

---

# Stock Market Close Price Forecasting Using LSTM And Candle Stick Charting

---

<sup>1</sup>R.Sujatha,<sup>2</sup>R.Ramesh,<sup>3</sup>M.Ponraviraj,<sup>4</sup>K.Hariharan,<sup>5</sup>B.Saran,<sup>6</sup>M.Saravanakumar

<sup>1</sup> Assistant Professor, Department of Information Technology, M Kumarasamy College of Engineering, sujathar.it@mkce.ac.in

<sup>2</sup> Assistant Professor, Department of Computer Science and Engineering, K.Ramakrishnan College of Engineering, rameshcse@krce.ac.in

<sup>3,4,5,6</sup> UG Students, Department of Information Technology, M.Kumarasamy College of Engineering, ponraviraj0206@gmail.com

## ABSTRACT:

Stocks that are on an exceptionally fundamental level related with each other will regularly move together. Considering such typical examples is acknowledged to assist with leading advancement deciding endeavors. Regardless, such signals are not piddling to show in light of the fact that the relationship among stocks are not really presented and ought to be surveyed from erratic data. Motivated by this insight, stock expense assumption accepts a basic part in building a trading system for monetary benefactors. The productive deciding of stocks' future expense will help the monetary benefactors with extending their advantage. In any case, it is difficult to anticipate unequivocally the example of the protections trade as a result of the muddled association between stock expenses and external elements like news, overall economy, general suppositions, and other tricky financial information. We propose a construction that solidifies the between relations of firm stock expense of following days share cost. Significant learning approach accepts fundamental part in estimate of money related time series data. One of the strategies to do insightful examination using time series data is long transitory memory (LSTM). The figure precision of LSTM has been assessed using three estimations - RMSE, MSE and MAE.

## KEYWORDS:

LSTM, CNN, ML, DL, Trade Open, Trade Close, Trade Low, Trade High

## 1. Introduction

Securities exchange Future Forecasting is the endeavor on unambiguous assessment, depiction, and presumption. The information given by Maruti is utilized for our analysis. [1] Apart from this information is also gathered from the money related exchanges such as monster headway stocks and the others. The open and closing stock of every day for an organization is monitored in order to predict the opening stock value of the next day. Our work aims at helping those who are interested in investing in stock market. The future value of the stock is predicted using LSTM.

## 2. Purposes of the Stock Market – Capital and Investment Income

As discussed in [2], the money related exchange is important for two reasons. First is to give money to affiliations which helps to foster the affiliations. If any stock value is less, then the affiliation pays for an undertaking bank to deal with the stock responsibility. By offering stock recommendations as opposed to getting the capital expected for increase, the affiliation avoids accomplishing responsibility and paying interest charges on that responsibility. The association keeps away from achieving commitment and paying interest charges on that commitment. For example, accepting a monetary patron buys parts of an association's stock at \$10 a deal and the expense of the stock in this manner climb to \$15 a proposition, the monetary benefactor can then get a half advantage on their theory by selling their bits.

Banks who are involved in this stock trading make use of the advice from esteem stock research specialists as discussed in the work in [3]. These people are researchers who check whether the association's stock is likely to rise or fall. Store bosses or a portfolio boss, which fuses adaptable venture executives, share save chiefs, and exchange traded hold (ETF) bosses, are critical monetary trade individuals since they exchange tremendous measures of stocks.

## Analyzing Stocks- Market Cap, EPS, and Financial Ratios

Many Trade analysts and monetary patrons could look at a grouping of factors to show how a stock reaches a

conceivable up or down in terms of cost in near future. There are many factors that help in analyzing the stock and help instockassessment. Monetarytradecapitalizationcontributes totheoverall largenumberofremarkablesegmentsofthestock. A Market capitalization that is higher contributes more segments of stock to the association even though the association is more fiscally sound.

Coming up next are two or three the key money related extents that monetary sponsor and specialists consider: Cost to Earnings (P/E) Ratio: An association stock expense is equivalent to its EPS.

Commitment to equity share

Commitment to Equity Ratio: A lower commitment to esteem extent, exhibiting fundamental supporting from monetary patrons, is ideal.

### 3 Return on Equity (ROE) Ratio

The benefit from ROE as pointed out by X. Qiu and et. al in [4] is considered to be a good indication of an association's improvement in potential as it shows the relation between the association's net addition relative to the total worth interest in the association.

By and large income: There are a couple of in general income extents that monetary benefactors should seriously think about, including working generally income and net generally income.

### Two Basic Approaches to Stock Market Investing - Value Investing and Growth Investing

[5] D. G. Gloubois There are limitless procedures for stock picking that agents and monetary patrons use, but basically all of them are some sort of the two fundamental stock buying approaches of critical worth contributing or improvement contributing. Regard contributing is more based on avoiding risk than improvement contributing is, in spite of the way that regard monetary supporters really attempt to buy stocks when they trust the stock expense to be an underrated bargain. This has been stated by D.G. Gloubois in his work. [5]

### Open High Low Close in Stocks

In trading with the stock, the words high and low indicates the limit and the smallest expenses in a given time frame. The words open and close are the expenses at which the stock value begins at the start of the day and the value of the stock at the end of a comparable period.

## 5 Time Matters

The period of time is very important for any assurance esteem as they focus on OHLCV. This indicates the open, high, low, close, volume and is stated by K.A. Althelaya [6]. But if regardless showed, the period is for the most part regular; in any case, representatives join various periods while keeping an eye on the worth movement of a security.

### "Open" and "Close" Prices

Many have the habit of going through the protection trade fragment of the morning paper everyday. We see as our main stock, yet see there is some difference from one worth near it. The header on one of them is "open". Open means the expense at which a stock started trading while the underlying ringerrang.

$$\text{Block input } (z^t): \quad z^t = g(W_z x^t + R_z y^{t-1} + b_z) \quad (1)$$

$$\text{Input gate } (i^t): \quad i^t = \sigma(W_i x^t + R_i y^{t-1} + b_i) \quad (2)$$

$$\text{Forget gate } (f^t): \quad f^t = \sigma(W_f x^t + R_f y^{t-1} + b_f) \quad (3)$$

$$\text{Cell state } (c^t): \quad c^t = z^t \odot i^t + c^{t-1} \odot f^t \quad (4)$$

$$\text{Output gate } (o^t): \quad o^t = \sigma(W_o x^t + R_o y^{t-1} + b_o) \quad (5)$$

$$\text{Block output } (y^t): \quad y^t = h(c^t) \odot o^t \quad (6)$$

### "High" and "Low" Prices

The terms "high" and "low" cost are considered to be the possible fuse by the financial periodicals and destinations. The former word indicates the most excessive expense at which a stock was traded during a period. The later word indicates the most diminished cost of the period. The high and the discouraged spots of a stock for the day is called as

intraday high and low. This is stated in [7] E. Cambria and et al.,

The qualification between the stocks open and close apportioned by the opening value is the stocks return or its presentation in terms of its rate. Let us extend our view point on the stocks presentation, example one-year, we would use the end cost from a year earlier and balance it with the end cost from today to get the yearly return.

## 6 Problem Identified

In [9] K. Nagarathinam states that the financial business was one of the chief dares to embrace the use of AI and significant learning initi hypothesisassessment and exercises to upgrade their clients. Going beforeAI, significant learning, and the entirety "Quant"change in the 2000's so far, examiners and monetary supporters relied upon less precisely subordinate methodologies. Head andconcentratedassessment governedand,inspiteofthewaythattheyreallymakeupasignificantpieceoftheexamination,they'rebyandbygottotogetherwithfiguresandassessmentdo ne by PCs.

Asbyfarmostknow,the protectionstradeistheplacewherepeopleexchange stocks. Theshowofexchangingthese stocks (for instance trading) occurs in physical and virtual circumstances called "Exchanges". These exchanges are housesfor records (typically acknowledged ones are the Dow Jones Industrial Average and NASDAQ Composite). The exchanges arewheretheexpense of stocks thatmakeuptherundownsareset. The role and the functions of the exchanges are stated in the work of M.Kim in [10]

The significant learning study is a subfield of AI computations impelled by the concerns and the structure and limit of the frontal cortex; this is call fake mind association. Byfar most of the learning method uses cerebrum network designing, which is the explanation you want to focus on significant learning model is commonly referred to as a significant mind association. Here, the maxim "significant" generally speaking, implies the amount of hidaway layers of the mind association. In the standard mind network involves only 2-3 mystery layers, while the deep cerebrum networks with upto 150 mystery layers. M. Bildirici [11] and his team has explained this in their work.

In this proposed work, Long Shot Term Memory (LSTM) have been utilized for predicting the intraday closing expense for Maruti association having a spot with different areas of movement. The financial data: Open, High, Low and Close expenses of stock are used for making new factors which are used as commitments to the model. The models are surveyed using standard key MARKET:RMSE and MAPE.

### Computation Selection

LSTM does not deal with singed data, it needs progression of data for preparing the required information and to store the evident information. LSTM focuses projecting the output based on assumptions and time series data. It is comfortable to work with the backslide issue.

### A Stock Price Predictor Using LSTM

The proposed framework that learns online anticipating the close costs of the stock with the assistance of Long Short-Term Memory (LSTM). The Long Short-Term Memory (LSTM) is a phony unpredictable neural system (RNN) plan used in the field of deep learning, not in the least like standard feed forward mind systems, LSTM has input affiliations. Not at all does the procedure not revolve around single information (e.g., pictures) but moreover on full information plans, (Forexample, atalk or a video). Forexample, LSTM is material for undertakings, for instance, undivided, affirmation, talk affirmation and recognition of idiosyncrasies in coordinated busy time gridlock or IDS (interruption region frameworks).

### The Working of the proposed system

H. Yan Tedious cerebrum network is kind of Long transient memory (LSTM). Due to the design of LSTM, the issue of long stretch dependence can be agreed to irregular neural network. To hinder the information's nerve, long flitting memory is used. The outcome of network endeavors to be in indirect circle that decays or explodes through the analysis circle, and the long stretch dependence is essential point of convergence of LSTM. In this module four estimations used in the endeavors are introduced.

### Mean Absolute Percentage Error

W. Bao, Mean Absolute Percentage Error (MAPE) is oftentimes used to study the introduction of the assumption strategies. MAPE is similarly an extent of assumption accuracy for deciding strategies in the AI area, it regularly presents precision as a rate. Condition (1) shows its condition.

### LSTM Prediction model

$$MAPE = \frac{1}{n} \sum_{t=1}^n \left| \frac{A_t - F_t}{A_t} \right| \times 100; \quad (1)$$

### Mean Absolute Error

The LSTM model was verifiable Python, a huge level mind networks API in view of top of Tensor Flow, which is an open-source programming library made by Google. It uses the recently referenced dataset, with S&P 500 element over the course of the period of time of 2003-01-01 to 2021-02-12 and isolates it in 80% readiness data and 20% testing data. With

Mean out and out screw up (MAE) is an extent of the differentiation between two characteristics. MAE is an average of the difference between the assumption and the genuine characteristics. MAE is a run of the mill measure of prediction bungle for backslide assessment in the AI area. The formula is shown in Equation (2). where  $A_t$  is the real worth and  $F_t$  is the assumption regard. In the situation, the altogether worth

of the difference between those is divided by n (number of tests) and added for each anticipated worth.

$$MAE = \frac{1}{n} \sum_{t=1}^n |A_t - F_t|, \quad (2)$$

Relative Root Mean Square Error

Root Mean Square Error (RMSE) is obtained by finding the standard deviation of the assumption bungle in the backslide work. The residuals call ed conjecture bungle show the distance between the real characteristics and an assumption model and also they depict the way in which they are spread out in the model. The estimation gives a clear picture of how the data is concentrated close to the best fitting model. The result of this RMSE is the square base of the squared contrasts among the expectations and the certified discernments. Relative Mean square error is like Root Mean square error which takes the total squared botch and normalizes it by parceling with the full scale squared bungle of the indicator model. The related formula is shown in the equation.

The formula is shown in Equation (3). where  $A_t$  is the saw worth,  $F_t$  is the assumption worth and  $n$  is the amount of tests.

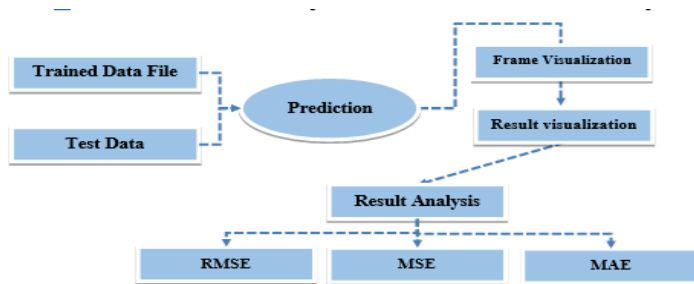
$$RRMSE = \sqrt{\frac{1}{n} \sum_{t=1}^n \left( \frac{A_t - F_t}{A_t} \right)^2}, \quad (3)$$

### Mean Squared Error

The Mean Squared Error (MSE) measures the idea of a pointers and its worth is for the most part nonnegative (regards more like zero are better). The MSE is the second depiction of the goof (about the origin), and unites both the difference in the estimate model (how by and large spread the predictions are beginning with one data test then onto the following) and its tendency (how close the typical expected value is from the insight). The formula is shown in Equation (4). where  $A_t$  is the seen worth,  $F_t$  is the estimate worth and  $n$  is the amount of tests.

$$MSE = \frac{1}{n} \sum_{t=1}^n (A_t - F_t)^2, \quad (4)$$

the picked feature(s) of the current and past number of days (identical to time adventure) as information (X), the end cost of the following day is expected as result (y).



From the above diagram it shows that prediction done by the trained and test data after the prediction result analysis of RMSE, MSE and MAE is based on Frame visualization and Result Visualization.

## RESULT

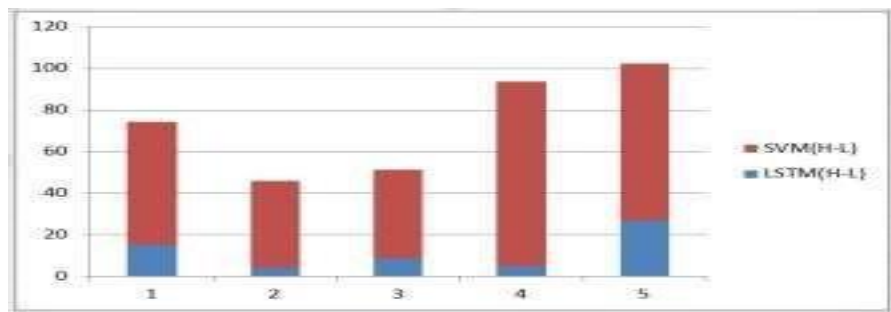


Fig1. SVM vs LSTM with high and low parameter

From the above diagram it shows that comparison on SVM and LSTM with the parameters high and low parameters LSTM provides greater accuracy.

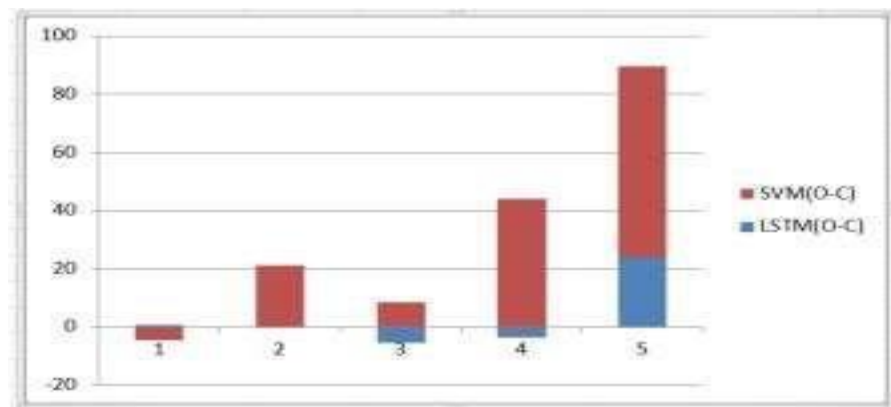


Fig2. SVM vs LSTM with open and close parameter

From the above diagram it shows that comparison on SVM and LSTM with the parameters open and close parameters LSTM provides greater accuracy.

## Conclusion and Future Enhancement

To help predict the stock index, a less error of the predictive model is needed which may take into account the processing of the input data. RNN cannot learn to connect information because old stored memory will be increasingly useless with time running due to overwritten or replaced new memory. Forecast using the LSTM method starts with entering inputs and outputs previously into the forget layer.

The future enhancement includes comparing the accuracy of LSTM with other prediction algorithm. We have taken 5 year of data and predicted in future only 1 year of data has been taken and predicted with less time facility.

## REFERENCES:

- [1] A. Altunkaynak and T. A. Nigussie, "Monthly water consumption prediction using seasonal algorithm and wavelet transform based models," *J. Water Resour. Planning Manage.*, vol. 143, no. 6, Jun. 2017, Art. no. 04017011.
- [2] P. Liang, H.-D. Yang, W.-S. Chen, S.-Y. Xiao, and Z.-Z. Lan, "Transfer learning for aluminium extrusion electricity consumption anomaly detection via deep neural networks," *Int. J. Comput. Integr. Manuf.*, vol. 31, nos. 4\_5, pp. 396\_405, Apr. 2018.
- [3] H. Z. Wang, G. B. Wang, G. Q. Li, J. C. Peng, and Y. T. Liu, "Deep belief network based deterministic and probabilistic wind speed forecasting approach," *Appl. Energy*, vol. 182, pp. 80\_93, Nov. 2016.
- [4] X. Qiu, H. Zhu, P. Suganthan, and G. A. Amaratunga, "Stock price forecasting with empirical mode decomposition based ensemble support vector regression model," in *Proc. Int. Conf. Comput. Intell., Commun., Bus. Anal. Singapore: Springer*, 2017, pp. 22\_34.
- [5] D. G. Gloubos, "Estimating corporate failure as an auditor's going concern evaluation factor," *M.S. thesis, School Bus. Admin., Univ. Macedonia, Thessaloniki, Greece*, 2016.
- [6] K. A. Althelaya, E.-S.-M. El-Alfy, and S. Mohammed, "Evaluation of bidirectional LSTM for short-and long-term stock market prediction," in *Proc. 9th Int. Conf. Inf. Commun. Syst. (ICICS)*, Apr. 2018, pp. 151\_156.
- [7] E. Cambria, "Affective computing and sentiment analysis," *IEEE Intell. Syst.*, vol. 31, no. 2, pp. 102\_107, Mar. 2016.
- [8] J. Grifith, M. Najand, and J. Shen, "Emotions in the stock market," *J. Behav. Finance*, vol. 21, no. 1, pp. 42\_56, 2020.
- [9] K. Nagarathinam and R. S. Kathavarayan, "Moving shadow detection based on stationary wavelet transform," *EURASIP J. Image Video Process.*, vol. 2017, no. 1, p. 49, Dec. 2017.
- [10] M. Kim, "Cost-sensitive estimation of ARMA models for financial asset return data," *Math. Problems Eng.*, vol. 2015, pp. 1-8, Jan. 2015.
- [11] M. Bildirci and Ö. Ö. Ersin, "Nonlinearity, volatility and fractional integration in daily oil prices: Smooth transition autoregressive ST-FI(ap) GARCH models," *Romanian J. Econ. Forecasting*, vol. 3, pp. 108\_135, 2014.
- [12] A. Lendasse, E. de Bodt, V. Wertz, and M. Verleysen, "Non-linear financial time series forecasting Application to the bel20 stock market index," *Eur. J. Econ. Social Syst.*, vol. 14, no. 1, pp. 81\_91, 2000.
- [13] E. Chong, C. Han, and F. C. Park, "Deep learning networks for stock market analysis and prediction: Methodology, data representations, and case studies," *Expert Syst. Appl.*, vol. 83, pp. 187\_205, Oct. 2017.
- [14] H. Yan and H. Ouyang, "Financial time series prediction based on deep learning," *Wireless Pers. Commun.*, vol. 102, no. 2, pp. 683\_700, Sep. 2018.
- [15] W. Bao, J. Yue, and Y. Rao, "A deep learning framework for financial time series using stacked autoencoders and long short term memory," *PLoS ONE*, vol. 12, no. 7, Jul. 2017, Art. no. e0180