

PREDICTING STOCK MARKET TRENDS USING MACHINE LEARNING AND DEEP LEARNING ALGORITHM

Ananya Divakar Goudar

Department of ISE
Vemana Institute of Technology
Bangalore, India

Hema K S

Department of ISE
Vemana Institute of Technology
Bangalore, India

Inchara T R

Department of ISE
Vemana Institute of Technology
Bangalore, India

Meghana Kalmat

Department of ISE
Vemana Institute of Technology
Bangalore, India



CrossMark

Publication History

Manuscript Reference No: IRJCS/RS/Vol.09/Issue08/SPAUCS10105

Research Article | Open Access

Peer-review: Double-blind Peer-reviewed

Article ID: IRJCS/RS/Vol.09/Issue08/SPAUCS10105

Volume 2022 | Article ID SPAUCS10105 <http://www.irjcs.com/volumes/Vol09/iss-08/25.SPAUCS10105.pdf>

Received: 25, July 2022 | Accepted: 04, August 2022 | Published: 12, August 2022

doi: <https://doi.org/10.26562/irjcs.2022.v0908.25>

Citation: Ananya,Hema,Inchara,Meghana(2022).Machine Learning and Deep Learning Algorithm. International Research Journal of Computer Science (IRJCS), Volume IX, Issue VIII, 2022, Pgs 281-285

BibTeX key:

Academic Editor-Chief: Dr.A.Arul Lawrence Selvakumar

Copyright: ©2022 This is an open access article distributed under the terms of the Creative Commons Attribution License; which Permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited

Abstract: Intention of StockMarket Prediction is to forecast future worth of a company's monetary stocks. utilising machine learning produces forecasts based on the values of current stock market indices Using their prior values as training data is a new development in stock market prediction technology. Predicting the performance of the stock market is one of the most challenging tasks.Prediction involves a huge number of variables, such as the distinction between physical and psychological factors, rational and irrational conduct, and more. Share prices are unpredictable and challenging to forecast accurately as a result of a combination of all these variables. Many researchers have conducted studies on the upcoming market developments movement. Data is a key source of efficiency because stock is made up of dynamic data. The prediction's efficiency has an effect on the same chances. To make accurate predictions, machine learning employs a variety of models. Machine learning techniques have the potential to uncover previously unknown patterns and insights using features like an organization's most recent announcements, quarterly sales statistics, and so forth, which can then be utilised to produce impeccably correct forecasts. The project's goal is to predict stock values using machine learning that is based on long short-term memory (LSTM) and regression. All of the variables—open, close, low, high, and volume are taken into account.

Keywords: Stock prediction, Machine Learning, Regression, LSTM, open, close, low, high.

I. INTRODUCTION

The phrase "stock" has likely appeared in one form or another to all of us. Particularly stock is connected to business partners and organisations that are preparing to establish themselves in the marketization sector. Another word for stock that is frequently used in ordinary speech is share. It is known as an investment strategy, and it is regarded as a long-term investment that will safeguard and provide adequate funds for retirement. A fraction of a corporation can be bought by purchasing its shares. People make the same investments to gain a long-term advantage that they believe is less valued right now but has the potential to increase in value over time. An investment addresses long-term goals with reasonable aims over the long term. The value of the share you invest in now must provide you with the best yield tomorrow, but that yield is not the same.

The resources and forces employed to drive the market off or on the set, as well as the market itself, are unpredictable. It has never been at the same level, and the trend will likely continue to be unpredictable for some time. A few approaches for prediction and closeness have been devised, and rough values and estimates are reproduced with the best intentions. All resources, though, are unreliable and unpredictable by nature. The greatest technique to determine reliability is to be aware of the state of the market and perform study into it. Many agents have made a fortune using this method. They provide predictions and offer advice, but their fees are greater and the stock evaluation is never less than uniform. Even during a single day, the market might experience numerous highs and lows depending on the resources available to external and internal agents as well as their timing. A fascinating starting point is stock. The fair share or ownership representation that explains the security measures and the agreement between two parties a person and a company is referred to as stock. Since the beginning, the term "stock" has been a fancy term because of its propensity for ambiguity. People who were investigating the same thing and used it every day had made a fortune from it. To help you comprehend and invest in the same, a variety of agents are available on the market, but their fees are laborious and exorbitantly high.

The primary resources for the company are the funds used to carry out daily operations and generate a profit. When there is a need for higher budget estimation and to overgrow from the resources, they require finance, and it is stressful to apply for, receive, and have a financing loan since the interest rate is higher than that of other forms of investment, reducing the product margin. Another approach for a business to raise money, boost output for a higher yield, and ultimately make the most of the business strategy is through the sale of stock. This has been shown to be a more advantageous method than dealing with the financial crisis during a time of necessity to invest and advance in the business world. For an investor, it is a risk phenomenon where they invest their saving and hope it brings back the return in a higher yield. If the evaluation of the same increases, then the stock evaluation and its price increase causing financial gain to both parties. In Indian culture, it is even regarded as a side issue of business and is viewed as a stroke of luck. A person becomes a shareholder when they buy stock in a company and receives a share of the firm's profit or gain as compensation for their investment. According to their demands, investors can buy and sell stocks. They can share their shares with each other or other people, and there are numerous stockbrokers playing with the same in the company.

II RELATED WORK

Many different fields, such as text classification, image processing, speech recognition, and many more, have embraced deep learning for classification and prediction problems. A promising field of research is presently deep learning. Since then, it's been utilised to forecast time-series data. Randomness, nonlinearity, and unpredictable data can all be handled with it [1]. For instance, Chong et al. [6] used deep feature learning to extract features from a data model for predicting the stock market that collects information without referencing earlier research, from the stock return time series a working understanding of the predictors, and tested it on high-frequency data from the Korean stock exchange. Recently, many different time-series analysis tasks have been carried out using LSTM networks, which are effective at learning temporal patterns [7]. LSTM is preferred over conventional RNN because it can efficiently learn long-term dependencies utilising memory cells and gates and addresses the issue of disappearing (or bursting) gradients. As a result, several studies on financial time series modelling include LSTMs.

Zhou et al. [2] employed GAN to predict stock prices on the China Stock Exchange market for particularly volatile equities. As far as we are aware, GAN hasn't been widely utilised on financial markets elsewhere in the globe, and certainly not on the Indian Stock Market. In this investigation, we contrast the effectiveness of LSTM and GAN models used to stock price forecasting in the Indian market. A. LSTM Dependencies that span extremely long times can be taught to the LSTM. It goes beyond vanishing. A sophisticated architecture known as the LSTM replaces a standard neuron by switching the gradients problem that a general RNN encounters. Unit or block [3]. The equations shown below can be used to summarise how the LSTM

$$\begin{aligned} \text{works.zt} &= \tanh(Wzxt + R zht-1 + b z) & it &= \sigma(Wixt + R iht-1 + b i) & ft &= \sigma(Wfxt + R fht-1 + b f) \\ ot &= \sigma(Woxt + R oht-1 + b o) \\ st &= zt \cdot it + st-1 \cdot ft & ht &= \tanh(st) \cdot ot \end{aligned}$$

where it denotes the input gate and ot the output gate. The forget gate, memory cell, and concealed state are denoted by the letters ft, st, and ht, respectively. The definitions of the functions and tanh are found in numbers two and three, respectively.

$$\begin{aligned} \sigma(z) &= \frac{1}{1 + e^{-z}} \\ (2) \tanh(z) &= \frac{e^z - e^{-z}}{e^z + e^{-z}} \\ (3) \text{ B. GAN} \end{aligned}$$

Goodfellow et al. presented generative adversarial networks [4], where picture patches are produced from employing two concurrently trained networks to generate random noise. It has shown excellent outcomes when used for a variety of purposes. In particular, the creation of images. GAN are members of the family models, generative. GAN models discover the provided collection of distribution (training data) and generate data that is comparable to dispersion of original data. The model has two output options: samples that either represent or come from a distribution.

In the past, training models needed a dataset with

$$\min G \max D (EX \sim P_{data}(x) [\log D(x)] + EZ \sim P_{data}(z) [\log(1 - D(G(z)))])$$

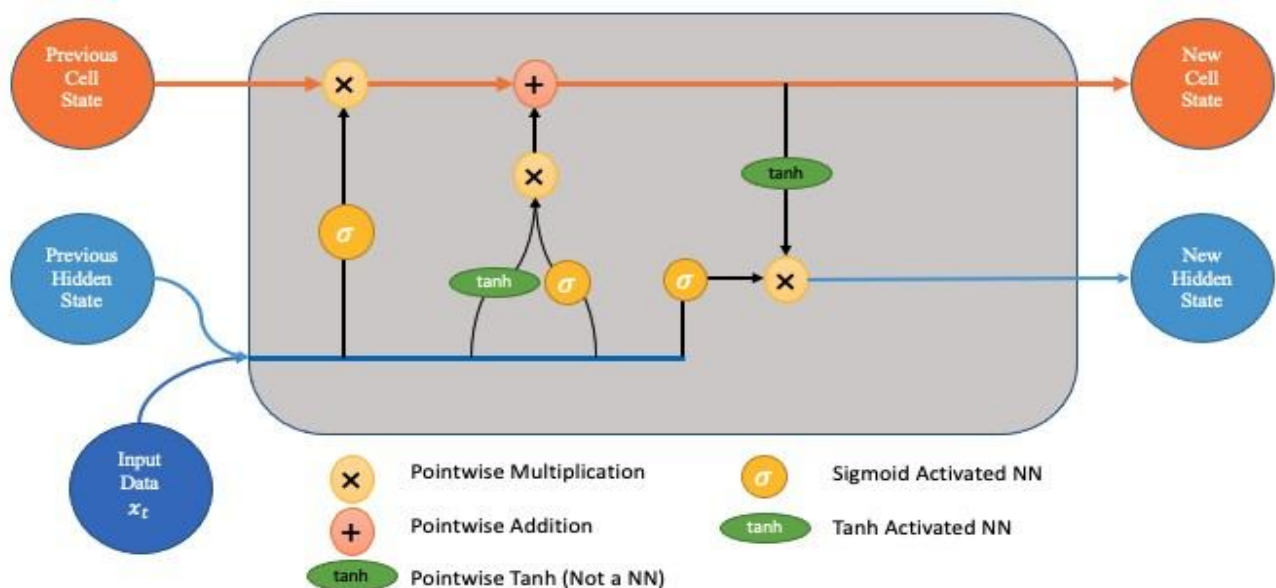
Where the predicted values for the P_{data} and P_z distributions, respectively, are x and z . Stochastic Gradient Descent can be used to solve issues (SGD). This optimization problem (see 5). The improvement is difficult, and there ought to be many potential options (sets of weights). Consist of global best. Local optima, on the other hand, perform poorly but are a little easier to find.

$$W_{t+1} = W_t - \eta \partial L \partial W \quad (5)$$

W is the set of weights, while W_{t+1} and W_t are sets of weights at iterations $t + 1$ and t , respectively. L is the cost function, and is the learning rate. The step size, or the degree to which the weights in the model change to enter the space of solutions the data are all labelled. To overcome the limitation imposed by ignorance, generative models can be trained using a dataset that combines labelled and unlabeled data. A Generator and a Discriminator are the two models that make up the generative adversarial network. The generator's goal is to produce samples that appear to have been taken from a distribution that corresponds to our target distribution using input noise, which frequently has a uniform distribution. The Discriminator then predicts which samples are real and which ones are fake based on the samples produced by our Generator (fake data). Afterwards, samples are chosen from a distribution that mimics our target distribution (actual data). The output of the discriminator is as follows : (Generator 1) The Generator function is a function G that is differentiable with respect to its parameters. G , along with the input data z , where z is the noise resulting from an allocation P_z , such as a Gaussian distribution, The power source then converts z to samples $G(z)$, which are then used to create the generator's distribution P_g , which is anticipated to be the same as the actual data distribution P_{data} after training. The graphic After using samples $G(z)$ to train the discriminator, the generator is told to mislead the discriminator by giving the generated samples high probability. 2) Discriminator: In a similar way, it is possible to define the discriminator function D as differentiable with respect to its parameters. D should be given x and $G(z)$, where x stands for the learning rate. It is the most important hyper parameter to adjust in order to achieve good performance on the task. In this session, you will explore the learning goals.

III. PROPOSED SYSTEM

The two broad groups of prediction techniques are statistical techniques and artificial intelligence techniques. As examples, consider the logistic regression model, the ARCH model, and other statistical methods. Artificial intelligence approaches include multi-layer perceptrons, convolutional neural networks, naive Bayes networks, back propagation networks, single-layer LSTMs, support vector machines, recurrent neural networks, etc. A network with long short-term memory was used. Network of long-term memory: The Long Short-Term Memory is a form of recurrent neural network. LSTM procedure: The special network structure known as LSTM is composed of three "gate" components. The input gate, forgetting gate, and output gate are the three gates that make up an LSTM unit. When information enters the LSTM network, it may be selected by rules. Information that does not comply with the algorithm will be erased by the forgetting gate, leaving only the data that does.



The historical data that were collected from the Internet and used as experimental data in this study. The experiments made use of three data sets. Finding an optimization algorithm with a quicker convergence speed and less resource requirements is necessary. First, the output of an LSTM is fundamentally dependent on three factors at any one time:

- (1) The cell state is the current long-term memory of the network.
- (2) The output at the preceding instant or the previous concealed state,
- (3) The input information for the present time step **LMS filter**:

The LMS filter is a specific type of adaptive filter used to address linear issues. The purpose of the filter is to discover the filter coefficients by minimizing the least mean square of the error signal in Order to minimize a system.

Algorithm LMS:

Input:

x : input vector
d: desired vector
u: learning rate
N: filter order

Output:

y: filter response
e: filter error

begin

```
M = size(x) ;
xn(0) = w, (0) = [0.0 ... 0];
while n < M do
X(n+1) = (x(n); xn(1: ND);
y(n) = we *Xp3
e(n) = d(n) —y(n);
W(n+1) = Wat 2pe(n) Xn;
```

End

end

Typically, we evaluate both a linear and a non-linear algorithm because we are unsure if the problem can be solved effectively with a linear approach. We will use LMS to demonstrate that stock market prediction can be done with linear algorithms with a reasonable degree of precision since the internet always depicts non-linear techniques. However, because this filter resembles a system, if we apply it on our data first, the filter coefficients will have learned to function as predicted when we add new data.

SYSTEM ARCHITECTURE

An orderly description of the structure, behaviour, and system perspectives is called system architecture. The system architecture outlines the key elements, their connections, organisational structures, and interactions. Business strategy, quality attributes, and many other aspects are included in the architecture and design of software.

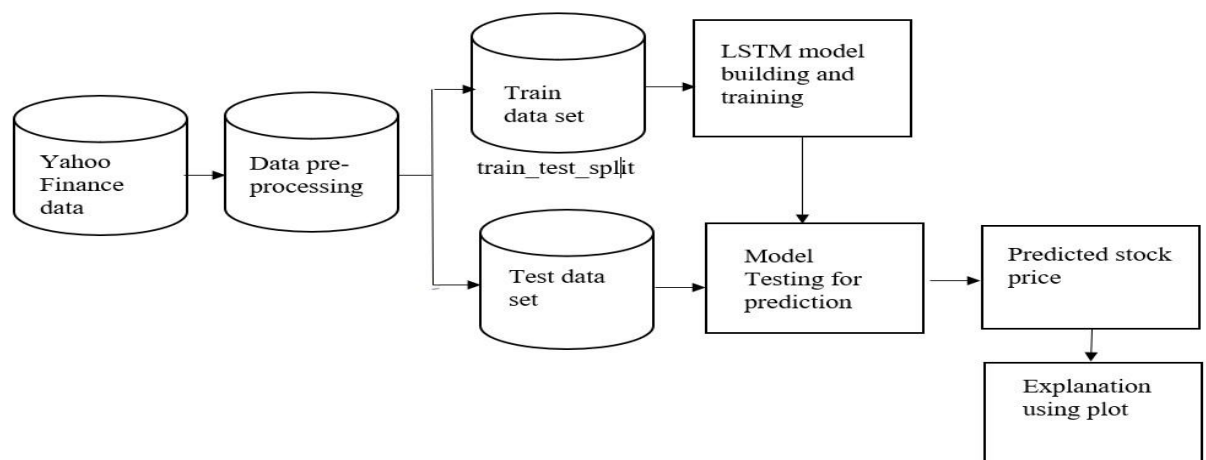


Figure 1: System Architecture

- Fig. 1 above depicts the system's overall view. It displays the various modules used to construct the system, such as
- Data
- Data Wrangling
- Applying Algorithm

- Training Model
- Testing Model
- Evaluating the model

The flow of the data in the system is as follows:

- It takes the input from the finance website and pre-processes it so that it can be fed to the algorithm (LSTM).
- The input is fed into the LSTM and the model is trained.
- After training, the model is tested on different inputs.
- The models are then used to forecast the closing price of the stock, which is displayed graphically.

CONCLUSION AND FUTURE WORK

Stock market prediction is the process of attempting to predict the future value of business stock or other financial instruments traded on an exchange. Accurately predicting the future price of a stock could yield a big reward. The efficient-market hypothesis holds that since stock prices accurately reflect all publicly available information, any price changes that are not based on newly disclosed information are inherently unpredictable. This machine learning method might develop into the technology that supposedly makes it possible for everyone to understand about future prices. This is an attempt to utilise machine learning techniques to more precisely and consistently anticipate how much a company's shares will cost in the future. The key contribution of the researchers is the application of the novel LSTM Model as a way of computing stock prices.

FUTURE WORK

This is an attempt to utilize machine learning techniques to more precisely and consistently anticipate how much a company's shares will cost in the future. The researchers' main contribution is their use of the innovative LSTM Model to compute stock prices.

REFERENCES

1. I. K. Nti, A. F. Adekoya, and B. A. Weyori, "A systematic review of fundamental and technical analysis of stock market predictions," *Artificial Intelligence Review*, pp. 1–51, 2019.
2. I. Goodfellow, J. Pouget-Abadie, M. Mirza, B. Xu, D. Warde-Farley, S. Ozair, A. Courville, and
3. Y. Bengio, "Generative adversarial nets," in *Advances in neural information processing systems*, 2014, pp. 2672–2680.
4. X. Zhou, Z. Pan, G. Hu, S. Tang, and C. Zhao, "Stock market prediction on high-frequency data using generative adversarial nets," *Mathematical Problems in Engineering*, vol. 2018, 2018.
5. Stock Market prediction using Supervised Machine Learning Techniques: An Overview Zaharaddeen Karami Lawal: Hayati Yusuf Zakari.
6. Indian Stock Market Prediction using Deep Learning Ayan Maiti: Pushparaj Shetty D