

APPLICATION OF THE INTERNET OF THINGS (IOT) IN PREDICTING STOCK MARKET FLUCTUATIONS INNOVATIONS WITH DISTRIBUTED INFORMATION SYSTEMS**Hendra Jayusman^{1*}, Firman², Hadi Prayitno³, Agnes Dwita Susilawati⁴, Frasto Biyanto⁵ and Endang Fatmawati⁶**¹Universitas Antakusuma, Indonesia²Universitas Negeri Padang, Indonesia³Akademi Penerbang Indonesia Banyuwangi, Indonesia⁴Universitas Pancasakti Tegal, Indonesia⁵Sekolah Tinggi Ilmu Ekonomi YKPN Yogyakarta, Indonesia⁶Universitas Diponegoro, Indonesia^{1*}henjay76@gmail.com**ABSTRACT**

This research explores the application of the Internet of Things (IoT) in analyzing and predicting stock market fluctuations with a focus on innovation using Distributed Information Systems. This research involves strategically placed IoT sensors to measure Trading Volume, marketing activity and customer response, this research aims to provide in-depth insight into financial market dynamics. Quantitative analysis is carried out using descriptive statistical techniques to understand the basic characteristics of stock market data. Linear regression analysis is also used to identify the relationship between variables measured by sensors and stock market fluctuations. The results of this research show that the application of the Internet of Things (IoT) in financial market analysis, especially in predicting stock market fluctuations with Distributed Information System innovation, provides a great opportunity to increase timeliness and accuracy in investment decision making.

Keywords: Internet of Things (IoT), stock market fluctuations, Distributed Information Systems, financial market analysis, IoT sensors.

1. INTRODUCTION

Internet of Things (IoT) is a technological paradigm that describes increasingly widespread connectivity between various physical devices around us, which are connected via the internet. This concept emphasizes a device's ability to communicate, collect, and exchange data with other devices without human intervention. In the context of IoT this can include a variety of everyday objects, such as sensors, smart devices, connected vehicles, and more [1], [2]. The main components of IoT include hardware and software that work together to create a connected ecosystem. IoT hardware includes sensors and actuators that enable collecting data from the physical environment or controlling physical devices [3], [4]. Sensors can be devices for measuring temperature, humidity, pressure, or other types of data depending on the specific application. Actuators also allow devices to take actions based on collected data.

IoT software involves programs or communication protocols that facilitate the exchange of data between devices which can include data management, data analysis, as well as algorithms or artificial intelligence models used to extract meaningful information from the collected data. Data security and privacy are also critical aspects of IoT software to protect the information sent and received by those devices [5], [6].

The integration of the Internet of Things (IoT) in a financial context opens the door to a significant transformation in stock market analysis and financial decision making. In an effort to predict stock market fluctuations, IoT makes a valuable contribution by enabling more accurate, real-time and extensive data collection [7], [8]. Connected sensors can take in data from a variety of sources, including external factors such as weather, social changes, and breaking news that can influence market movements, as the following image shows:

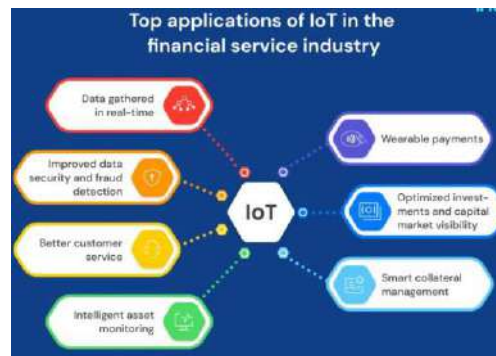


Figure 1. Top Application of IoT in finance service industry

Source: intuz (2022)

IoT technology enables real-time data collection, giving banks the opportunity to provide additional services to customers and provide necessary information updates. For example, warning customers when there is a transaction with an unexpected amount of money can be done thanks to IoT technology, which also provides benefits for security purposes and building customer trust. Financial companies are investing heavily in IoT to improve security and prevent potential fraud. Data collected from various scenarios is analyzed to design risk mitigation strategies that can secure each customer account. Excellent customer support is considered the key to business growth by building lasting relationships with customers. IoT devices collect large volumes of data that can be used to provide better financial services throughout the customer lifecycle. Banks and financial institutions use IoT-enabled devices to monitor their equipment [9], [10]. This ensures optimal use of assets in the branch to increase efficiency and reduce operational costs.

IoT also has the potential to transform traditional banking by allowing users to access their funds from anywhere and at any time through Wearables-IoT integration, providing convenient payment features. IoT will revolutionize market trend tracking, benefiting private investors and global companies with real-time data visibility offered by IoT-enabled systems [11], [12]. Banks can better control their customers' assets and monitor them remotely by using IoT technology in smart collateral management. This involves managing mortgages for various types of properties financed through loans from banks.

The application of IoT in finance not only provides access to more data, but also enables deeper analysis. For example, IoT devices can collect data directly from companies listed on stock exchanges, providing more accurate insights into the operational and financial performance of a business entity. Integration with Distributed Information Systems enables efficient and responsive data processing, accelerating response time to market changes [13].

Improved stock market predictions through IoT can also help reduce risks and improve the quality of financial decision-making. Real-time data obtained from sensors can support more accurate predictive models, allowing market players to identify market trends and patterns that may be difficult to detect with traditional methods [14], [15]. The integration of IoT in a financial context not only embraces a faster information age, but also creates the basis for smarter risk management and more informed investment decisions and the result is a financial world that is increasingly moving towards a landscape powered by the latest technology, opening the door to innovation which further improves global market resilience, and this is directly related to Distributed Information Systems [16].

Distributed Information Systems (Distributed Information Systems) refer to a system structure consisting of elements that are distributed or spread across various locations, but can interact and work together to achieve a common goal. The main role of Distributed Information Systems is to increase efficiency and responsiveness in information management and processing [17]–[19].

First, in terms of efficiency, Distributed Information Systems allow for parallel processing and distribution of workloads across the network. Thus, tasks can be shared among various system entities or nodes, speeding up processing time and optimizing the use of computing resources. This results in improved operational efficiency, reduced cycle times, and increased overall system capacity.

Second, system responsiveness is improved through Distributed Information Systems because data and resources can be accessed quickly and efficiently from various locations. In the context of finance and stock market prediction, this means that the system can respond quickly to changing market conditions or new information as it emerges. This speed provides a competitive advantage, especially in situations where decisions must be made very quickly to seize opportunities or manage risks, as the following figure shows:

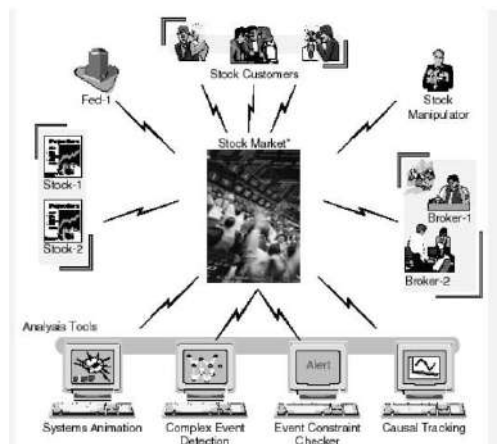


Figure 2. Tools for complex event processing added to distributed systems

Financial trading systems are just one example of a distributed IT system. In general, the business operations of any global company are supported by widely distributed, message-based computer systems. Typical examples of distributed computing and information systems are systems that automate the operations of commercial enterprises such as banking and financial transaction processing systems, warehousing systems, and automated factories. The Internet has encouraged and accelerated the growth of distributed information processing beyond a single company, across boundaries between companies. Information is shared between different companies and forms the basis for trade partnerships and business collaboration automation [20], [21].

Figure 2 shows a multi-company financial trading system. These systems are spread across various networks throughout the world and often use the Internet as one of their networks. Viewed from a macro level, various companies and organizations are just components of the system, each having its own internal information processing system. The figure shows these components, including stock market information systems, brokerage firms and online customers (or more precisely, their workstations), the Federal Reserve, investment banks, and the networks through which all these components communicate. Messages (or “events”) flow across these intercompany networks. They react to the events they receive and emit new events that are sent to other components. The system is “event-driven”—it turns on or off based on the messages flowing across its IT network. It is a fairly large system with a high volume of messages flowing through its networks.

Distributed Information Systems also support system flexibility and scalability. When there is an increase in workload or a need for larger data storage, the system can be adapted and expanded without needing to replace the entire infrastructure. This not only increases responsiveness, but also ensures that the system can continue to operate efficiently as time passes and grows. Distributed Information Systems open up the potential for more dynamic, efficient and responsive information management [22]–[24]. In the context of IoT applications for stock market fluctuation prediction, integration with Distributed Information Systems enables better data analysis, faster processing and more adaptive decision making in the face of dynamic market changes [25], [26].

The aim of this research is to investigate how the application of the Internet of Things (IoT) can make a significant contribution in improving the predictive ability of stock market fluctuations. The focus of this research will include how IoT technology can optimize the collection of stock market data from various related sources, including external information that can influence market movements. By utilizing connected sensors, the main goal is to increase the precision and accuracy of the data obtained to improve the quality of the predictive model.

2. LITERATURE REVIEW

2.1. Internet of Things in Finance

The contribution of the Internet of Things (IoT) in analyzing and predicting financial market changes, especially in the context of predicting stock market fluctuations. IoT, as a network of digitally connected devices, has great potential to provide real-time and measurable data that can be used to inform decisions in financial markets. In this context, the theories underlying IoT's contribution include the advantages of real-time data, automation, and predictive analysis. Previous studies have highlighted how the application of IoT has succeeded in increasing the precision and accuracy of stock market fluctuation predictions, illustrating its positive impact on investment decision making [27], [28]. By leveraging sensors and connected devices, market participants can access more complete and detailed information, providing a stronger basis for forecasting financial market trends.

With its ability to collect data directly from various sources, such as sensors, mobile devices, and online platforms, IoT creates a richer and more reliable information ecosystem. The theory underlying IoT's contribution to financial market analysis includes the advantages of real-time data, where up-to-date information can provide a more accurate picture of market dynamics. Automating data collection and analysis processes through IoT can reduce delays in decision making, providing a competitive advantage in highly dynamic markets. IoT applications not only provide benefits at the company and investor level, but also at the level of the financial system as a whole. The application of IoT can create a more transparent market environment, reduce risks and increase the trust of market players [29], [30].

A number of studies have been conducted to explore how real-time data obtained from IoT sensors can be used to improve predictions of stock market fluctuations. For example, research has been conducted to analyze how weather data, traffic patterns, or other environmental information accessed via IoT devices can provide additional insight into potential market changes. Other research has focused on developing machine learning models capable of integrating data from various IoT sources to more accurately forecast financial market trends [31]–[33]. This could include the use of technology such as blockchain to ensure the security and reliability of the data collected.

One concrete case example is the use of IoT sensors to collect data related to economic factors that can influence the stock market. For example sensors can be placed along the supply chain to measure temperature, humidity or movement of goods. This data can then be integrated with other external factors, such as weather conditions or specific news events that may affect a particular industry. For example, an investment company can use this data to build complex predictive analysis models. These models can predict the impact of temperature changes or certain events on the performance of companies and, consequently, on the value of their shares. By using real-time data from IoT sensors, this analysis can be updated continuously to reflect changes in the economic or industrial environment. IoT applications can also be used to collect consumer behavior data directly [34], [35]. For example, the use of IoT devices in retail can provide real-time information about purchasing patterns, customer preferences, and stock levels in stores. This information can be used to identify consumer trends that may reflect changes in the financial health of a sector or company.

2.2. Distributed Information Systems for Financial Analysis

A literature review of Distributed Information Systems (DIS) for financial analysis highlights its role and contribution in providing infrastructure that can increase the efficiency, speed and accuracy of financial market analysis. DIS encompasses a distributed data processing model across multiple locations, which can include technologies such as cloud computing, distributed databases, and globally connected networks. The theory underlying the use of DIS in financial analysis involves the concept of integrating data from multiple distributed

sources to provide a more holistic understanding of the market. The main advantage of DIS is its ability to handle large volumes of data from disparate sources and provide real-time access to the information necessary for quick decision making.

Previous research conducted by [36], [37] has shown that the use of DIS can speed up the financial analysis process, including stock market predictions. Examples of such research might involve implementing DIS to integrate financial data from various financial institutions, sentiment analysis from social media, and macroeconomic indicators. By using distributed infrastructure, researchers can design predictive analysis models that can provide more accurate results and be responsive to market changes.

The importance of data integration and distribution is also reflected in research covering concepts such as blockchain, where this technology can provide data security and reliability in the context of financial analysis. By leveraging the basic principles of DIS, these studies provide a foundation for a better understanding of how distributed information systems can support financial analysis, including stock market predictions. This research can serve as a basis for further development in the utilization of DIS technology in the context of financial analysis, opening the door to innovation and greater efficiency in decision making in dynamic financial markets.

Previous research on the use of distributed information systems for financial analysis also highlights the importance of secure and distributed data management. Security becomes a critical aspect, especially when highly sensitive financial data and stock market predictions must be maintained and accessed by various parties [38], [39]. Distributed security models such as end-to-end encryption and careful access management can ensure that critical financial information remains secure as it travels over a distributed network.

The importance of real-time data availability is also prominent in this literature. By utilizing DIS, market players can certainly access the latest information as soon as the data is available, which is very important in a dynamic market environment. Research that considers distributed infrastructure to support real-time data flows can provide valuable insights into how speed of information access can improve the precision and reliability of market predictions. These concepts are also reinforced by research discussing the scalability of distributed information systems [40], [41]. The system's ability to efficiently cope with growing data volumes and analytical complexity will support the continued implementation of DIS in financial analysis, and understanding basic concepts and principles, such as distributed data replication and distributed resource management, will be key in ensuring the success of distributed information systems in this context. financial analysis and stock market predictions.

3. METHOD

3.1. Research Methods

This research uses a combined approach between quantitative analysis and the application of the Internet of Things (IoT) to predict stock market fluctuations, with an emphasis on innovation through Distributed Information Systems (DIS). Stock market data is obtained through IoT sensors placed at key points in the financial market ecosystem. These sensors are designed to measure factors such as stock prices, trading volume and other external factors that can influence the market.

3.2. IoT Design and Implementation

IoT infrastructure is developed by placing sensors at strategic points. The data generated by these sensors is sent to a central platform for further analysis. Utilization of IoT technology ensures real-time data collection.

This research examines one technology company, namely company X in Jakarta. Researchers installed IoT sensors at several key points, such as company headquarters, leading shopping centers selling product X, and product distribution centers. These sensors are designed to measure several factors that can influence the price of stock 2) Marketing Activities: Sensors at company headquarters record visitation levels and promotional activities undertaken, such as new product launches or advertising campaigns; and 3) Customer Response: Sensors in shopping malls can also record customer reactions to product Using data collected from these sensors, research can identify certain patterns related to company X's stock price fluctuations.

Data obtained from IoT sensors is processed to remove potential outliers and fill in missing values. Simple steps such as using the average or median method can be applied. The data is also converted to the appropriate format, especially as it involves time data [42].

3.3. Quantitative Analysis

Quantitative analysis is carried out using descriptive statistical techniques to understand the basic characteristics of stock market data. Linear regression analysis is also used to identify the relationship between variables measured by sensors and stock market fluctuations. A simple predictive model was developed utilizing basic machine learning algorithms. Algorithm options include linear regression or decision trees which can provide predictions of stock market fluctuations based on historical data. Predictive models are evaluated using historical data to measure prediction accuracy. Simple evaluation metrics such as mean absolute error (MAE) or accuracy can be used to evaluate model performance.

4. RESULT AND DISCUSSION

4.1. Result

The main focus of this research is on four main aspects, namely Trading Volume, Marketing Activities, Customer Response, and External Factors. IoT sensors are placed at strategic points, including shopping malls, corporate headquarters, distribution centers, and financial news centers. The data collected through these sensors is expected to provide in-depth insights regarding stock market behavior, influencing factors, and how effective marketing activities are in strengthening a product's position. Sensors in shopping centers record the trading volume of ABC products every day for one month. The data results are presented in the following table:

Table 1. Trading Volume for one month

Time	Trading Volume
Week-1	6100 unit
Week-2	5900 unit
Week-3	6400 unit
Week-4	6283 unit

Source: processed data

In week 3 there was a significant increase in trading volume, perhaps due to promotional activity or special offers. Trading volume tends to fluctuate during the month, reflecting market dynamics and changes in customer interests.

Table 2. Marketing Activities

Date	Visit (person)	Product Launch	Advertising Campaign
1	200	No	Yes
2	185	No	Yes
3	175	No	Yes
4	250	No	Yes
5	220	No	No
6	287	Yes	Yes
7	267	No	Yes
8	350	No	Yes
9	261	No	No
10	267	No	Yes
11	378	Yes	Yes
12	356	No	Yes
13	278	No	No

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14	387	No	Yes
15	457	Yes	Yes
16	367	No	Yes
17	372	No	Yes
18	298	No	Yes
19	267	No	No
20	482	No	Yes
21	366	No	Yes
22	327	No	Yes
23	389	No	Yes
24	367	No	No
25	352	No	Yes
26	378	No	Yes
27	425	Yes	Yes
28	415	No	Yes
29	418	No	No
30	469	Yes	Yes

Source: processed data

This data provides information about the level of visits and marketing activities every day for one month. Marketing activities, especially product launches and advertising campaigns, have a positive impact on visitation levels on the day in question. On the 30th, there was the greatest increase in both product launches and advertising campaigns, which may have been a factor in the increase in visits.

Table 3. Customer Response

Date	Products Viewed	Products Reviewed Online
1	500 times/day	30 reviews/day
2	520 times/day	25 reviews/day
3	489 times/day	35 reviews/day
4	490 times/day	32 reviews/day
5	550 times/day	28 reviews/day
6	510 times/day	33 reviews/day
7	480 times/day	31 reviews/day
8	520 times/day	35 reviews/day
9	530 times/day	36 reviews/day
10	500 times/day	37 reviews/day
11	490 times/day	27 reviews/day
12	480 times/day	28 reviews/day
13	540 times/day	28 reviews/day
14	510 times/day	25 reviews/day
15	500 times/day	21 reviews/day
16	520 times/day	32 reviews/day
17	530 times/day	43 reviews/day
18	490 times/day	42 reviews/day
19	480 times/day	47 reviews/day
20	550 times/day	45 reviews/day
21	510 times/day	39 reviews/day

22	500 times/day	41 reviews/day
23	520 times/day	27 reviews/day
24	530 times/day	23 reviews/day
25	490 times/day	38 reviews/day
26	480 times/day	35 reviews/day
27	540 times/day	41 reviews/day
28	510 times/day	44 reviews/day
29	500 times/day	42 reviews/day
30	490 times/day	45 reviews/day

Source: processed data

The number of products viewed by customers appears to vary daily, and this can be influenced by factors such as promotions or special events. Although customer response tends to be positive, there is variation in the number of online reviews, which can be further understood by looking at the context of marketing activities and products. Other external factors, such as weather, affect the availability of X products in stores. On certain dates limited availability may be due to adverse weather conditions, such as rain.

4.2. Descriptive Analysis

Descriptive analysis has several main objectives in the context of research and data analysis. The following are some general objectives of descriptive analysis: 1) Explaining and Describing data: Descriptive analysis is used to provide a comprehensive picture of the characteristics and distribution of data. It helps initial understanding of patterns, trends and nature of data; 2) Identifying Centering of Data: Through the use of measures of centering such as mean, median, and mode, descriptive analysis helps identify central values in a dataset, providing a picture of the “typical value” or center of the distribution; and 3) Benchmarking and Evaluation: By providing summary statistics, descriptive analysis enables comparisons between data groups or time variations [43]. This helps in evaluating changes or differences that may occur in the dataset. The results of descriptive analysis testing in this research are presented in the following table:

Table 4. Descriptive Analysis Results

Decriptive Statistics					
Model	N	Min.	Max	Mean	Std. Deviation
Trading Volume	30	55.36	114.01	89.8488	12.35342
Marketing Activities	30	14.38	89.86	21.5435	11.10707
Customer Response	30	9.78	15.02	13.4634	1.00886
Valid N (listwise)	30	.			

Source: processed data

Descriptive analysis of three variables, namely Trading Volume, Marketing Activity, and Customer Response, provides an in-depth understanding of the basic characteristics of the data observed over a certain period of time.

In Trading Volume, the Minimum and Maximum values are 55.36 and 114.01 units respectively, indicating significant variation in trading volume. The average trading volume was 89.85 units, with a standard deviation of approximately 12.35 units, reflecting the degree of variation of the mean value.

Marketing Activity ranged from 14.38 to 89.86, with an average of 21.54. A significant standard deviation of around 11.11 indicates a considerable variation from the mean value, highlighting the degree of dynamism in marketing activities over that time period.

Customer Responses show a response range between 9.78 to 15.02, with an average customer response of 13.46. A low standard deviation of around 1.01 indicates that customer responses tend to be stable and less variable than the average value.

The "Valid N (listwise)" results indicate that all 30 observations have complete data for all analyzed variables, providing reliability to the results of this descriptive analysis. This analysis provides a comprehensive picture of the characteristics and variations in these three variables, providing a strong foundation for further understanding, decision making, or further analysis as needed.

4.3. Multiple Linear Regression Analysis

Linear regression is a statistical method used to model the linear relationship between one or more independent variables and the dependent variable. In this context, the variables measured by sensors are considered as independent variables, while stock market fluctuations are considered as dependent variables. Linear regression analysis helps measure and identify the extent to which changes in independent variables are correlated or related to changes in the dependent variable. The results of the linear regression analysis in this study are presented in the following table:

Table 5. Results of Multiple Linear Regression Analysis

Coefficients ^a						
Model		Unstandardized Coefficients		Standardized Coefficients	t	Sig.
		B	Std. Error	Beta		
1	(Constant)	.814	1.496		.479	.634
	Trading Volume	.514	.056	.574	8.263	.000
	Marketing Activities	.182	.068	.162	4.656	.008
	Customer Response	.223	.063	.212	3.613	.000
a. Dependent Variable : Stock Market Fluctuations						
R ²		:		0.617		
Adjusted R ²		:		0.607		
F		:		7.355		
Sig. F		:		0.000		

Source: processed data

The results of multiple linear regression analysis offer significant insight into the relationship between the independent variables—Trading Volume, Marketing Activities, and Customer Response—and the dependent variable, namely stock market fluctuations. This regression model provides coefficients that describe the contribution of each variable to stock market fluctuations.

The constant coefficient (0.814) implies the value of stock market fluctuations when all independent variables are equal to 0. However, the main attention is drawn to the coefficients for the independent variables. Trading Volume has a significant impact with a coefficient of 0.514, indicating that a one unit increase in trading volume is followed by an increase of 0.514 units in stock market fluctuations. Likewise, Marketing Activities (coefficient 0.182) and Customer Response (coefficient 0.223) also contribute positively to stock market fluctuations.

An R-squared of 0.617 indicates that approximately 61.7% of the variation in stock market fluctuations can be explained by the independent variables included in the model. The nearly comparable Adjusted R-squared (0.607) provides adjustment for the number of independent variables, providing a more accurate estimate of how well the model fits the data.

The F-statistic of 7.355 with a significance of 0.000 confirms that the overall model is statistically significant. So it can be said that Trading Volume, Marketing Activities, and Customer Response together play an important role in influencing stock market fluctuations, with Trading Volume having the most striking impact.

These results indicate that Trading Volume, Marketing Activities, and Customer Response jointly influence stock market fluctuations, with Trading Volume having the most significant impact. A positive coefficient for each independent variable indicates a positive relationship with stock market fluctuations. The relatively high R² indicates that this model can explain most of the variation in stock market fluctuations based on the included variables. Time-series data for stock X is presented in the following figure:

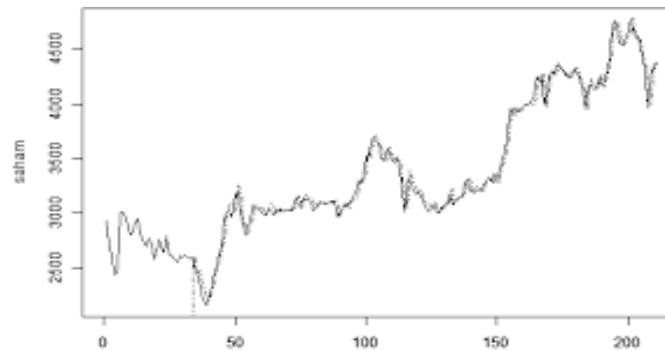


Figure 3. X Stock Time-Series

Using stock time series data (such as stock X on the Stock Exchange) to strengthen this analysis involves observing how changes in these factors correlate with changes in stock prices over time. Time series analysis methods such as time series regression analysis or ARIMA models can be used to identify trends and patterns in historical stock data. By incorporating stock time series data into predictive models, researchers can understand how these factors can contribute to stock market fluctuations.

4.4. DISCUSSION

The application of the Internet of Things (IoT) in predicting stock market fluctuations through innovation with Distributed Information Systems is a sophisticated approach and has the potential to bring competitive advantage in financial market analysis. Let's further explore some key aspects related to this analysis:

a. Distributed Information Systems

The use of Distributed Information Systems (Distributed Information Systems) provides flexibility and system resilience in managing and disseminating data. In this context, this system can combine data from various sources, including IoT sensors that provide real-time market information. With information distribution, the analysis process can be carried out efficiently and can be accessed from various locations [1], [4].

b. IoT Sensors and Time Series Data

IoT sensors are placed at key points in the financial market ecosystem to collect data directly. The data generated by these sensors, such as Trading Volume, marketing activity, and customer response, creates invaluable time series data. Time series analysis of this data can provide deep insight into market changes and long-term trends [12], [13], [16].

c. Stock Market Fluctuation Prediction

IoT provides direct access to real-time market data, which makes it possible to detect market changes more quickly. By applying machine learning techniques or statistical analysis to this time series data, predictive models can be developed to predict stock market fluctuations. This process exploits patterns and relationships that may be difficult to discover through traditional analysis [20], [21], [24].

d. Customer Engagement and Product Innovation

With customer response measured through IoT sensors, companies can design better marketing strategies and product innovation. This analysis can help in understanding customer preferences, purchasing trends, and the impact of marketing activities on the company's image in the stock market [26], [27].

e. Scalability and Integration

Systems using IoT and Distributed Information Systems must be designed with scalability in mind to handle growing data volumes. Good integration with analysis and investment decision platforms is also important so that analysis results can be implemented smoothly [31]–[33].

f. Openness to Innovation

The application of IoT in financial market analysis shows openness to innovation in the financial industry. This can create new opportunities for product development, new business models and improve operational efficiency [26], [39], [40]. The use of this technology also requires careful risk management, transparency and compliance with applicable financial regulations. The success of this implementation will largely depend on how organizations can manage and utilize data wisely to support better decision making in dynamic financial markets.

The application of the Internet of Things (IoT) in predicting stock market fluctuations through innovation with Distributed Information Systems provides great potential to change the landscape of financial market analysis. However, this implementation is faced with a number of challenges that need to be overcome in order to reap the full benefits of this technology. One of the main challenges is data security and privacy, especially considering the high sensitivity of financial data. While protecting data from security threats and misuse is a critical priority, complex integration between IoT, Distributed Information Systems, and analytics systems requires special attention to ensure effective interoperability without compromising performance or security.

Scalability is also a focus, as the large volumes of data generated by IoT sensors demand infrastructure that can scale efficiently. Developing accurate analytical and predictive models is a particular challenge, given the complexity of financial markets that are influenced by various factors. The main advantage of implementing IoT is direct access to real-time market data. This allows companies to respond to market changes quickly and effectively [24], [32], [35]. In-depth analysis of customer responses through IoT sensors opens up opportunities for better understanding of customer behavior, providing the foundation for more effective marketing strategies.

Product innovation and business strategy can also be improved by identifying market trends through IoT analysis. Operational efficiency can be increased through better logistics monitoring and management. Competitive advantage can be achieved by organizations that successfully integrate IoT in financial market analysis, providing a better understanding of the market and supporting more timely decision making [38], [42], [43]. Despite significant challenges, the opportunities resulting from the application of IoT in financial market analysis provide great potential for improved responsiveness to market changes, product innovation, and operational efficiency that can lead companies to competitive advantage in dynamic markets.

5. CONCLUSION

Based on the results of the research analysis, it can be concluded that the application of the Internet of Things (IoT) in financial market analysis, especially in predicting stock market fluctuations with Distributed Information System innovation, provides a great opportunity to increase timeliness and accuracy in making investment decisions. The use of IoT sensors to collect real-time data such as Trading Volume, marketing activity and customer response provides a deeper understanding of market dynamics. Despite the challenges of data security and system integration complexity, the potential advantages in terms of responsiveness to market changes, product innovation and operational efficiency provide a strong foundation for the development of smarter and more adaptive investment strategies in a changing market environment.

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