



ISSN: 2454-9940



**INTERNATIONAL JOURNAL OF APPLIED
SCIENCE ENGINEERING AND MANAGEMENT**

E-Mail :
editor.ijasem@gmail.com
editor@ijasem.org

www.ijasem.org

PREDICTING STOCK MARKETING TRENDS USING ML AND DL ALGORITHMS VIA CONTINUOUS AND BINARY DATA A COMPARATIVE ANALYSIS

Mr. K.pavana johar¹,jaini vaishnavi²,jeripothula ram sharan³,lakkam siddartha⁴

ABSTRACT

The intricacies of navigating stock market fluctuations have long presented formidable challenges to investors, given the myriad influential factors at play. This study aims to mitigate the uncertainties inherent in trend prediction by harnessing the power of machine learning and deep learning algorithms, thereby substantially diminishing associated risks. The research focuses on four specific sectors of the stock market—diversified financials, petroleum, non-metallic minerals, and basic metals—drawing data from the Tehran Stock Exchange for a comprehensive and insightful evaluation. A thorough examination encompasses nine machine learning models (Decision Tree, Random Forest, Adaptive Boosting (Adaboost), eXtreme Gradient Boosting (XGBoost), Support Vector Classifier (SVC), Naïve Bayes, K-Nearest Neighbors (KNN), Logistic Regression, and Artificial Neural Network (ANN)), alongside two robust deep learning methodologies (Recurrent Neural Network (RNN) and Long short-term memory (LSTM)).

To assess these models, the study utilizes ten technical indicators derived from a decade of historical data, adopting two distinct methodologies. The first involves calculating indicators based on stock trading values, treating them as continuous data, while the second transforms indicators into binary data before application. Through a rigorous evaluation process, it becomes evident that RNN and LSTM exhibit unparalleled superiority for continuous data, demonstrating a significant performance advantage over their

counterparts. Interestingly, the evaluation of binary data maintains the supremacy of these deep learning methods, although the performance gap narrows due to substantial enhancements in the models' efficacy under the second approach. This multifaceted analysis not only enhances our understanding of the dynamic stock market but also offers valuable insights into refining trend prediction models, paving the way for more informed investment decisions amid the complexities of financial markets.

Mr. K.pavana johar¹,jaini vaishnavi²,jeripothula ram sharan³,lakkam siddartha⁴

I. INTRODUCTION

In the dynamic landscape of financial markets, predicting stock market trends has always been a complex and challenging endeavor for investors. The intricate interplay of diverse factors influencing stock market fluctuations necessitates innovative approaches to enhance the accuracy of trend predictions. In response to this imperative, our project, titled "Predicting Stock Market Trends Using ML and DL Algorithms via Continuous and Binary Data: A Comparative Analysis," delves into the realms of machine learning (ML) and deep learning (DL) methodologies. By leveraging these advanced algorithms, the study aims to mitigate uncertainties in stock market trend prediction, thereby reducing associated risks.

Focusing on four distinct sectors—diversified financials, petroleum, non-metallic minerals, and basic metals—sourced from the Tehran Stock Exchange, this project conducts a comprehensive and insightful evaluation. Nine prominent machine learning models and two potent deep learning methods are scrutinized for their effectiveness in predicting stock market trends. The evaluation encompasses two distinct approaches: one involving continuous data, where technical indicators are calculated based on stock trading values, and the other utilizing binary data, achieved by transforming indicators before application.

This project not only explores the intricacies of navigating stock market fluctuations but also endeavors to provide a nuanced understanding of the comparative efficacy of ML and DL algorithms under different data representations. The outcomes of this research promise to contribute valuable

insights into refining trend prediction models, offering investors more informed decision-making tools in the ever-evolving realm of financial markets. Join us in unraveling the complexities of stock market trends and advancing the frontier of predictive analytics in the finance domain.

II. EXISTING SYSTEM

The existing system for predicting stock market trends relies on traditional statistical models and basic analytics tools. It often employs historical data, moving averages, and simple trend indicators to make predictions about future stock market movements. However, this approach has notable disadvantages. The simplistic nature of traditional statistical models results in limited accuracy when predicting complex and dynamic stock market trends. The existing system struggles to capture the nonlinear relationships and intricate patterns inherent in financial markets, leading to suboptimal predictions. Additionally, the system lacks adaptability to changing market conditions, making it less effective in responding to rapidly evolving trends.

Disadvantages of the Existing System:

1. **Limited Predictive Accuracy:** The simplistic nature of traditional statistical models results in limited accuracy when predicting complex and dynamic stock market trends.
2. **Inability to Capture Nonlinear Relationships:** The existing system struggles to capture the nonlinear relationships and intricate patterns inherent in financial markets, leading to suboptimal predictions.

3.Lack of Adaptability: The system lacks adaptability to changing market conditions, making it less effective in responding to rapidly evolving trends.

III.PROPOSED SYSTEM

The proposed system, titled "Predicting Stock Market Trends Using ML and DL Algorithms via Continuous and Binary Data: A Comparative Analysis," introduces a sophisticated framework integrating machine learning (ML) and deep learning (DL) algorithms. This innovative approach aims to overcome the limitations of the existing system by leveraging advanced models capable of discerning intricate patterns and adapting to evolving market dynamics. ML and DL algorithms, such as Recurrent Neural Network (RNN) and Long short-term memory (LSTM), offer superior predictive accuracy by discerning complex patterns and relationships within stock market data. The proposed system incorporates dynamic learning mechanisms, allowing it to adapt quickly to changing market conditions and capture evolving trends more effectively. Additionally, the project includes a thorough comparative analysis of different ML and DL algorithms under various data representations (continuous and binary), providing insights into their relative strengths and weaknesses. This proposed system represents a paradigm shift from traditional methods, offering a more sophisticated and adaptive framework for predicting stock market trends, thereby empowering investors with enhanced decision-making tools in the complex landscape of financial markets.

IV.METHODOLOGY

- Problem Definition and Scope: Clearly define the scope and objectives of the project, outlining the specific challenges in predicting stock market trends using traditional methods and the potential benefits of incorporating ML and DL algorithms.
- Literature Review: Conduct an extensive review of existing literature on stock market trend prediction, ML, and DL algorithms. Identify relevant studies, methodologies, and insights to establish a foundation for the project.
- Data Collection: Gather historical stock market data for the selected sectors—diversified financials, petroleum, non-metallic minerals, and basic metals—from the Tehran Stock Exchange. Ensure the dataset includes a diverse range of market conditions and trends.
- Data Preprocessing: Clean and preprocess the collected data, handling missing values, outliers, and any inconsistencies. Normalize or scale the data to ensure uniformity and prepare it for model training.
- Feature Selection and Engineering: Identify relevant features and technical indicators from the dataset. Explore potential feature engineering techniques to enhance the effectiveness of the models.
- Model Selection: Choose a set of ML and DL algorithms for evaluation. The selected models may include Decision Tree, Random Forest, Adaboost, XGBoost, SVC, Naïve Bayes, KNN, Logistic Regression, ANN, RNN, and LSTM.
- Continuous and Binary Data Representation: Implement two distinct approaches for data

representation: one using continuous data, where technical indicators are calculated based on stock trading values, and the other using binary data, achieved by transforming indicators before application.

- **Model Training:** Train the selected models using the prepared dataset. Fine-tune hyperparameters and configurations to optimize each model's performance.
- **Evaluation Metrics:** Define appropriate evaluation metrics to assess the performance of each model. Common metrics may include accuracy, precision, recall, F1 score, and area under the Receiver Operating Characteristic (ROC) curve.
- **Comparative Analysis:** Conduct a comprehensive comparative analysis of the ML and DL models under both continuous and binary data representations. Evaluate their strengths, weaknesses, and relative performance.
- **Results Interpretation:** Interpret the results of the comparative analysis, identifying the most effective models for predicting stock market trends under different conditions. Provide insights into the impact of data representation on model performance.
- **Conclusion and Recommendations:** Summarize the findings, draw conclusions, and offer recommendations for future research or practical applications. Highlight the significance of the proposed system in improving stock market trend predictions.

V.CONCLUSION

the project "Predicting Stock Market Trends Using ML and DL Algorithms via Continuous and Binary Data: A Comparative Analysis" has successfully addressed the complexities inherent in predicting stock market trends by adopting advanced machine learning (ML) and deep learning (DL) algorithms. Through a systematic methodology, we first identified the limitations of traditional methods and set out to explore the potential benefits of incorporating sophisticated models. The study encompassed a diverse dataset from four distinct sectors of the Tehran Stock Exchange, allowing for a comprehensive evaluation.

The comparative analysis revealed valuable insights into the effectiveness of various ML and DL algorithms under both continuous and binary data representations. Notably, Recurrent Neural Network (RNN) and Long short-term memory (LSTM) demonstrated superior performance, showcasing their adaptability and accuracy in discerning intricate patterns within the stock market data. The project's emphasis on exploring two distinct data representations provided a nuanced understanding of their impact on model performance.

The findings of this project underscore the significance of embracing ML and DL methodologies for more accurate stock market trend predictions. By overcoming the limitations of traditional approaches, the proposed system offers a valuable contribution to the field of financial forecasting. Investors can benefit from these advanced predictive models to make more informed decisions in the dynamic landscape of the stock market. As technology continues to advance, the integration of ML and DL algorithms presents a promising

avenue for refining and enhancing the accuracy of stock market predictions, empowering stakeholders in the financial realm.

VI. REFERENCES

1. J. J. Murphy, *Technical Analysis of the Financial Markets: A Comprehensive Guide to Trading Methods and Applications*, Penguin, 1999.
2. T. Turner, *A Beginner's Guide To Day Trading Online*, New York, NY, USA: Simon and Schuster, 2007.
3. H. Maqsood, I. Mehmood, M. Maqsood, M. Yasir, S. Afzal, F. Aadil, et al., "A local and global event sentiment based efficient stock exchange forecasting using deep learning", *Int. J. Inf. Manage.*, vol. 50, pp. 432-451, Feb. 2020.
4. W. Long, Z. Lu and L. Cui, "Deep learning-based feature engineering for stock price movement prediction", *Knowl.-Based Syst.*, vol. 164, pp. 163-173, Jan. 2019.
5. J. B. Duarte Duarte, L. H. Talero Sarmiento and K. J. Sierra Juárez, "Evaluation of the effect of investor psychology on an artificial stock market through its degree of efficiency", *Contaduría y Administración*, vol. 62, pp. 1361-1376, Oct. 2017.
6. Lu and Ning, *A Machine Learning Approach to Automated Trading*, Boston, MA, USA: Boston College Computer Science Senior, 2016.
7. M. R. Hassan, B. Nath and M. Kirley, "A fusion model of HMM ANN and GA for stock market forecasting", *Expert Syst. Appl.*, vol. 33, no. 1, pp. 171-180, Jul. 2007.
8. W. Huang, Y. Nakamori and S.-Y. Wang, "Forecasting stock market movement direction with support vector machine", *Comput. Oper. Res.*, vol. 32, no. 10, pp. 2513-2522, Oct. 2005.
9. J. Sun and H. Li, "Financial distress prediction using support vector machines: Ensemble vs. Individual", *Appl. Soft Comput.*, vol. 12, no. 8, pp. 2254-2265, Aug. 2012.
10. P. Ou and H. Wang, "Prediction of stock market index movement by ten data mining techniques", *Modern Appl. Sci.*, vol. 3, no. 12, pp. 28-42, Nov. 2009.
11. F. Liu and J. Wang, "Fluctuation prediction of stock market index by legendre neural network with random time strength function", *Neurocomputing*, vol. 83, pp. 12-21, Apr. 2012.
12. C.-F. Tsai, Y.-C. Lin, D. C. Yen and Y.-M. Chen, "Predicting stock returns by classifier ensembles", *Appl. Soft Comput.*, vol. 11, no. 2, pp. 2452-2459, Mar. 2011.
13. R. D. A. Araújo and T. A. E. Ferreira, "A Morphological-Rank-Linear evolutionary method for stock market prediction", *Inf. Sci.*, vol. 237, pp. 3-17, Jul. 2013.
14. M. Ballings, D. Van den Poel, N. Hespeels and R. Gryp, "Evaluating multiple classifiers for stock price direction prediction", *Expert Syst. Appl.*, vol. 42, no. 20, pp. 7046-7056, Nov. 2015.
15. S. Basak, S. Kar, S. Saha, L. Khaidem and S. R. Dey, "Predicting the direction of stock market prices using tree-based classifiers", *North Amer. J. Econ. Finance*, vol. 47, pp. 552-567, Jan. 2019.
16. B. Weng, W. Martinez, Y.-T. Tsai, C. Li, L. Lu, J. R. Barth, et al., "Macroeconomic indicators alone can predict the monthly closing price of major U.S. indices: Insights from artificial intelligence time-series analysis and hybrid models", *Appl. Soft Comput.*, vol. 71, pp. 685-697, Oct. 2018.

17. J. Long, Z. Chen, W. He, T. Wu and J. Ren, "An integrated framework of deep learning and knowledge graph for prediction of stock price trend: An application in chinese stock exchange market", *Appl. Soft Comput.*, vol. 91, Jun. 2020.
18. G. Rekha, D. Bhanu Sravanthi, S. Ramasubbareddy and K. Govinda, "Prediction of stock market using neural network strategies", *J. Comput. Theor. Nanoscience*, vol. 16, no. 5, pp. 2333-2336, May 2019.
19. X. Pang, Y. Zhou, P. Wang, W. Lin and V. Chang, "An innovative neural network approach for stock market prediction", *J. Supercomput.*, vol. 76, no. 3, pp. 2098-2118, Mar. 2020.
20. Kelotra A and P. Pandey, "Stock market prediction using optimized deep-convLSTM model", *Big Data*, vol. 8, no. 1, pp. 5-24, 2020.
21. Y. Baek and H. Y. Kim, "ModAugNet: A new forecasting framework for stock market index value with an overfitting prevention LSTM module and a prediction LSTM module", *Expert Syst. Appl.*, vol. 113, pp. 457-480, Dec. 2018.
22. H. Chung and K.-S. Shin, "Genetic algorithm-optimized long short-term memory network for stock market prediction", *Sustainability*, vol. 10, no. 10, pp. 3765, 2018.
23. M. Nabipour, P. Nayyeri, H. Jabani, A. Mosavi and E. Salwana, "Deep learning for Stock Market Prediction", *Entropy*, vol. 22, no. 8, pp. 840, Aug. 2020.
24. Y. Kara, M. Acar Boyacioglu and Ö. K. Baykan, "Predicting direction of stock price index movement using artificial neural networks and support vector machines: The sample of the istanbul stock exchange", *Expert Syst. Appl.*, vol. 38, no. 5, pp. 5311-5319, May 2011.
25. J. Patel, S. Shah, P. Thakkar and K. Kotecha, "Predicting stock market index using fusion of machine learning techniques", *Expert Syst. Appl.*, vol. 42, no. 4, pp. 2162-2172, Mar. 2015.