

USING MACHINE LEARNING FOR SHORT-TERM EXTRAPOLATION IN FINANCIAL FORECAST BASED ON FIELD-PROGRAMMABLE GATE ARRAY

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ABSTRACT

A weather report is a prediction about the lending market that is made. There could be insufficient factors or stock market assessment in the finance industry outlook. Financial statistics can be precise or imprecise, average, model-based, or information for each occasion. Hundreds of impact analysis and test systems are included in the gate series, which are used to generate financial forecasts. We initially used a digital signal processing sensor and an IoT (Internet of Things) network to measure exchange rates, for example. Especially over the past, but just not the least, this approach makes it difficult to cut spending and establish untrue values to combat inflation. The proposed approach for infrastructure and finance includes FPGAs and deep learning. Predictive analysis issues and possibilities arising from financial statement unique features, signal-to-noise ratios, repetitive predictor variables, numerical volatility, and ecological predictability resulting from competitors' demand and shareholder learning are all significant issues. The mean, variability, and probability allocation of asset prices are estimated using machine learning models. Financial forecast analysis is a part of FPGA (Field-Programmable Gate Array) research, which raises data mining concerns because many forecast models are used.

Keywords: Exchange Rate Financial Prediction Platform, FPGA (Field-Programmable Gate Array), And Machine Learning.

I. INTRODUCTION

Financial systems are usually straightforward to conquer. Cash in liquid currency used in financial sales should be focused on both currency value calculations. Deep learning assesses the global financial climate's risks and benefits and estimates the foreign exchange market. The value of future benefits or prediction attribute denotes the undervalued value. Expected estimates select the data collection for the building of the data. The estimate of currency rates based on the predicted results is by two pure methods. In order to determine the value of a stock, professionals use structured equations. In order to estimate or produce buys/sells, the equation inflation rate can be used. Shopping and payments can be made through actions. There may be a big variation between the expected alternation and the predicted reversal rate; obviously, it depends on the formula and the inflation rate on the market at any time. If the price is significantly different or higher, the instructor can recoup the quantity. The difference between the halted currency value and the current rate of interest is called the risk price. If the specialist decides that there is a difference as a result of a false decision, it will generate a buy or sell signal.

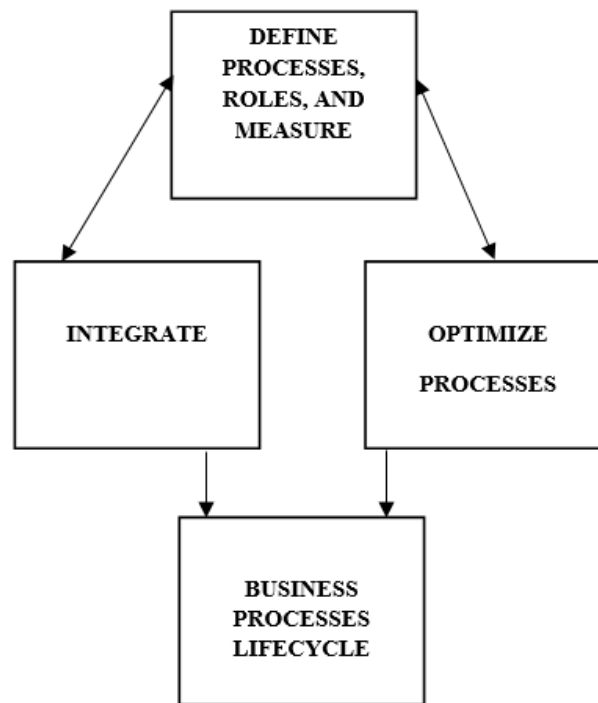


Figure 1 Block diagram of Exchange Rate Financial Platform

Figure 1 shows how the predicted time frame is used for measuring the parameters of the model. Suppose that the networks will develop further and that yet another projection in this time frame was not a "real opportunity." Only attributes are used for these step-by-step hints. The associated FPGA (Field-Programmable Gate Array) evaluation error and validity data are not included in the model and parameter estimation. The "real opportunity" was a one-step reference during this period, also known as blackouts. It generates estimates and nearly always error numbers in the future. The evaluation reports the findings of this test process in order to determine if the sample population can be used by chance to produce models. In our daily lives, money is an important and crucial factor. At the same period, as they started to spend in other regions and stocks, mostly reliant on exchange rates, Cash, the nation and its citizens began investing in a number of companies.

The issue of FPGA is bonds (i.e., financial sector ups and downs), policy and investment dynamics FPGA is (Field-Programmable Gate Array). The rate of exchange, which is commonly recognized as the currency, is one of the most critical decisions on the capital market. Due to its dynamic and uncertain nature, the global trading market is hard to predict. As a result, it is important for the finance sector to estimate foreign exchange rates. The trade and investment value can be managed by assessing the trade and investment impact of exchange rates, which leads to a more responsive international trade and, in particular, to significant advantages for investors.

In today's dynamic trading world, maintaining the right currency values has become a big problem. Currencies affect financial decisions and the behavior of participants enormously. Due to abrupt dynamic data changes, it is difficult to incorporate large amounts of data and evaluate real and relevant rates of translation. The Smart Exchange Rate Prediction Algorithm and the machine learning process obtain real-time exchange rate data and calculate the waste of processing time of potential exchange rates for regression model estimation of complex currency exchange rates. This approach is more effective and accurate when processing large amounts of historical and complicated data. The enormous gains it provides are an application in the field of economics that generally seeks to estimate the currencies.

On the other hand, the forecast is unsure and could lead to significant losses. Intelligent computing of model time series data is part of machine learning methods. Recent competitions also have used hybrid models, state-of-the-art models, neural networks, vector support machines, and blurry rules of natural language. The most difficult situation is when traditional mathematical models fail to estimate the time series and are dependent on a large number of the entry parameters when attempts are made to estimate time series in economics.

Financial forecasts can be challenging, and some challenges and solutions are discussed in this article. Inefficient testing for minutes is required in order to design the reference: data is evaluation algorithm, preprocessing

matches, compilation and trading templates, and estimate and tuning. The inquiry is necessary because there is no known solution, but experience from others will nevertheless help solve challenges and speed things up. A peculiar concept that has been tried on several occasions is the financial (as well as cumulative) assessment. Excessive performance in a highly competitive world requires more insight and complexity. The efficient method reported is a conventional solution, while the neural network is a personalized hybrid plug-in system, which often does not provide simple data.

II. RELATED WORK

The basis of the verdict are based the future decisions and regulations. It gives the decision-makers options to achieve goals and directions and allows for route discovery and differentiation [1]. The interaction between the dual structure, the system of economic innovation and the climate is limited by the degree that makes it difficult to accurately predict future trends of growth[2]. Since the sample is small, however, no matter whether the economy is vulnerable to major changes, irregularities or qualitative evidence, the analysis focuses on conventional model predictions such as time series predictions, the more significant, the fewer data are given, as well as a multiple regression prediction grey forecast model[3].

Dynamic methods are now being proposed to predict the accuracy of established models to forecast financial problems. A new financial crisis for the sector will continue as the business climate continues[4]. In response to the financial crisis, there are two forms of recent economic data development. The distribution of functions is the first update, which is a virtual concept that derives financial details. As the financial crisis shifts and health priorities shifts[5] the provision of the data would become more current financial data. The actual definition of financial detail and the second post-distribution transition are inconsistent. The financial crisis has changed the idea of a welfare objective. Note that if there is some term drift [6] the new classification is outdated and inaccurate. In order to change the model, it is necessary to reverse the financial crisis. The stock estimate is a crucial problem in academics and economics. This dilemma is often tackled with technology[7]. A model attendance model is developed that analyses, uncover, and forecasts the relationship between news content and market markets. Both posts would have a positive or negative impact on the market[8].

The main variables are assessed, and it is possible to determine if the net profit production is within a certain percentage of the target, according to eight typical characteristics. This makes the determination of the net profit growth rate over the last few percentage points critical. (nine) (nine) Review the most crucial problems with processed versions in data acquisition. As a result, the data is reliable, but it is difficult to achieve the decision tree algorithm[10]. The data characteristics can be divided into various domain treatment classes. In other words, the decision tree uses split parameters to determine the split point for the attribute. The data are arranged by the split point. Processing is not necessary as the target data is not autonomous[11].

An early-warning data analysis model was created for different sectors in the financial crisis. Each industry has a set of business features that require an increase in financial crises in the various sectors, the decrease of financial information between industries, and the contrast of emergency alert patterns in financial data gaps. [13]. [14]. Second, such contextual factors have to be taken into account, and an early warning pattern could emerge for a financial crisis. Indicators are used non-financially [13]. The validity of the financial documents could be detected by an emergency alert model for a possible economic downturn. Finally, its errors should be used to learn from the next financial crisis warning paradigm. (#14) - (#14) [15.] The following: In the context of corporate financial battles, economic and social problems have stimulated the quest for better intellectual understanding and capacity. [No. 16] [No. 16] [17.] [17.] The research focuses on using a conventional pattern recognition program to establish reliable models for solving current financial problems with difficult financial forecasts. Combine two of the most common approaches, parallel and series. [90] [90] [90] [99] [20][21] [20][21] [24][22][23][23][24] [25] The report [25] [23], [24], [25], [27], [28], and the following: A single classification output correlation is the best energy flowing for providing the best system decision. Due to the complexity of the grouping, the value of the entry margin is difficult to classify. This method creates a double-level grading system that increases the accuracy of calculation and eliminates margin value issues. [20].

III. PLATFORM BASED ON MACHINE LEARNING MATERIALS AND METHOD FOR EXCHANGE TRATE DELEVERATION

Financial predictions are a strategy method that links the future management of an enterprise with its anticipated economic, technological, competitive, and social environment. Strategies and attitudes are commonly used in policy growth to achieve short, medium, and long-term goals. Machine training is used to monitor financial

supervision involving the format reports and different operating budgets of the reference financial statements. One of three ways can be an economical estimate. The official financial statements, the cash budget, and the operational budgets are the three forms of financial statements. Official financial statements are company asset basis forecasts, projected performance, and financing needs.

A cash budget is a systematic list of the actual cash flows and outflows of an organization. They are accounting for operations. The departmental income and expenditure formulas are mentioned in various ways. Machine learning is an official statement by the executive and the cash flow estimates, which helps you understand the financial capacity of the company. These statements could be accompanied by revenue statements and balance sheets. The appropriate operating manual applies to the operating schedule of the business (P&L). Furthermore, the official balance sheet reflects the financial position of the company at the point of collection and the overall impact of future action. Machine training is a brief quote that shows insights into the company, the accounts, the outsourced results and scenarios for future processes and the expected capital inflows of the report in the third report. During the forecast period, the different funds are expected to change.

The financial outlook as quotation, growth, or prospects of a company will be shown in Figure 2. The company's cash flow statements, as seen above, are available for reference financing. Refer to whether it is necessary to understand the variations in the model for future predictions and financial performance. Recall calculating the major aspect of the gain computation, in this case. Both the health of companies and the willingness of users to meet their needs are affected by the condition of the world economy. Calculations of benefits and expenses shall be carried out to correctly initiate this process and analyze future revenue and expends of the company.

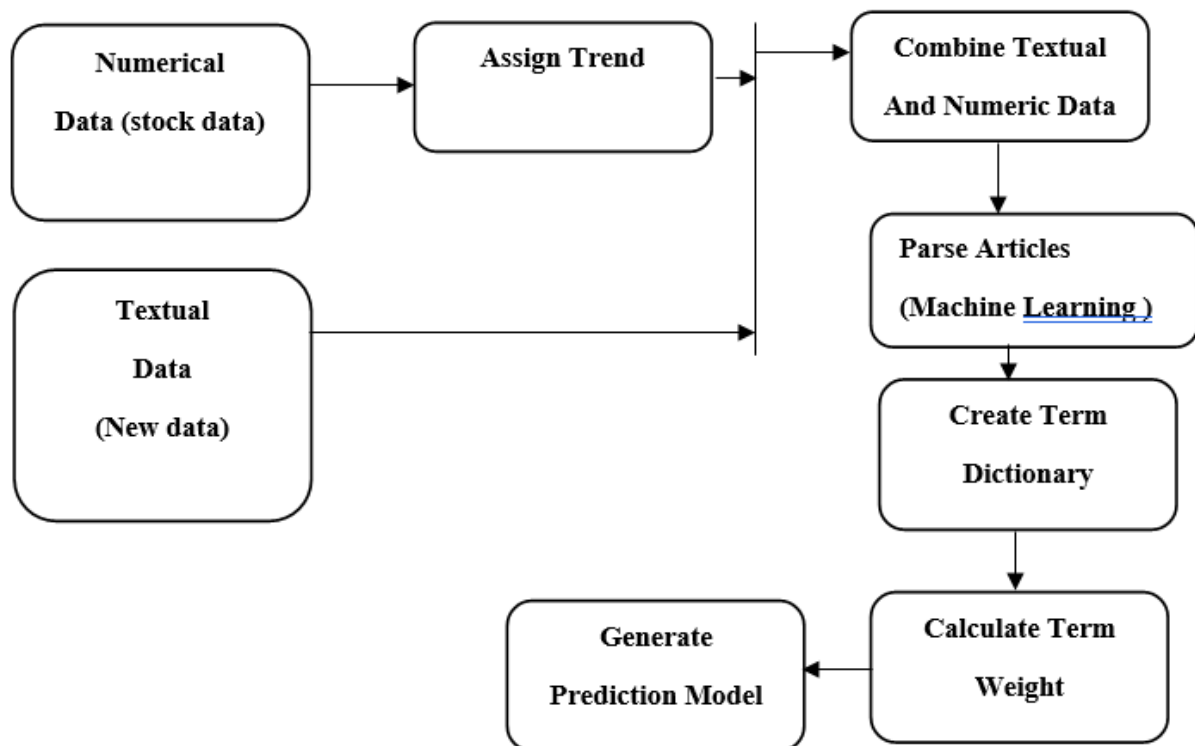


Figure 2 Financial statement-based machine learning block diagram. Figure 2

Historical research explores a machine learning company's previous financial performance with forecasts that this data will succeed—the financial statement, balance sheet, and cash flow statement. The latter is the case because the company has progressed in recent years, and we can evaluate on the basis of these reports what projections for the coming year are being made. The advantage of this process is that it is simple and requires no detailed technical knowledge. One of the flaws, however, is that historical references in the general market and comparative accounts are not included both temporarily and methodically. As a result, checks for investor book returns are wrong. It is wrong.

3.1 PLATFORM ON FINANCIAL PREDICTION (FIELD PROGRAMMABLE FIAL THE GATE ARRAYS)

The analytical projections go beyond the financial results of previous years. Research-based forecasts are more precise than the typical market product. Find out often what the adversaries expect and compare it with your own. As new consumer trends develop, a finance advisor supports innovation in the fuel industry. Technological changes and their impact on future growth, as well as other developments that could affect the entire industry. For example, if you are in the Telecommunications sector, the market position, new trends, and the expected revenues of your competitors will be part of the financial forecast focused upon the studied market size as it is more straightforward for users and shareholders to objectively attract and forecast growth based on survey predictions.

Companies review and use this information to predict future outcomes using a financial projection of the Field-Programmable Gate Array. The prediction's goal is, therefore, to make future performance more accessible. Based on these, the company is making predictions. Financial simulations are used for FPGA (Field-Programmable Gate Array), and company predictions should be based upon different financial reports. With regard to business and financial planning, the organization uses expenditures to estimate income and income expectations for the future. When the correct feedback from the person responsible for the accounting department is correctly provided by the programmable gate array, your business revenue outlook and the cash outflow may be used in the returns statement, balances, and other business metrics.

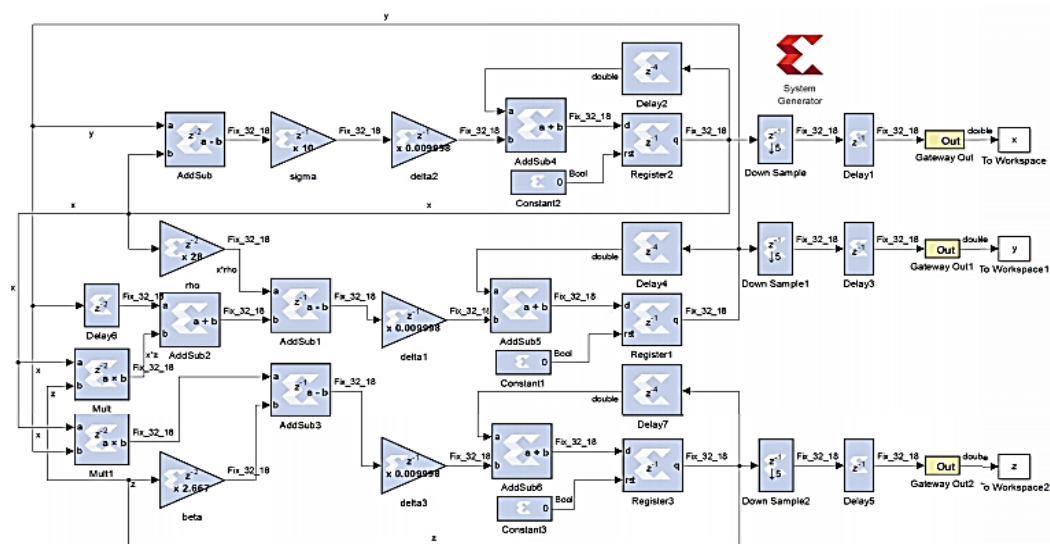


Figure 3 FPGA platform-based output.

As shown in Figure 3, for several factors, it is important to predict programmable gate-based financial statements. It provides investors with plans and information. It is best to use the percentage during the tax phase to estimate the financial statement. This procedure often saves time with fewer estimation data. This method helps to calculate the latest sales figure. The top-down approach addresses both the macroeconomic factors and the overall business situation. These statistics are translated into a company sector or product sales forecast and development projections. In order to establish demand and price statistics for every brand, we use the following technique to calculate sales forecasts around the company.

3.2 DISCUSSION RESULT.

As a revenue statement with few substantial differences, the financial statement can be estimated. We would use the same format as we have seen so far for products that fluctuate directly in price, and we will explain how we deal further with those goods. Using historical evidence and mathematical data, future interactions are predicted. Predictions aren't necessarily reliable when such relationships change. Because of ambiguity, some assessments in different contexts are inaccurate; progress should therefore be taken into account—the likelihood of each condition and an estimate. If the schedule is lengthened, the forecast is less precise. You should propose halving the forecast period if you want to improve the accuracy of the forecast. The extent of the planning process is determined by how often you review new plans. Stability of income, corporate setbacks, and financial conditions are all important. It is more accurate to estimate the number of large interconnected projects than to estimate the number of individual projects. When many items are kept together, the group errors are omitted. General financial estimates, for example, are more accurate than sectoral projections.

IV. FINANCIAL BASED ON MACHINE LEARNING PLATFORM

Figure 4 shows the latest financial plan and how Excel is used to build long-term predicting strategies and to use the prediction solutions to incorporate agiler solutions. Managers will utilize Excel data to expand the entire financial forecasting infrastructure, for example, risk and uncertainty forecasting, risk assessment, Monte Carlo simulation, and trying to learn about the financial, scientific, and modeling aspects. Assist finance professionals in their order to prepare and policy to incorporate uncertainty

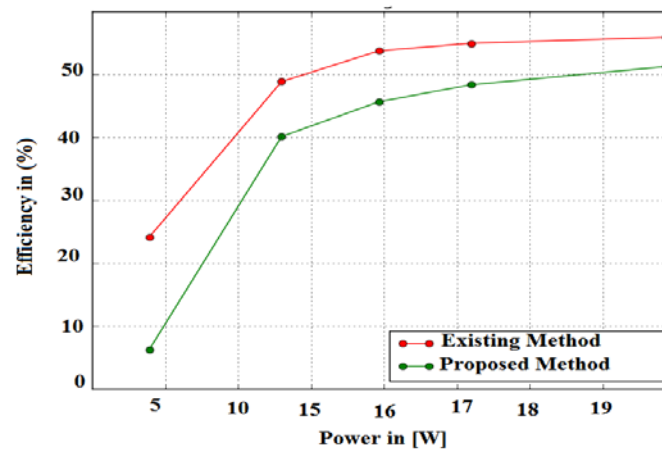


Figure 4 Machine learning efficiency analysis for financial forecasting

This book offers a comprehensive guide for the Simulation Model for Financial Statements, simple and consistent implementation, and a step-by-step system to implement an overall evaluation plan. The platform includes a complete operational model that can be individualized to deliver accounting methods or other key economic measures and tools that can easily be implemented to help them make key decisions.

4.2 TABEL 1:Exchange Rate Financial Prediction Platform Based On Machine Learning

	Number of Products (Per unit = 1)	Number of shares (Per unit = 1)	Growth of shares (Per unit = 1)	Efficiency in Percentages (%)
Gross Margin	392	873	1059	56
Net Income before	115	534	275	58
Net assets per share	115	157	36	57
Divided per share	16	17	18	69

Table 1 shows the most relevant pricing for a wider range of stocks and products—program manager for finance prediction, sensitivity analysis, master-analytic financial projections, and deep forecasts using appropriate models. Modern financial analyses require intensive preparedness, and modeling involves the development of the budgeting and forecasting systems of financial analysts. FPGA technology enables user-friendly and practical modeling of the Field-Programmable Gate Array for long-term benefits. Readers will examine their plans, methods, and particular requirements for accuracy improvement, smooth workflows, improved financial processes, and more effective research procedures. Machine learning can be easily customized to inform readers and to generate a variety of financial predictions and other important economic indicators to help them make intelligent choices.

4.1 EXCHANGE RATE CIRCUIT DIAGRAM FINANCIAL PREDICTION BASE ON MACHINE LEARNING AND FPGA PLASTIC PLACEMENT BASE FINE PREDICTION (FIELD-PROGRAMMABLE GATE ARRAY)

The global financial policy method is comprised of financial projections, research, and modeling. Financial predictions and analysis can be supported by money managers (including the uncertainty of their plans and budgets)—the transparent way to give longer-term income metrics a much more detailed meaning. In short, with a step-by-step process for global description readiness, the FPGA (Field-Programmable Gate Array) Execution Instructions. The site enables readers to quickly make decisions, use appropriate tools to promote the consequences of the recent financial crisis, provide financial information and other essential economic action, and provide financial signals and other critical financial indicators. Figure 5 shows how to adapt the production to the specific needs of financial modeling and provides a full operating model.

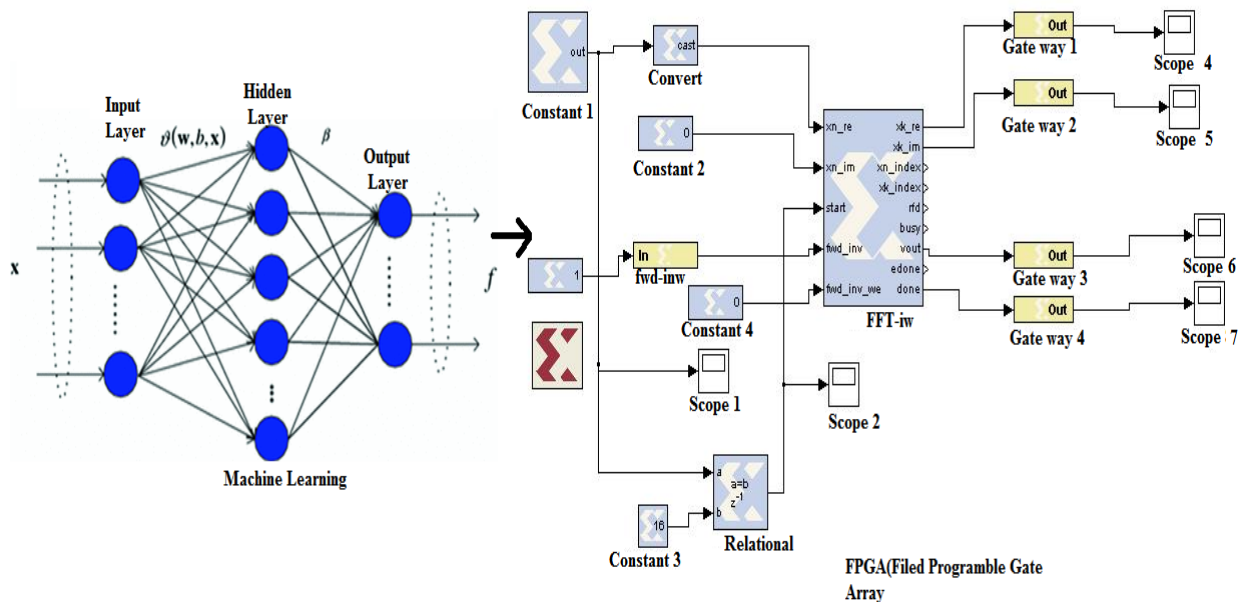


Figure 5 Exchange rate chart Financial platform based on the learning of machinery and FPGA (Field-Programmable Gate Array)

4.2 FPGA FIELD-PROGRAMMABLE GOTO PREDICTION (FIELD-PROGRAMMABLE GATE ARREY)

This machine training emphasizes prior knowledge, accurate review of and responses to problems, legal analysis, and problem-solving. Economic values, tools, and techniques form the basis of financial forecasts. FPGAs are valuable for the design of new technologies. Field-Programmable Gate Array. Trigonometry is used with quantitative as well as qualitative methods for dealing with uncertainty in the economic sector to achieve different results. Financial forecasts also provide possibly the best risk functions that help to optimize, average diversifying, and maximum trading tool estimates. The instantaneous or exceptional distribution of the probability is part of the learning process of the machine.

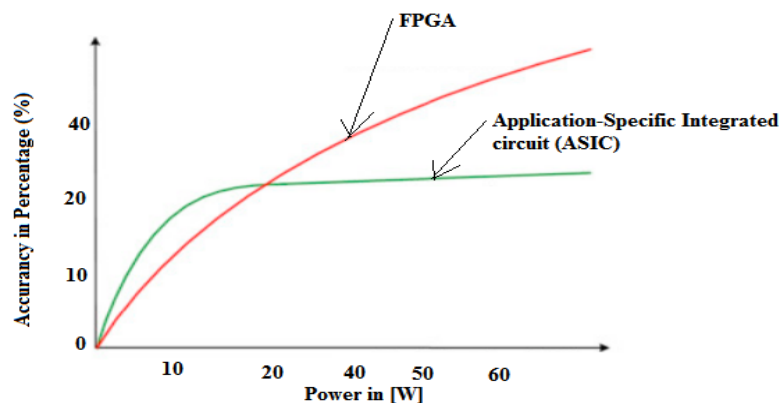


Figure 6 Accuracy analysis based on FPGA for financial predictions
(FIELD-PROGRAMMABLE GATE ARRAY).

The accessible functions are illustrated in Figure 6. A comprehensive evaluation of the actions taken, including a range of returns and expenditure forecasts for an investment portfolio, is now being conducted to achieve results from utility forecasts.

V. CONCLUSION

Financial forecasts are subordinate to transition, like many other aspects of economics. The climatic trends of an institution's establishment are based on flexibility. Long-term studies show that SMBs can decrease their use of tiny protrusions while developing at the same time. FPGA (Field-Programmable Gate Array) (Field Programmable Gate Array) (Field-Programmable Gate Array) (Field-Specific Injury methods to the strengths of the machine-learning entrepreneurs and large and microtechnology companies seem to be helpful. FPGA (Field-Programmable Gate Array) (Field-Practicable Gate Array) Price forecasts must be evidence-focused. The future of modern technology, because of the inherent error in results connected to the unceasing maturity of the authorities over time, is a noticeable substantial economic advance.

REFERENCES

1. J. Shi and L. Cheng, "Financial Crisis Dynamic Prediction Based on Sliding Window Technology and Mahalanobis-Taguchi System," 2011 International Conference of Information Technology, Computer Engineering and Management Sciences, Nanjing, Jiangsu, 2011, pp. 65-68, doi: 10.1109/ICM.2011.307.
2. Shaolin Lu, "Enterprise financial cost management platform based on FPGA and neural network," Microprocessors and Microsystems, 2020, 103318, ISSN 0141-9331, <https://doi.org/10.1016/j.micpro.2020.103318>.
3. M. İ. Y. Kaya and M. E. Karşilgil, "Stock price prediction using financial news articles," 2010 2nd IEEE International Conference on Information and Financial Engineering, Chongqing, 2010, pp. 478-482, doi: 10.1109/ICIFE.2010.5609404.
4. B. Xinzhong and H. Guangshuo, "A Study of Listed Companies' Financial Distress Prediction Using Rough Set Conditional Entropy Method," 2010 3rd International Conference on Information Management, Innovation Management and Industrial Engineering, Kunming, 2010, pp. 460-463, doi: 10.1109/ICIM.2010.117.
5. H. Yuzhu, L. Zengxin and H. Zaiqiang, "Financial Distress Prediction Model of Small and Medium-sized Listed Companies," 2011 International Conference on Information Management, Innovation Management and Industrial Engineering, Shenzhen, 2011, pp. 189-192, doi: 10.1109/ICIM.2011.50.
6. M. Zhou, "Short Term Prediction Method of Financial Crisis Based on Artificial Intelligence," 2020 International Conference on Intelligent Transportation, Big Data & Smart City (ICITBS), Vientiane, Laos, 2020, pp. 1026-1029, doi: 10.1109/ICITBS49701.2020.00228.
7. Y. Li, "Research on Financial Risk Prediction and Prevention Countermeasures Based on Big Data," 2019 11th International Conference on Measuring Technology and Mechatronics Automation (ICMTMA), Qiqihar, China, 2019, pp. 564-567, doi: 10.1109/ICMTMA.2019.00130.
8. L. Hengjun, "Rough Set Neural Network-Based Financial Distress Prediction," 2014 Sixth International Conference on Measuring Technology and Mechatronics Automation, Zhangjiajie, 2014, pp. 578-581, doi: 10.1109/ICMTMA.2014.141.
9. Y. Tian, W. Chen and S. Zhu, "The Coupling Degree Prediction between Financial Innovation Process and Innovation Environment Based on GM(1,1)-BPNN," 2014 Sixth International Conference on Intelligent Human-Machine Systems and Cybernetics, Hangzhou, 2014, pp. 257-260, doi: 10.1109/IHMSC.2014.70.
10. Y. Yang and C. Yang, "Research on the Application of GA Improved Neural Network in the Prediction of Financial Crisis," 2020 12th International Conference on Measuring Technology and Mechatronics Automation (ICMTMA), Phuket, Thailand, 2020, pp. 625-629, doi: 10.1109/ICMTMA50254.2020.00139.
11. Jaswanth K S, Dr. D. Stalin David, "A Novel Based 3d Facial Expression Detection Using Recurrent Neural Network", International Journal of Scientific Research in Computer Science, Engineering and Information Technology (JSRCSEIT), ISSN : 2456-3307, Volume 6 Issue 2, pp. 48-53, March-April 2020.
12. D Stalin David, 2020, 'An Intellectual Individual Performance Abnormality Discovery System in Civic Surroundings' International Journal of Innovative Technology and Exploring Engineering, Volume 9, Issue 5, PP.2196-2206.
13. D Stalin David, 2020, 'Machine learning for the prelude diagnosis of dementia', International Journal of Pharmaceutical Research, Volume 13, Issue 3, PP.2329-2335.
14. Stalin David D , Saravanan D, 2020, 'Multi-perspective DOS Attack Detection Framework for Reliable Data Transmission in Wireless Sensor Networks based on Trust', International Journal of Future Generation Communication and Networking , Volume 13, Issue 4, PP.1522–1539.
15. Stalin David D, Saravanan D, "Enhanced Glaucoma Detection Using Ensemble based CNN and Spatially Based Ellipse Fitting Curve Model", Solid State Technology, Volume 63, Issue 6, PP.3581-3598.
16. Stalin David D, Saravanan M, Jayachandran A, "Deep Convolutional Neural Network based Early Diagnosis of multi class brain tumour classification", Solid State Technology, Volume 63, Issue 6, PP.3599-3623.
17. R.Parthiban, Dr.K.Santhosh Kumar, Dr.R.Sathya, D.Saravanan, "A Secure Data Transmission And Effective Heart Disease Monitoring Scheme Using Mecc And Dlmnn In The Cloud With The Help Of Iot", International Journal of Grid and Distributed Computing, ISSN: 2005 – 4262, Vol. 13, No. 2, (2020), pp. 834 – 856.
18. R.Bhavya, G.I.Archanaa, D.Karthika, D.Saravanan, "Reflex Recognition of Tb Via Shade Duplicate Separation Built on Geometric Routine", International Journal of Pure and Applied Mathematics 119 (14), 831-836.
19. D Saravanan, R Bhavya, GI Archanaa, D Karthika, R Subban, "Research on Detection of Mycobacterium Tuberculosis from Microscopic Sputum Smear Images Using Image Segmentation", 2017 IEEE International Conference on Computational Intelligence and Computing Research (ICCIC).
20. U.Palani, D.Saravanan, R.Parthiban, S.Usharani, "Lossy Node Elimination Based on Link Stability Algorithm in Wireless Sensor Network", International Journal of Recent Technology and Engineering (IJRTE), Volume 7, Issue 6S5.
21. S.G.Sandhya, D.Saravanan, U.Palani, S.Usharani, "Handover Priority to the Data at Knob Level in Vanet", International Journal of Recent Technology and Engineering (IJRTE), Volume 7, Issue 6S5.
22. D.Saravanan R.Parthiban, U.Palani S.G.Sandhya, "Sheltered and Efficient Statistics Discrimination for Cluster Based Wireless Antenna Networks", International Journal of Recent Technology and Engineering (IJRTE), Volume 7, Issue 6S5.

23. D.Saravanan, Dr. K.Santhosh Kumar, R.Sathya, U.Palani, "An Iot Based Air Quality Monitoring And Air Pollutant Level Prediction System Using Machine Learning Approach – Dlmnn", International Journal of Future Generation Communication and Networking, Vol. 13, No. 4, (2020), pp. 925–945.
24. D Saravanan, R Parthiban, S Usharani, K Santhosh Kumar, "Furtive Video Recorder Using Intelligent Vehicle with the Help of Android Mobile", International Journal of Pure and Applied Mathematics, Volume 119, Issue 14, 2018.
25. Hong Liu and Zhi Liu, "Research and application of combination prediction model for financial equipment sales," 2010 2nd International Conference on Industrial and Information Systems, Dalian, 2010, pp. 390-393, doi: 10.1109/INDUSIS.2010.5565830.
26. Jao-Hong Cheng, Li-Wei Lin, Liang-Chien Lee and Jing-Han Chang, "How to reduce the false alarm rate beyond voting system for financial distress prediction," 2015 12th International Conference on Fuzzy Systems and Knowledge Discovery (FSKD), Zhangjiajie, 2015, pp. 892-897, doi: 10.1109/FSKD.2015.7382061.
27. Li Yun-Fei, Chen Yan-mei and Wang Jun, "The empirical research of listed company financial distress prediction based on financial information," 2011 2nd International Conference on Artificial Intelligence, Management Science and Electronic Commerce (AIMSEC), Dengleng, 2011, pp. 1659-1662, doi: 10.1109/AIMSEC.2011.6010552.
28. C. Cheng, W. Xu and J. Wang, "A Comparison of Ensemble Methods in Financial Market Prediction," 2012 Fifth International Joint Conference on Computational Sciences and Optimization, Harbin, 2012, pp. 755-759, doi: 10.1109/CSO.2012.171.
29. L. Li and H. Guo-hui, "Grey assessment and prediction of the financial agglomeration degree in central five cities," Proceedings of 2013 IEEE International Conference on Grey Systems and Intelligent Services (GSIS), Macao, 2013, pp. 187-190, doi: 10.1109/GSIS.2013.6714762.
30. J. Tang, "A Short-Term Prediction Approach of Financial Crisis Based on Artificial Intelligence Classification," 2020 12th International Conference on Measuring Technology and Mechatronics Automation (ICMTMA), Phuket, Thailand, 2020, pp. 990-994, doi: 10.1109/ICMTMA50254.2020.00214.
31. R. S. T. Lee, "Chaotic Type-2 Transient-Fuzzy Deep Neuro-Oscillatory Network (CT2TFDNN) for Worldwide Financial Prediction," in IEEE Transactions on Fuzzy Systems, vol. 28, no. 4, pp. 731-745, April 2020, doi: 10.1109/TFUZZ.2019.2914642.
32. H. R. Putri and A. Dhini, "Prediction of Financial Distress: Analyzing the Industry Performance in Stock Exchange Market using Data Mining," 2019 16th International Conference on Service Systems and Service Management (ICSSSM), Shenzhen, China, 2019, pp. 1-5, doi: 10.1109/ICSSSM.2019.8887824.
33. N. Kanungsukkasem and T. Leelanupab, "Financial Latent Dirichlet Allocation (Finlda): Feature Extraction in Text and Data Mining for Financial Time Series Prediction," in IEEE Access, vol. 7, pp. 71645-71664, 2019, doi: 10.1109/ACCESS.2019.2919993.
34. Lorenzo Servadei, Elena Zennaro, Tobias Fritz, Keerthikumara Devarajegowda, Wolfgang Ecker, Robert Wille, "Using Machine Learning for predicting area and Firmware metrics of hardware designs from abstract specifications", Microprocessors and Microsystems, Volume 71, 2019, 102853, ISSN 0141-9331.
35. Jayachandran, A., and D. Stalin David. "Textures and Intensity Histogram Based Retinal Image Classification System Using Hybrid Colour Structure Descriptor." Biomedical and Pharmacology Journal, vol. 11, no. 1, 2018, p. 577+. Accessed 12 Feb. 2021.
36. D. Stalin David, 2019, "Parasagittal Meningioma Brain Tumor Classification System based on MRI Images and Multi Phase level set Formulation", Biomedical and Pharmacology Journal, Vol.12, issue 2, pp.939-946.
37. D. S. David and A. Jeyachandran, "A comprehensive survey of security mechanisms in healthcare applications," 2016 International Conference on Communication and Electronics Systems (ICES), Coimbatore, 2016, pp. 1-6, doi: 10.1109/CESYS.2016.7889823.
38. Stalin David, D., Jayachandran, A. A new expert system based on hybrid colour and structure descriptor and machine learning algorithms for early glaucoma diagnosis. Multimed Tools Appl 79, 5213–5224 (2020). <https://doi.org/10.1007/s11042-018-6265-1>.
39. D. Stalin David and A.A. Jose, 2020. Retinal image classification system for diagnosis of diabetic retinopathy using SDC Methods. Artech J. Eff. Res. Eng. Technol., 1: 87-93.
40. D. Stalin David and T. Joseph George, 2020. Identity-based Sybil attack detection and localization. Artech J. Eff. Res. Eng. Technol., 1: 94-98.
41. David, D.S. and L. Arun, 2020. Classification of brain tumor type and grade using MRI texture and shape in a machine learning scheme. Artech J. Eff. Res. Eng. Technol., 1: 57-63.
42. David, D.S., 2020. Retinal image classification system for diagnosis of diabetic retinopathy using morphological edge detection and feature extraction techniques. Artech J. Eff. Res. Eng. Technol., 1: 28-33.
43. David, D.S., 2020. A novel specialist system based on hybrid colour and structure descriptor and machine learning algorithms for early diabetic retinopathy diagnosis. Artech J. Eff. Res. Eng. Technol., 1: 50-56.
44. David, D.S. and M. Samraj, 2020. A comprehensive survey of emotion recognition system in facial expression. Artech J. Eff. Res. Eng. Technol., 1: 76-81.
45. David, D.S. and L. Arun, 2020. Multi-view 3D face renovation with deep recurrent neural networks. Artech J. Eff. Res. Eng. Technol., 1: 64-68.
46. D.Saravanan, S.Rajasekaran, Dr. D.Stalin David, P.Hemalatha, Dr.U.Palani. (2021). Detection of Sick Cell Anemia from Microscopic Blood Images Using Different Local Adaptive Thresholding Techniques. Annals of the Romanian Society for Cell Biology, 6549. Retrieved from <http://annalsofscb.ro/index.php/journal/article/view/3254>.
47. Dr.BrahmaDesamViswanathan Krishna, Dr.G.Amuthavalli, Dr.D.StalinDavid, E. FantinIrudaya Raj, D.Saravanan. (2021). Certain Investigation of SARS-COVID-2-Induced Kawasaki-Like Disease in Indian Youngsters. Annals of the Romanian Society for Cell Biology, 1167–1182. Retrieved from <http://annalsofscb.ro/index.php/journal/article/view/4469>.