

# Stock Price Prediction Web Application Using Python

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*Abstract*— The thing of stock request vaticination is to read the value of a company's fiscal stocks in the future. The use of machine literacy, which produces vaticinations grounded on the values of current stock request indicators by training on their previous values, is a new trend in stock request vaticination technology. Several models are used by machine literacy to grease and authenticate vaticination. The study focuses on LSTM- grounded machine literacy and retrogression for stock value vaticination. Open, near, low, high, and volume are all factors. One of the most significant conditioning in the world of finance is stock trading. Stock request vaticination is an act of trying to determine the unborn value of a stock or other fiscal instrument traded on a fiscal

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# I. INTRODUCTION

Stock prediction using machine learning is a powerful technique that can help traders and investors to make informed decisions by using historical and real-time data to predict future stock prices. Machine learning algorithms are trained on historical stock prices and other relevant data such as financial news, economic indicators, and company financial reports to identify patterns and trends that can be used to make predictions about future stock prices.

The real-time aspect of this technique involves using current market data to make predictions in real-time. By continuously updating the model with the latest data, traders and investors can get a more accurate picture of the market and make timely decisions. An end-to-end web application for stock price prediction using machine learning in real-time can provide a user-friendly interface for traders and investors to access the latest predictions and insights. This type of application can be a valuable tool for anyone who wants to stay on top of the stock market and make informed investment decisions.

In the recent times, adding elevation of machine literacy in colorful diligence have enlightened numerous dealers to apply machine literacy ways to the field, and some of them have produced relatively promising results. This will develop a fiscal data predictor program in which there will be a dataset storing all literal stock prices and data will be treated as training sets for the program. The main purpose of the vaticination is to reduce query associated to investment decision timber. The probable stock request vaticination target can be the unborn stock price or the volatility of the prices or request trend. In the vaticination there are two types like dummy and a real time vaticination which is used in stock request vaticination system. In Dummy vaticination they have define some set of rules and prognosticate the unborn price of shares by calculating the average price. In the real time vaticination mandatory used internet and saw current price of shares of the company. Computational advances have led to preface of machine literacy ways for the prophetic systems in fiscal requests.

Long Short-Term Memory (LSTM) is a type of recurrent neural network (RNN) architecture that is used for sequence modeling, particularly in tasks that require memory of past events. LSTMs were introduced by Hochreiter and Schmidhuber in 1997, and have since become a popular tool in natural language processing, speech recognition, and other applications.

The key innovation of LSTM is the addition of "memory cells" to the standard RNN architecture. These cells allow the network to selectively remember or forget information from past inputs, based on the current input and the network's internal state. [1]

Stock prices are unpredictable due to different factors that are involved in the stock request, similar as geopolitical pressure, company earnings, and commodity prices, affecting stock price. occasionally stock prices reply to domestic query similar as reserve bank policy, government policy, affectation, and global request query. The volatility estimation of stock is one of the gruelling tasks for dealers. Accurate vaticination of stock price helps investors to reduce the threat in portfolio or investment. Stock prices are nonlinear. [2]



A correct vaticination of stocks can lead to huge gains for the dealer and the broker. constantly, it's brought out that vaticination is chaotic rather than arbitrary, which means it can be prognosticated by precisely assaying the history of separate stock request. It predicts a request value close to the palpable value, thereby adding the delicacy. preface of machine literacy to the area of stock vaticination has appealed to numerous inquiries because of its effective and accurate measures

#### **II. LITERATURE REVIEW**

Raghav Nandakumar [2] the original focus of our literature check was to explore general online literacy algorithms and see if they could be acclimated to our use case i.e., working on real-time stock price data. These included Online AUC Maximization, Online Transfer literacy, and Online point Selection. still, as we were unfit to find any implicit adaption of these for stock price vaticination, we also decided to look at the being systems, dissect the major downsides of the same, and see if we could ameliorate upon them. We zeroed in on the correlation between stock data ( in the form of dynamic, longterm temporal dependencies between stock prices) as the crucial issue that we wished to break. A brief hunt of general results to the below problem led us to RNNs and LSTM

Abhishek Karpe [3]The sentiments from colorful investors, investment enterprises, and dealers are important to calculate a plan where the request can go, but this data can't be trusted fully as there is a trap employed by colorful drivers who are financially strong to control the request and make a state of confusion among the common retail investors. Hence specialized analysis of the stocks is also important to prognosticate the long-run performance of a stock or an association.

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( PRE) fashion and deep neural network( DNN). First, stock specialized pointers are considered to identify the uptrend in stock prices [4].

Stock prediction is a challenging problem, and there has been a considerable amount of research in this area over the years. In this literature review, we will discuss some of the key works related to stock prediction.

One of the earliest approaches to stock prediction was based on fundamental analysis. This approach involves analyzing a company's financial and economic data to make predictions about its future performance. However, this approach has some limitations, as it relies heavily on the accuracy and availability of the data.

Technical analysis is another popular approach to stock prediction, which involves studying past market data, such as stock prices and trading volumes, to identify patterns and trends that can help predict future price movements. Various techniques have been proposed to identify such patterns, including moving averages, momentum indicators, and chart patterns.

Machine learning techniques have also been applied to stock prediction. These approaches involve training models on historical data to identify patterns and make predictions. Some of the most popular machine learning algorithms used for stock prediction include decision trees, neural networks, and support vector machines.

Deep learning techniques, which involve training deep neural networks on large amounts of data, have also been applied to stock prediction. These approaches have shown promising results, as they can capture complex patterns and relationships in the data.

Another area of research related to stock prediction is sentiment analysis, which involves analyzing news articles, social media posts, and other sources of information to gauge public sentiment about a company or the stock market as a whole. This approach has shown to be useful in predicting short-term price movements.

## **III. METHODOLOGY**

Stock Vaticination is a complex process that involves assaying once and present data to make prognostications about unborn stock prices. There are colorful methodologies for stock vaticination, including specialized analysis, abecedarian analysis, and machine literacy. Specialized analysis involves studying literal request data similar as price and volume to identify patterns and trends that can help prognosticate unborn stock prices. Specialized judges use maps and other tools to identify support and resistance situations, moving pars, and other pointers that can give sapience into request trends.

Abecedarian analysis, on the other hand, involves assaying a company's fiscal data, including earnings reports, balance wastes, and income statements, to determine the beginning value of the company and its stock. Judges may also consider macroeconomic factors similar as interest rates, affectation, and profitable growth when making stock prognostications. Machine literacy ways have come decreasingly popular for stock vaticination in recent times.

Machine literacy algorithms can dissect large quantities of data and identify patterns that may not be apparent to mortal judges. These algorithms can also learn from once data and acclimate their prognostications over time as new information becomes available. Some popular machine literacy ways for stock vaticination include retrogression analysis, decision trees, neural networks, and support vector machines.

These ways can be applied to colorful types of data, including fiscal data, news papers, and social media sentiment analysis. It's worth noting that stock vaticination is innately changeable, and no methodology can guarantee accurate prognostications all the time. nevertheless, combining different methodologies and approaches can increase the delicacy of prognostications and help investors make informed opinions.

In *Figure 1* The cascade methodology is a design operation approach that involves a direct and successional process,



where each phase of the design is completed before moving on to the coming. While it may not be a typical approach for stock vaticination, then's an illustration of how the cascade methodology could be applied to a stock vaticination design Requirement Gathering The first phase would involve gathering the conditions for the stock vaticination model. This would involve determining the data sources that would be used, the type of data to be collected, and the overall ideal of the design.

Analysis In this phase, the data would be anatomized to determine the trends, patterns, and other applicable information. This would involve using statistical ways, data mining, and machine literacy algorithms to identify patterns and connections in the data.

Design The design phase would involve creating the model armature and opting the applicable algorithms and ways that would be used for the stock vaticination. This would involve creating a detailed plan that outlines the way involved in the vaticination process.

Implementation in this phase, the actual Implementation of the model would take place. This would involve rendering and programming the model, testing it, and enriching it to insure it works as intended. Testing Once the model has been enforced, it would be tested to insure it's accurate and dependable. This would involve comparing the prognostications generated by the model with factual stock prices to determine its position of delicacy.

Deployment In the final phase, the stock vaticination model would be stationed, and ongoing monitoring would be conducted to insure it continues to serve as intended.



Figure 1 Waterfall methodology.

## V. RESULTS

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	TELI ive share price evolution	
Predict		
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#### Figure 2 Prediction page.

In Figure 2 The user will enter the input to the ticker name of whichever company stocks they required. Then, will enter the no of days of data they required for the prediction accordingly. The left-hand side shows, the livestock predictions. The right-hand shows the future stocks which will be predicted in the next seven days.



#### Figure 3 Sentiment analysis page

In Figure 3 Sentiment analysis is the process of analyzing text data to determine the sentiment or emotion expressed in it. VADER (Valence Aware Dictionary and sEntiment Reasoner) is a popular open-source tool for sentiment analysis. It is a rule-based model that is trained on a lexicon of words and their associated sentiment scores.

The VADER tool is designed to handle social media text data that is typically short, informal, and often contains sarcasm, irony, or other forms of ambiguity. It can identify the sentiment expressed in text data as positive, negative, or neutral, and it can also provide a sentiment score that ranges from -1 (extremely negative) to +1 (extremely positive).

The VADER tool uses a combination of rules and heuristics to identify sentiment. It takes into account the intensity of words, the presence of emoticons or emojis, and the context in which the words are used. For example, the word "love" would be considered highly positive, while the word "hate" would be considered highly negative.

To use VADER for sentiment analysis, one needs to tokenize the text data into individual words and sentences, and then apply the VADER tool to each sentence to obtain a sentiment score. The sentiment scores can then be aggregated to obtain an overall sentiment score for the entire text data.



VADER has been widely used for sentiment analysis in social media data, product reviews, and customer feedback. It is a useful tool for quickly analyzing large volumes of text data and can provide valuable insights into the sentiment of customers, users, or the general public. However, it should be noted that VADER, like any other sentiment analysis tool, has its limitations and may not always accurately capture the nuances and complexities of human emotions.

## **VI. CONCLUSION**

In conclusion we can say that the fashion ability of stock request trading is growing fleetly which is encouraging experimenters to find out new styles for the vaticination using new ways. The soothsaying ways isn't only helpful to experimenters, but it also helps investors or any person dealing with the stock request. To help prognosticate stock indicators or soothsaying model with good delicacy is needed. Overall, we conclude that Stock vaticination using Machine literacy can increase the vaticination delicacy. A model can learn the trends with huge quantum of data handed and train themselves to prognosticate the movement of the stock price. We can conclude that literal data can be used as important tool that allows us to read the movement more directly

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