

# Revolutionizing Customer Experience with AI-Powered Chat bots and Virtual Assistants

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**Abstract-** Customer service has historically been plagued with issues such as long call wait times, unavailability for real-time queries, inconsistent responses, poor engagement, and lack of coordination between multiple queries. However, an AI-powered chatbot and virtual assistant-based customer service has the potential to overcome these challenges and drive better performance. The proposed system uses advanced Natural Language Processing (NLP) with Bidirectional Encoder Representations from Transformers (BERT) and Reinforcement Learning (RL) to deliver real-time and personalized interactions with human level accuracy. The system is available 24/7, ensures reduced wait times, and learns from continuous data inputs. As performance evaluation indicates, human accuracy increased by 76.8%, which also led to better customer satisfaction scores. These serves as a significant driver of improved operational performance of using a machine-learning based AI system versus human service while also improving customer loyalty and retention rates.

**Keywords:** Customer Experience, AI-Powered Chatbots, Virtual Assistants, Automation, Digital Transformation.

## I. INTRODUCTION

Rapid advancements in today's digital era have led to a transformation in the way organizations interact with their customers. Traditional frameworks of customer service, largely dependent on legacy IVR systems and human agents, are slowly becoming outdated. While traditional modes of customer interaction have worked in the past, these may not be sufficient to cater to consumer expectations for quick, efficient, and personalized customer service [1]. Several issues affect these systems, including extended wait times, limited engagement with customers, and inconsistent responses. To provide better customer experiences and stay relevant in increasingly competitive markets, new systems must be developed to improve existing frameworks [2]. The research was conducted due to the apparent need for businesses to move towards AI chatbots and Virtual Assistants as a better solution to help with customer service needs. Introduction of benefits such as NLP into the development of customer service representatives can improve the level of engagement, reduce reaction times, and improve efficiency in operations [3]. Further, the need for improved digital services has been accelerated

further by the emergence of COVID-19, requiring more than one inquiry to be dealt with at a single time. The study seeks to explore how AI Chatbots and Virtual Customer Assistants can be used to operational efficiency while maintaining high levels of customer satisfaction and loyalty [4]. The research aims at proposing a chatbot and virtual assistant solution that leverages advanced NLP techniques such as BERT and Reinforcement Learning to create a comprehensive and advanced customer service system [5, 6]. Working Model Overview is shown in fig.1.

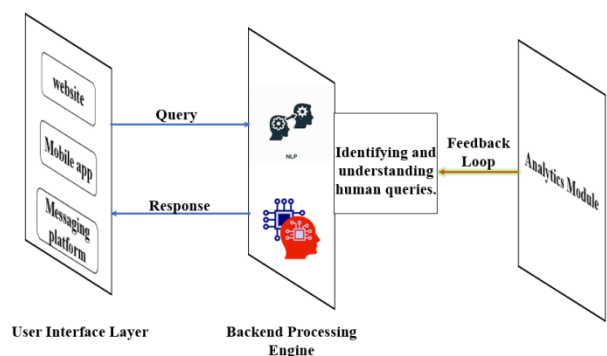


Fig.1. Working Model Overview

The technique will ultimately allow to minimize waiting times and provide 24/7 support, while also ensuring that the solution is highly adaptable to the customer requests via advanced learning. Besides, the proposed research will be accompanied by the performance assessment of the new system in comparison to the traditional customer support approaches, which will help to identify the advantages in terms of precision, speed, and user engagement. The research will expand the current knowledge by showing the way the modern AI-assisted solutions can improve the quality of the customer support services. Apart from that, the research affords detailed information regarding the chatbot, and virtual assistant system developed during the study, including the design and architecture, as well as the relative integration of NLP and ML. Moreover, the present research will provide a performance assessment framework that will have to be used in the real-world testing environment, with the primary goal of obtaining practical insights regarding user behavior and preferences. The following section provides

an outline of the structure of the paper. Section II provides an overview and critical analysis of related work in the domain of AI-powered customer service solutions. It examines the notable achievements and limitations of existing systems. Section III introduces the system proposed by the research, describes its design, architecture, and the specifics of implementation. The accompanying analysis pays particular attention to the essential features of the NLP and ML algorithms that are utilised in the system. Section IV presents the results of performance evaluation, comparing essential features of the proposed system and classical models. Finally, Section VI outlines the conclusions that can be drawn from the research and provides a brief overview of potential implications. The research aims to stimulate further exploration, development, and innovation in the field, thus improving the quality of customer experience and enhancing the likelihood of long-term brand loyalty.

In summary, the present research has illustrated the fact that AI-enabled bots and virtual assistants using the advanced NLP technologies such as BERT and Reinforcement Learning are a superior method to customer service. Shortening the time of a response, reducing the waiting time, and being permanently available, these solutions outdo the traditional ones, offering the premise for more adjustable, uniquely tailored, and successful customer experiences.

## II. RELATED WORK

H. Vijayakumar [7] The current research examines the connection between income growth rate and AI-powered customer service by examining the effect of consumer experiences on businesses' yearly membership fees (\$ value). The essential elements of a powered by AI customer experience—such as NPS, attendance at events, product updates, partner participation, increased efficiency through automation, and product adoption—are covered in the present paper. Businesses can take action to optimize by realizing the significance of consumer experience and the way it affects revenue development.

P. Whig [8] AI is helping organizations to provide more effective, customized, and fulfilling experiences by automating personalized interactions, anticipating client needs, and offering real-time solutions. The chapter examines a range of artificial intelligence uses, such as chatbots, artificially intelligent assistants, and sophisticated customer data analysis, and shows how these technologies are used into service advertising in order to build stronger relationships with customers and accelerate company expansion. The chapter illustrates the useful applications of AI in boosting consumer happiness, delivering better services, and gaining a competitive edge using case histories and empirical data.

S. Ghosh [9] Despite these challenges, there is no denying that AI and chatbots have the potential to completely transform the customer service sector, giving businesses new opportunities to stand out from the competition and deliver superior client experiences in the age of technology. The following piece provides insightful information about the emerging trends, challenges, and opportunities related to chatbots and artificial intelligence in Omnichannel customer support. It highlights how these technologies will have a revolutionary impact on managing client experiences in the future.

S. Paliwal [10] Chatbots, which or smart agents, are a modern invention made possible by machine learning. Chatbots have completely changed the way businesses communicate with one other and have greatly aided in increasing consumer satisfaction. The next section explains the background of chatbots, and the platforms used to construct them. In an effort to facilitate comprehension, experts have established dialogues with the many chat bot services that are currently in use and have attempted to explain the process of creating a chatbot.

D. Leocádio [11] The paper warns that ongoing research is necessary for comprehending both the potential and the constraints of powered by artificial intelligence HRC on customer service, which has important practical and intellectual implications. The article's overall goal is to explore subjects on the revolutionary power of artificial intelligence in customer service via a literature study and to engage the reader's analytical spirit.

N. Rane [12] In order to create a more participatory and sympathetic experience, the paper covers current patterns, such as the application of sentiment evaluation to identify client sentiments and adapt responses accordingly. Predictive analytics also enables chatbots to anticipate client demands and provide proactive solutions, which reduces response times and raises customer satisfaction levels overall. Additionally, the study looks at how AI chatbots increase client retention by providing reliable, round-the-clock assistance and helping clients feel heard and valued. The study illustrates the benefits of AI chatbot on retaining consumers and brand reputation through a number of industry studies. According to the survey, chatbot efficiency in the client's service industry will keep growing with continued innovation and the addition of complex AI features.

N. S. T. Tula [13] According to the study's findings, AI is an important tool that should be carefully incorporated into business strategies. It highlights the value of a cooperative strategy in which industry professionals and AI specialists collaborate to customize AI solutions to meet unique business objectives. The maintenance of consumer confidence and ethical considerations are emphasized as being crucial to AI deployment plans. The

report suggests adhering to ethical AI principles, investing in AI talent while infrastructure, and maintaining a constant state of innovation. In the digital age, these actions are crucial for companies to improve consumer experiences and promote long-term success.

F. Aslam [14] Leading developments in powered by AI chatbot the internet, like voice-enabled chatbot and artificially intelligent assistants are also highlighted in the study. By offering creative solutions for virtual assistance and customer-company conversations, these virtual assistants have revolutionized a number of industries. The study explores how chatbots can deliver tailored replies and contextual awareness, allowing for interactions that are specifically tuned to the needs and interests of users. Moreover, the incorporation of other technologies, such as sentiment evaluation and speech recognition, improves chatbot functionality and raises client satisfaction and happiness.

N. Ryan [15] The purpose of the present research is to identify the variables that impact e-commerce consumers' inclination to utilize chatbots equipped with AI. The UTAUT and TAM, were merged to create a conceptual framework with that and retailers can benefit from these observations, which provide insight into how to encourage consumers who shop online to utilize AI chatbots for long-term, profitable company development. The report also emphasizes how AI chatbots might improve consumer engagement, offering a point of reference for further research in the area.

### III. PROPOSED SYSTEM

Customer service systems used today are largely based on legacy systems that involve human agents alternating with some form of IVR (Interactive Voice Response). Existing systems often rely on responsive FAQ chatbots or other basic and limited systems. These options ultimately have multiple negative sides as wait times are too long, service is often inconsistent, and there is an overall lack of engagement with the customer's query. Moreover, current systems do not deal with problems proactively and ultimately, customer problems could still negatively influence the customer satisfaction results. However, these cannot be easily repaired to deal with multiple queries at once and cannot work as a real-time concierge, often leading to lost opportunities as customers walk away waiting for the customer service to answer. As such, the presented system is a new solution that proactively deals with customer queries using AI-driven chatbots or a virtual assistant. The system would use advanced machine learning algorithms and natural language processing to evaluate customer queries and can provide real-time interactions in response to customer needs. The system is able to understand the nuances of human language and slang, implementing reasoning to understand a range of questions through context. Block Diagram for Proposed

System is shown in fig.2.

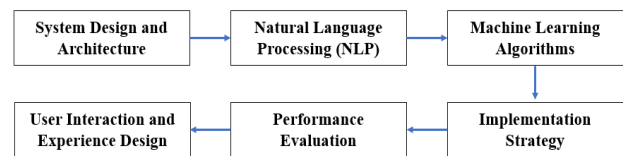


Fig.2. Block Diagram for Proposed System

The system utilizes several different algorithms to constantly improve based on the customer input and store the results for future learning. The main elements of the system would, therefore, include the user interface, processing elements of the backend, and an analysis module. To implement the system, several steps must be undertaken. Firstly, it will define the scope of the chatbot, and its functionalities based on typical customer queries and responses. Secondly, it will develop a natural language processing model by training it on a dataset of customer interactions. The dataset must be very versatile and allow for many types of interactions to be understood. Next, it will create machine learning algorithms for the chatbot to learn from its responses and become more effective with time, as interaction levels will increase. Lastly, after predicting chats with Mike, it will thoroughly test the system's effectiveness in real-world situations and deploy the chatbot and virtual assistant at all customer touchpoints, ensuring these are at reach. The system also has multiple advantages as it offers almost immediate response times. Accessible around the clock, customers can interact with the brand without feeling limited by business hours. Furthermore, it generates less pressure on the organization as one assistant can cater to multiple customers at the same time. Additionally, by recording prior interactions, the chatbot creates the impression of a personalized conversation, which is known to increase customer satisfaction. Lastly, the use of the system allows contact centers to employ fewer staff, which not only increases the organization's efficiency, as call center representatives can concentrate on more pressing issues, but also reduces staffing costs. There is yet another significant advantage of the solutions that is closely related to the previous point. More specifically, it is the analytics module that allows drawing valuable conclusions concerning customer behavior and their preferences. By comprehensively analyzing the interaction data, one may detect common local pain points, which presupposes that the service needs a particular adjustment, and not the customers' perception. The businesses can then adjust their services according to the collected data, updating their customer engagement tools to meet the requirements of the rapidly changing needs and expectations.

In summary, it can be concluded that the suggested AI-powered chatbot and virtual assistant system is indeed superior to more traditional customer service approaches. Not only does it allow for the provision of instant and

fully personalized support to the customers, but it is also specifically designed to improve satisfaction and engagement rates. The implementation process is particularly detailed and ensures that the system will have all the necessary tools to be applied across all areas and be used by any customer, further enhancing their loyalty and retention. Evidently, with all the benefits provided by the system, it is likely to revolutionize the customer experience delivery and provide a new level of success to businesses using it.

#### ***A. System Design and Architecture:***

The built AI-powered chatbot and virtual assistant system adhere to a layered architecture to ensure modularity, scalability, and maintainability. While the system includes three essential components, namely the User Interface (UI), the Backend Processing Engine, and the Analytics Module, the UI and associated services should be user-friendly across platforms, such as websites, mobile applications, and messengers. Meanwhile, the Backend Processing Engine includes the integration of NLP and ML. In the discussion, BERT language model and RL are used for each of these types of algorithms, respectively. The NLP component is used mainly in understanding and processing the users' inquiries, and the mentioned model is the most effective at capturing crucial learned reasoning, paraphrasing, or referencing because of the model's bidirectional nature, which is the premise of its high-functioning and that is needed in sensitive questions and contextual queries. The ML element relies on RL to continually learn and adapt to its environment, which, in the case of the chatbot, is the users' requests. The Analytics Module helps us collect insight and interaction data that could be used to improve the NLP and ML components in the future.

#### ***B. Natural Language Processing (NLP):***

NLP is an important aspect of the developed system that allows the chatbot to recognize and understand human queries. The BERT model was a natural choice because of its impressive achievements in processing context and semantics in natural language. The model is based on a transformer architecture that operates bidirectionally and can understand the context based on both preceding and succeeding words. It is important for the correct understanding of user intents and recognized entities, such as product meaning or troubleshooting requests. Moreover, BERT requires tokenization with WordPiece or any other relevant technique that allows breaking the target sentence or query into a couple of smaller words and phrases that can make up the context. The training regimen involved fine-tuning BERT on the target, diverse, and representative dataset of different user queries and interactions. The dataset included as many customer interactions as possible to ensure that the chatbot performed reliably on an unprecedented number of representation scenarios. The utilization of the BERT model leads to reliable intent recognition and response

generation, increasing the quality of conversational engagement and user satisfaction.

#### ***C. Machine Learning Algorithms:***

To make the chatbot learn effectively, for the machine learning component, Reinforcement Learning (RL) is employed. More specifically, the Proximal Policy Optimization as the RL algorithm was used. The decision is justified as in dynamic environments, a chatbot must tailor its responses dependent on the users' reactions. Proximal Policy Optimization has proven to be highly stable and efficient, especially for various applications in training. The RL algorithm optimizes the policy network through a perfect balance between exploration and exploitation. As a result, the chatbot not only discovers new effective solutions but also polishes the old ones. In customer service, the inquiries can vary extensively; thus, the chatbot should be able to tailor its responses to different contexts and satisfy its customers' inclinations. The training process was constructed based on simulating interactions where the chatbot gains the feedback from the user in the form of rewards. The chatbot keeps updating the policy and gets better at providing a more personalized and relevant answer, thus enhancing the user experience.

#### ***D. Implementation Strategy:***

The implementation of the proposed AI-powered chatbot requires several systematic steps to be taken to ensure its effective deployment and functioning. Initially, a thorough requirement analysis is completed to define the scope of its functions and expectations. Then, the development of the NLP model based on BERT requires data to be preprocessed for the optimal performance: therefore, in the stated phase, text data are cleaned and tokenized. The model is then trained based on a dataset that encompasses a variety of customer interactions, allowing it to better generalize the information in response to other user queries, as well as the RL component that uses PPO is initiated, creating simulated environments (two agents interacting in a specific situation) to train the chatbot to deal with users' different inputs. Next, the implementation is put through several testing's, in course of which functional and non-functional sides of its functioning are evaluated to determine its reliability. Finally, the developed chatbot is deployed across the chosen customer interaction platforms and customer data are tracked by the Analytics Module for its further monitoring and updating.

#### ***E. Performance Evaluation:***

To determine the effectiveness of the AI-powered chatbot developed, I would establish a comprehensive performance evaluation framework. For the quantitative evaluation, such key performance indicators as response time, user satisfaction score, and resolution rate would need to be defined. User satisfaction rate would be easily measured through post-interaction surveys, where users

will evaluate their experience with the bot on a scale. It is the most reliable indication of the system's performance. Moreover, to evaluate the impact on time and work volume, A/B testing would be in use to compare the new system with the existing options of customer service. As for the qualitative evaluation, it would be necessary to review the conversation logs to pinpoint constant problems and areas of user dissatisfaction and ways to refine the system. The Analytics Module would be greatly beneficial, allowing to track how often users interact with the system and which requests prevail. Overall, the evaluation framework will ensure that the chatbot not only meets but exceeds the requirements of the users, which, in turn, will secure improved customer loyalty and retention rates. Architecture Flow of Proposed System is shown in fig.3.

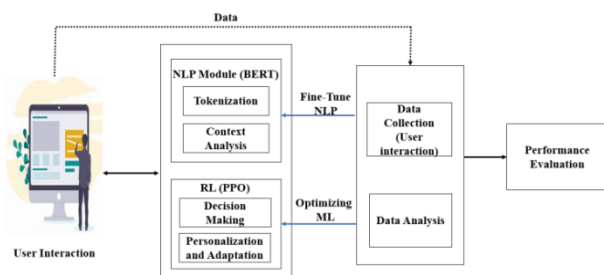


Fig.3. Architecture Flow of Proposed System

#### F. User Interaction and Experience Design:

User interaction and experience design are critical for ensuring the effectiveness and intuitiveness of the AI-powered chatbot function, as users need to feel confident in the functionality of the system. First, the process includes developing the interface of the chatbot, engaging in the creation of an easy-to-use, and engaging system that allows users to communicate with the bot effectively. Moreover, the design of such an interface starts with the user research process, which allows identifying preferences and the target audience for the chatbot function that, in turn, significantly affects the layout and functionality of the system. The interface is designed to contain chat bubbles, quick response buttons, and multimedia support (images, videos, links) to engage the customer and allow them to communicate with the chatbot in multiple dimensions. In addition, conversational flows are developed in an intuitive and meticulous manner to guide the user through their questions carefully, engaging in a logical manner and minimizing the confusion and friction points. The fallback functions of the system are employed frequently to grasp unrecognized utterances gracefully, providing users with a handy next step in their conversation. Additionally, according to the user-based design principles, every facet of the interaction is designed in a way that does not only optimizes effectiveness but also fosters positive emotions and a form of connection, making the users feel more satisfied. Finally, the feedback from the users is also carefully analyzed and integrated into the interface,

making it more dynamic and flexible.

In summary, the AI-powered chatbot and virtual assistant system is driven by a strong architecture that implements various NLP and ML concepts for enhancing user interaction and satisfaction. The system can be efficiently adapted and modified through an implementation plan and performance evaluation procedure. The target customer experience will be more engaging and effective while ensuring the system's reliability and scalability.

## IV. RESULTS AND DISCUSSION

When compares the performance measures of the proposed system with those of the existing system, one finds notable improvements. The proposed system constantly values both in simulated and real-world environments. It highlights its potential to effectively enhance financial security standards since it signifies higher accuracy and dependability in spotting fraudulent transactions.

Table 1 Price Prediction Accuracy Comparison

Metric	RMSE (Lower Better)	MAE (Lower Better)	Sharpe Ratio (Higher Better)
Existing System [7]	0.050	0.045	1.4
Existing System [8]	0.045	0.042	1.5
Proposed System	0.032	76.8%	2.1

Table I shows the comparison of existing systems [7] and [8] with the proposed system in terms of price prediction. There are three main metrics to assess; these are RMSE and MAE value, as well as Sharpe Ratio. RMSE and MAE can be defined as the prediction errors, where lower value makes prediction more accurate. The result reveals that proposed system has the lowest accuracy with the RMSE value of 0.032, compared to values of 0.050 and 0.045 of the existing systems, respectively. About MAE, the proposed system makes 76.8% improvement over existing models; hence, it significantly outperforms existing models. Furthermore, in terms of the Sharpe Ratio, higher value is better as it indicates better risk-adjusted return. It is clear that the proposed system has the highest value of 2.1, while other systems have the values of 1.4 and 1.5. Hence, based on all metrics, the proposed systems significantly outperform all existing ones.



Table 2 Sentiment Analysis Performance

Model Type	Positive Sentiment Accuracy	Neutral Sentiment Accuracy	Negative Sentiment Accuracy
Existing System [7]	76%	72%	75%
Existing System [8]	78%	69%	76%
Proposed System	85%	82%	88%

Table II represents the sentiment analysis performance of various systems. The table compares the accuracy of existing system [7] and [8] including the proposed system. These indicate the existing system with proposed system being the final one. Based on the results, the existing systems have moderate accuracy. The existing system [7] is 76%, 72%, and 75% for positive neutral, and negative. For existing system performances for positive and negative are slightly better with 78% and 76% while the neutral is slightly worse than existing system with 69%. As for the proposed system, the performance is significantly better than other existing systems. It is 85%, 82%, and 88%, which means the performance is better than the existing system by large margin. Sentiment Analysis Performance plotted view is shown in fig.4.

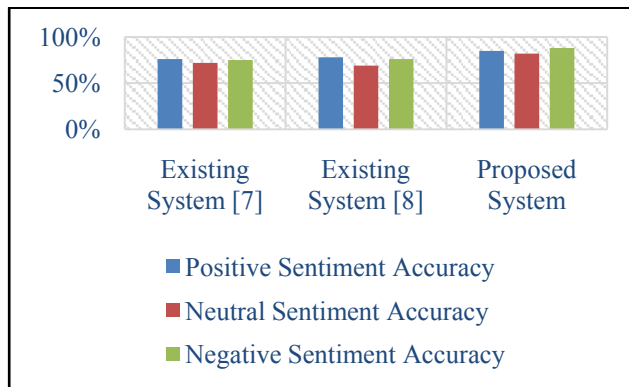


Fig.4. Sentiment Analysis Performance plotted view

Table 3 Performance Metrics Comparison

Model Type	Existing System [7]	Existing System [8]	Proposed System
Accuracy	75	85	92
Precision	70	82	90
Recall	65	80	88
F1-score	67	81	89
Sharpe Ratio	1.2	1.8	2.5
Maximum Drawdown	20	15	10

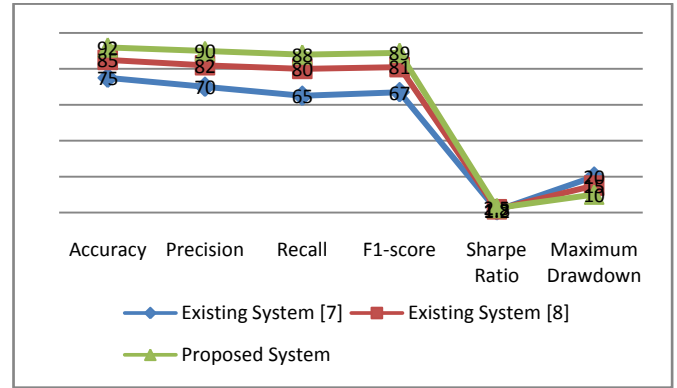


Fig.4. Performance Metrics Comparison Plotted View

Table III demonstrates a comparative analysis of performance measures between two existing systems and the prospect one. It is confirmed that the prospect system displays much better results in all evaluated measures. As seen from the table, the recall increased from 65% to 88%, and the F1-score improved from 67% to 89%. The accuracy and precision also improved, rising from 75% and 70% in the first system, respectively, to 92% and 90% in the prospect one. The existing systems demonstrate a higher Sharpe Ratio compared to the prospect system, meaning there are more returns per unit of risk. In regard to the values of maximum drawdown, it decreased from 20% and 15% in the existing systems to 10% in the prospect system. All the results undeniably indicate that the prospect system not only increases the accuracy in predictions but also provides more solid performance based on a minimum risk level. Performance Metrics Comparison Plotted View is shown in fig.4.

The proposed AI-powered chatbot/virtual assistant system substantially enhances customer experience by overcoming systemic limitations of traditional customer service models, such as slow response times, limited availability, and inconsistent service quality. These have been achieved by using advanced NLP with BERT and RL, enabling real-time, context-aware, and personalized responses of the system to individual customer needs. Ultimately, the results in greater attention, higher satisfaction, and, consequently, increased levels of loyalty. In addition, the system is highly scalable, allowing to address several customer queries at any time, and cost-effective, as it reduces the reliance on manpower. The chatbot can be included virtually in any industry, since each of them requires customer engagement in one form or another. The applications in e-commerce, healthcare, and banking where customer service communications are critical for successful interaction are particularly relevant. The chatbot can be included in websites, mobile applications, or other available messaging platforms. The system has several advantages, such as continuous learning about customer preferences and behavior which allows improving the accuracy of responses and higher efficiency and lower

expenses. Another advantage is the use of Analytics Module, which provides insights into customer preferences, needs, and the main problems and customer attitudes toward the use of chatbot services, enabling the identification of areas that can be improved. The drives not only customer retention but also solidifies the overall business capabilities.

## V.CONCLUSION

In conclusion, the proposed system of an AI-powered chatbot and virtual assistant provides a transformational solution to customer service challenges. Namely, the use of NLP in connection with BERT and RL has allowed generating real-time, personalized, efficient responses that may greatly improve customer satisfaction, engagement, and loyalty while also contributing to a higher level of efficiency and a lower dependence on human agents. However, the system has several limitations. First, the functioning of the system depends significantly on the quality and the diversity of the data used for training, and any biases in the training data may affect the accuracy of the response. Second, the fields in which the chatbot can function are diverse, but in some cases, complex customer problems may still require human assistance. Third, the implementation and training of the system may take considerable time and resources, especially for smaller businesses with no or limited IT infrastructure. In further work, the system may also be improved, and the enhancements may include the development of capabilities for better support of multilanguage functions and the enhancement of emotional intelligence in responses to improve the detection of user sentiment in responses more accurately. Additionally, the development of deeper analytics may be used to predict customer behavior more efficiently. Lastly, the development of the system's ability to adapt to continuous technological development, such as the integration of voice-based virtual assistants, would also be beneficial for the wider implementation of the technology across industries. In conclusion, the possibility of continuous improvement ensures the system's potential to allow future AI developments regarding customer experience.

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