



Intelligent Chatbot Ticketing System Using RNN and CNN

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Abstract: The Intelligent Chatbot Ticketing System leverages advanced artificial intelligence techniques, particularly Recurrent Neural Networks (RNN) and Convolutional Neural Networks (CNN), to automate and enhance the process of managing and responding to customer service tickets. This system aims to streamline the ticketing process by efficiently understanding and categorizing user inquiries, providing quick resolutions, and improving overall user experience. The RNN is employed to capture sequential patterns in user interactions, enabling the system to understand context and provide coherent, context-aware responses. Meanwhile, CNN is used for feature extraction from text inputs, helping the system identify important keywords and topics related to customer queries. This combination of RNN and CNN allows the system to not only classify and prioritize tickets effectively but also learn from past interactions to continuously improve its performance. The integration of these deep learning models ensures an intelligent, adaptive, and user-friendly ticketing system that reduces the manual effort required from customer service agents and provides timely assistance to users.

Keywords: Neural Networks, CNN, AI, Customer Service Automation, E-commerce Support.

1. Introduction

In recent years, the demand for efficient, scalable, and intelligent customer support solutions has grown significantly across various industries. Traditional methods of handling customer service, such as call centers and email support, often struggle with scalability, response time, and resource allocation. As a result, businesses are increasingly turning to automated systems, particularly chatbots, to address customer issues are logged, categorized, and resolved. However, managing and responding to these tickets efficiently requires a system capable of understanding the complexity and context of each customer request. This is where advanced machine learning techniques, such as Recurrent Neural Networks (RNNs) and Convolutional Neural Networks (CNNs), come into play.

RNNs excel at handling sequential data, allowing the system to understand the flow of conversation and context, while CNNs are particularly adept at extracting relevant features from textual data, such as keywords or patterns, to classify and prioritize issues. The Intelligent Chatbot Ticketing System Using RNN and CNN aims to combine these powerful techniques to create an advanced, automated solution for managing customer support tickets. By integrating RNN for sequence modeling and

CNN for feature extraction, this system is designed to efficiently categorize, prioritize, and respond to customer inquiries with minimal human intervention. The goal is to provide a seamless and intelligent user experience while reducing the burden on customer service representatives. Furthermore, this system can learn and adapt over time, improving its performance and accuracy based on past interactions, making it a scalable solution for businesses in various sectors such as IT support, e-commerce, and telecommunications. Through this approach, the proposed system addresses the core challenges of traditional ticketing systems, offering a robust solution that is both efficient and scalable.



Figure.1 ChatBot

2. Literature Survey

S. S. S. R. Depuru, S. S. R. Depuru, and L. Wang , (2022)" Design and Development of a Ticket Booking System Using Smart Bot", this paper presents the development of a smart chatbot designed to assist users in booking tickets. The chatbot leverages Natural Language Processing (NLP) techniques to interpret user inputs and provide relevant responses. While the study focuses on the application of NLP in ticket booking, it lays the groundwork for integrating advanced neural network architectures like RNNs and CNNs to enhance the chatbot's understanding and response generation capabilities [1].

S. K. S. Gupta, A. K. Maurya, and S. K. Singh, Chatbot Based Ticket Booking System for All Modes of Transport, Lecture Notes in Electrical Engineering, Springer, 2023, this research introduces a chatbot system capable of handling ticket bookings across various transportation modes. The system employs a voice-to-text input mechanism, converting spoken language into text, which is then processed to fulfill user requests. The study emphasizes the importance of tokenization and breaking down input into smaller tokens or keywords for accurate processing [2]. Although the paper does not explicitly mention the use of RNNs or CNNs, the methodologies discussed are pertinent to the development of intelligent chatbot systems that could benefit from such architectures.

A. Hussain and S. Athula , AI and Deep Learning-driven Chatbots: A Comprehensive Analysis and Future Directions this comprehensive analysis explores the role of AI and deep learning in the evolution of chatbots. The authors discuss various neural network architectures, including RNNs and CNNs, and their applications in enhancing chatbot functionalities. The paper provides insights into how these models can be utilized for natural language understanding and generation, which are critical components in developing intelligent ticketing systems [3].

S. K. Jain and M. Kumar," A Review on Chatbot Design and Implementation Techniques ", this review paper examines various design and implementation techniques for chatbots, comparing different models and approaches. The authors provide a comparative analysis between RNNs and CNNs, discussing their respective strengths and limitations in the context of conversational agents. The insights from this paper are valuable for understanding how to effectively integrate RNNs and CNNs in chatbot-based ticketing systems to improve performance and user satisfaction [4].

P. Dhankhar (2018), " RNN and LSTM based Chatbot using NLP", This paper discusses the development of a chatbot utilizing Recurrent Neural Networks (RNNs) and Long

Short-Term Memory (LSTM) units for natural language processing tasks. The study focuses on building an interactive chatbot capable of understanding and generating human-like responses by training on conversational datasets. The methodologies and findings presented are directly applicable to the development of intelligent ticketing systems that require context-aware and coherent interactions with users [5].

3. Proposed System

RNN Captures context and sequential patterns from text. CNN Extracts relevant features and patterns from the processed text. Hybrid Model Combines the strengths of RNN (sequence understanding) and CNN (feature extraction). Auto-generates responses based on classified intent. Can use rule-based templates or retrieval-based methods. If confidence is low, escalates to a human agent.

The proposed methodology offers several advantages over traditional ticketing systems. By using a combination of RNN and CNN models, the system is capable of understanding both the context of conversations and the specific features of customer issues. The RNN model excels at managing multi-turn conversations, making it particularly well-suited for environments where the flow of information is crucial for resolving issues [6].

This is in contrast to simpler rule-based systems, which often struggle with handling complex or lengthy interactions. The CNN model enhances the system's ability to classify tickets based on their content. By analyzing text at the character or word level, CNNs can detect key phrases and important patterns that may not be immediately obvious, thus improving the system's ticket categorization accuracy.

This allows the system to automatically route tickets to the right department, reducing the need for human intervention and speeding up the resolution process. However, the integration of these two models presents challenges [7]. For example, the RNN model requires significant computational resources due to the need to process long sequences of data. Additionally, the CNN model, while excellent at feature extraction, may require fine-tuning to ensure that it captures the most relevant aspects of ticket data. Training both models simultaneously requires a balanced approach to prevent overfitting and underfitting. Despite these challenges, the system's ability to learn from past interactions and continuously improve its responses is a significant benefit. The inclusion of reinforcement learning techniques can further enhance the system by allowing it to adapt based on user feedback, improving its performance over time.

3.1. Algorithm Approach

Recurrent Neural Network (RNN): Processes sequential data like sentences or queries. Maintains a "memory" of previous inputs to understand context. Understands the structure and flow of user queries. Good for modeling the natural language input.

Convolutional Neural Network (CNN): Typically used in image recognition, but very effective in text classification. Uses filters to extract important n-gram features from text (e.g., common word patterns). Identifies key features or phrases in the input [8]. Enhances the classification capability alongside RNN.

NaiveBayes ,SVM (Support Vector Machine),Random Forest These are typically used as a baseline or in hybrid models

3.2. Methodology

The proposed Intelligent Chatbot Ticketing System utilizes a hybrid approach that combines Recurrent Neural Networks (RNNs) and Convolutional Neural Networks (CNNs) to automate and streamline the ticketing process. The system consists of several key components: data collection, pre-processing, model training, and ticket categorization [9].

Data Collection and Pre-processing: The first step involves gathering a large dataset of customer queries and support tickets. These tickets are typically categorized based on the issue type, priority, and other relevant features. The data is pre-processed to clean and standardize the text, which includes removing stop words, special characters, and irrelevant information.

Tokenization: Stemming, and lemmatization are applied to prepare the data for analysis.

RNN Model for Sequential Data Processing: Once the data is pre-processed, an RNN-based model, specifically a Long Short-Term Memory (LSTM) network, is employed to handle the sequential nature of customer interactions. RNNs are capable of capturing the context and flow of a conversation, making them ideal for understanding multi-turn dialogues in a ticketing system. The LSTM model is trained to process sequences of messages, maintaining contextual information from past interactions to generate more accurate responses [10].

CNN Model for Feature Extraction: In parallel, a CNN model is used to extract important features from the ticket content. CNNs are particularly effective for identifying local patterns in the text, such as keywords related to specific issues (e.g., "login issue," "payment failure," etc.). The CNN scans the text in a sliding window fashion, allowing it to detect these critical features that help

categorize the tickets into predefined categories (e.g., technical support, billing inquiries, etc.).

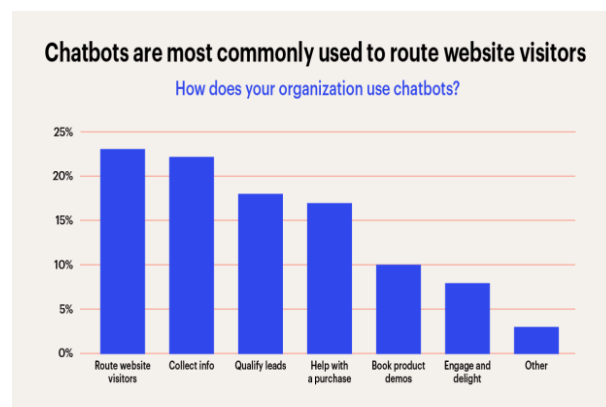


Figure.2 Visitor analysis using ChatGPT

Integration of RNN and CNN: After both models process the data, their outputs are combined. The RNN model provides a context-based response, while the CNN model categorizes the ticket based on the extracted features. The system then assigns the ticket to the appropriate support team or generates a response if it is a simple query.

Model Training and Evaluation: The system is trained on a labelled dataset of customer queries and tickets. The training process involves optimizing the parameters of the RNN and CNN models using backpropagation and gradient descent. Once trained, the models are evaluated using standard metrics like accuracy, precision, recall, and F1-score to assess their effectiveness in handling customer queries and ticket classification.

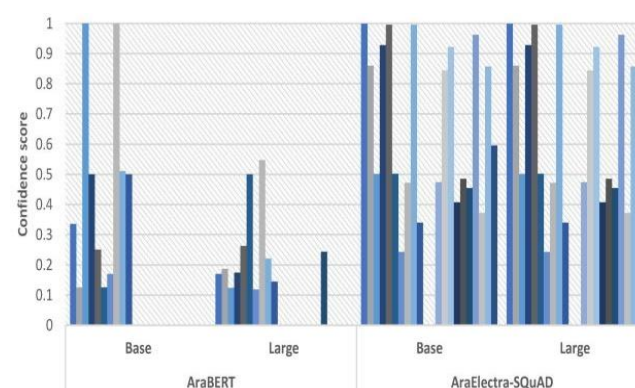


Figure.3 Visitor Active User Analysis using ChatGPT

4. Conclusion

The integration of Recurrent Neural Networks (RNNs) and Convolutional Neural Networks (CNNs) into the Intelligent Chatbot Ticketing System offers a highly effective and scalable solution for automating and improving customer service operations. By leveraging

RNNs, the system is capable of understanding the sequential nature of customer interactions, maintaining context across multiple exchanges, and generating coherent, context-aware responses. Meanwhile, the use of CNNs for feature extraction enhances the system's ability to identify important keywords and patterns within customer queries, enabling more accurate ticket categorization and prioritization. This combination of RNN and CNN allows the system to not only process and classify tickets efficiently but also learn from past interactions to continuously improve its performance.

The proposed system significantly reduces the manual effort required by customer service agents, while also providing quick and effective resolutions to users, thereby enhancing the overall user experience. The adaptability and learning capabilities of the system make it well-suited for industries like IT support, e-commerce, and telecommunications, where efficient customer support is essential. Future developments may further improve the system's capabilities, including refining model accuracy and incorporating advanced techniques like reinforcement learning or attention mechanisms to handle more complex customer interactions. Ultimately, this intelligent chatbot ticketing system has the potential to revolutionize the way businesses handle customer support, offering faster, more efficient, and highly responsive solutions.

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