”, *Procedia Computer Science* 170 (2020) 1168–1173, International Workshop on Statistical Methods and Artificial Intelligence (IWSMAI 2020) April 6-9, 2020, Warsaw, Poland.
- [2] Ishita Parmar, Navanshu Agarwal, Sheirsh Saxena, “Stock Market Prediction Using Machine Learning”, 978-1-5386-6373-8/18/\$31.00 ©2018 IEEE 2018 First International Conference on Secure Cyber Computing and Communication (ICSCCC).
- [3] Kunal Pahwa, Neha Agarwal, “Stock Market Analysis using Supervised Machine Learning” 978-1-7281-0211-5/19/\$31.00 2019 ©IEEE, 2019 International Conference on Machine Learning, Big Data, Cloud and Parallel Computing (Com-IT-Con), India, 14th -16th Feb 2019.
- [4] Hiransha M, Gopalakrishnan E. A, Vijay Krishna Menon, Soman K.P, “NSE Stock Market Prediction Using Deep-Learning Models”, *Procedia Computer Science* 132 (2018) 1351–1362, International Conference on Computational Intelligence and Data Science (ICCIDS 2018).
- [5] David M. Q. Nelson, Adriano C. M. Pereira, Renato A. de Oliveir, “Stock market’s price movement prediction with LSTM neural networks”, 978-1-5090-6182-2/17/\$31.00 ©2017 IEEE.
- [6] G. Shobana, K. Umamaheswari, “Forecasting by machine learning techniques and econometric s: A review2021”, IEEE Xplore Part Number: CFP21F70-ART; ISBN: 978-1-7281-8501-9, Proceedings of the Sixth International Conference on Inventive Computation Technologies [ICICT 2021].
- [7] Ji sang park, ji sung lee, dong Jim Kim, “Forecasting daily stock trends using random forest optimization”, 978-1-7281-0893-3/19/\$31.00 ©2019 IEEE.
- [8] Ryota kato, tomoharu nagao, “Stock market prediction based on interrelated time series data”, 978-1-4673-1686-6/12/\$26.00 ©2012 IEEE.
- [9] Srinath Ravikumar, Prasad Sara, “Prediction of Stock Prices using Machine Learning (Regression, Classification) Algorithms”, 978-1-72816221-8/20/\$31.00 ©2020 IEEE, 2020 International Conference for Emerging Technology (INCET) Belgaum, India. Jun 5-7, 2020.
- [10] Yash Mehta, Atharva Malhar, Dr. Radha Shankarmani, “Stock Price Prediction using Machine Learning and Sentiment Analysis” 978-1-72817029-9/21/\$31.00 ©2021 IEEE, 2021 2nd International Conference for Emerging Technology (INCET) Belgaum, India. May 21-23, 2021.
- [11] Mehak Usmani, Syed Hasan Adi, Kamran Raza and Syed Saad Azhar Ali, “Stock Market Prediction Using Machine Learning Techniques”, 978-1-5090-2549-7/16/\$31.00 ©2016 IEEE, 2016 3rd International Conference on Computer and Information Sciences (ICCOINS).