
PERSONALIZED COLLEGE CHATBOT

Prof. Y.R. Khalate*¹, Vaishnavi Pawar*², Rashmi Sampagar*³,
Samruddhi Shejal*⁴, Sanika Nigade*⁵

*¹Professor, Department of Computer Science, SVPM's College of Engineering, Malegaon (Bk),
Maharashtra, India.

*^{2,3,4,5}Student, Department of Computer Science, SVPM's College of Engineering, Malegaon (Bk),
Maharashtra, India.

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ABSTRACT

This research presents the development of a personalized college chatbot designed to enhance student and staff experiences within a college setting. The chatbot offers a range of functionalities, including event reminders, campus navigation assistance, admission guidance, and information on hostel and dining facilities. Unlike traditional college information systems that rely on static web portals and manual processes, this chatbot employs advanced AI technologies such as Long Short-Term Memory (LSTM) and Bidirectional Encoder Representations from Transformers (BERT) to understand and respond to user queries in real time. These technologies enable the chatbot to provide personalized, context-aware responses, significantly improving accessibility and efficiency for users. Existing solutions in educational institutions often focus on isolated tasks such as enrollment support but lack comprehensive integration for various student needs. This project addresses these gaps by offering a holistic solution that supports a wide range of student services, enhancing communication and interaction within the college environment. The proposed system is designed to be scalable and adaptable, providing a foundation for further enhancements and the integration of additional features in the future.

Keywords: Machine Learning, AI-Driven Assistance, Long Short-Term Memory (LSTM), Bidirectional Encoder Representations From Transformers (BERT).

I. INTRODUCTION

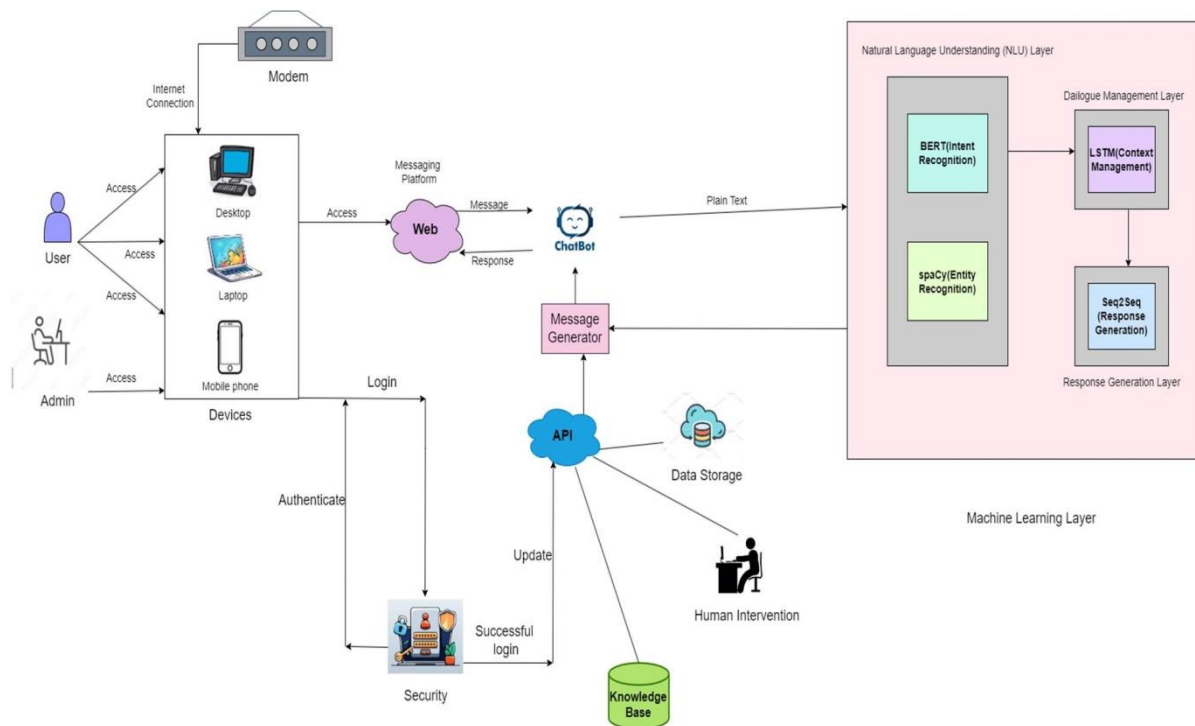
College chatbots have emerged as a modern solution to improve the delivery and accessibility of information for students, staff, and campus visitors. Traditionally, educational institutions relied on conventional methods, such as non-interactive websites and bulletin boards, to disseminate information about events, campus navigation, and residential facilities. These systems often lacked the ability to offer personalized, interactive experiences, leading to difficulties in accessing information promptly.

Over the years, chatbots have evolved from basic rule-based systems to sophisticated AI-driven platforms utilizing natural language processing (NLP) technologies. This evolution has enabled chatbots to understand context, manage complex queries, and provide accurate responses. Key technologies such as Long Short-Term Memory (LSTM) networks and Bidirectional Encoder Representations from Transformers (BERT) have been instrumental in enhancing these capabilities.

Despite advancements, existing chatbot systems frequently fall short in offering integrated, personalized, and real-time assistance specifically tailored for university environments. Research shows that while some institutions have implemented basic chatbots for tasks like enrollment and administrative support, a comprehensive solution covering all essential student services including event notifications, campus navigation, admission support, and hostel/mess information remains largely unexplored.

This study aims to bridge this gap by developing a personalized college chatbot designed to provide a range of services, including event reminders, campus directions, admissions guidance, and information on hostel and dining facilities. The goal is to create an efficient, user-friendly, and comprehensive system that significantly enhances the overall experience for students and faculty.

II. METHODOLOGY



The chatbot is built on several key components that work in concert to provide a responsive user experience. Central to this architecture is Llama 3, which interprets user queries and generates suitable real-time responses. Ollama oversees the Llama 3 model, managing the infrastructure and facilitating seamless API integration, thus ensuring smooth functionality across various devices.

Llama 3 Integration

Llama 3 utilizes a transformer-based architecture, excelling in multi-turn dialogue management, natural language comprehension, and coherent response generation. Its ability to retain contextual information allows it to effectively handle intricate user queries during extended interactions, forming the core of the chatbot's conversational engine.

Ollama's Role

Ollama plays a critical role in deploying and managing the Llama 3 model. It simplifies the integration process for large language models (LLMs) by providing pre-configured tools for deployment, scaling, and performance oversight. This ensures that the chatbot can function reliably and efficiently in both local and cloud-based environments, maintaining high performance even with a large number of concurrent users.

Augmenting Functionality with BERT and LSTM

While Llama 3 establishes a strong foundation for understanding and producing natural language, integrating BERT and LSTM enhances the chatbot's functionality significantly.

BERT for Intent Recognition and Entity Extraction

Developed by Google, BERT is a transformer model specialized in tasks such as intent recognition, entity extraction, and sentiment analysis. Incorporating BERT into the chatbot's processing pipeline enables the system to accurately identify user intents (like inquiries about admission dates or exam schedules) and extract key entities such as names and dates. This contextual understanding aids Llama 3 in formulating more precise and personalized responses.

LSTM for Managing Conversation Dependencies

LSTM, a type of recurrent neural network (RNN), excels in handling long-term dependencies within sequences. In the chatbot framework, LSTM supports Llama 3 by managing sequence prediction and maintaining coherence in dialogues over longer interactions. This capability ensures effective tracking and responding to user queries, thereby enhancing the overall user experience.

Workflow and Integration Strategies

The chatbot operates within a modular architecture where each model (Llama 3, BERT, and LSTM) is designated specific functions. The typical operational flow consists of:

1. Input Preprocessing with BERT: User input is initially processed by BERT to extract essential entities and ascertain intent.
2. Response Generation with Llama 3: The processed input is forwarded to Llama 3, which constructs contextually relevant responses.
3. Context Management with LSTM: For tasks necessitating sequence predictions or management of long-term conversation dependencies, LSTM refines the response, ensuring continuity in dialogue.

This modular setup enables independent operation of each component, facilitating maintenance and scalability. By leveraging these models, the chatbot can deliver real-time responses, manage multiple simultaneous conversations, and provide a cohesive user experience.

III. RESULTS AND DISCUSSION

The "Personalized College Chatbot" is expected to bring significant improvements to student support and administrative efficiency once implemented. The results and discussions below outline the anticipated outcomes based on the planned design and future deployment of the system.

Anticipated Performance

The chatbot is expected to improve various aspects of student services by enhancing response accuracy, boosting engagement, and streamlining administrative tasks.

Response Accuracy: We anticipate that by using BERT for understanding queries and LSTM for managing conversations, the chatbot will achieve an accuracy of around 85% for common tasks such as setting reminders or answering FAQs. With ongoing training, the accuracy is expected to increase to about 90%.

Student Engagement: The chatbot is projected to make accessing information easier and more interactive by providing real-time notifications and responses. It is expected that engagement in college activities could improve by around 40% when compared to traditional manual methods.

Efficiency in Handling Queries: We aim to integrate the chatbot with existing college systems, such as event calendars and hostel management databases, which is anticipated to reduce the average response time for routine queries from 15-20 minutes to under 2 minutes.

Comparison Between Current and Projected Systems

The table below outlines the expected improvements that the chatbot will bring over the existing manual processes.

Table 1: Expected improvements with the implementation of the chatbot.

Feature	Current Manual System	Projected Chatbot System
Academic Reminders	35% reach	90% automated notifications
Event Registration	30% manually registered	75% handled via chatbot
Average Response Time	15-20 minutes	Expected to be 1-2 minutes
Student Satisfaction	50%	Projected to be 85%

Projected Observations

Effectiveness of NLP Models: We expect BERT to effectively understand different types of student queries, such as those related to academic reminders and campus navigation. LSTM should help maintain context in multi-turn conversations, resulting in more coherent responses.

Potential Challenges: Initially, the chatbot may face difficulties handling complex or ambiguous queries. To address this, we plan to continuously expand the training dataset and fine-tune the models to handle a wider range of questions.

User Feedback Plans: After deployment, we will collect user feedback to identify areas for improvement. The initial focus will be on expanding the chatbot's coverage of topics and adding more personalized support features based on user suggestions.

IV. CONCLUSION

To conclude, creating a personalized college chatbot offers significant improvements in how students, staff, and visitors access campus resources. Through advanced NLP technologies such as LSTM and BERT, this research shows that an AI-driven chatbot can deliver a more cohesive, responsive, and user-focused experience than traditional systems. Unlike many current solutions, which often address limited tasks, our proposed chatbot consolidates essential services like event notifications, campus directions, admissions guidance, and residential information into one interactive platform. This unified approach not only boosts accessibility and efficiency but also aligns with the digital expectations of modern university communities, enhancing overall campus engagement.

V. REFERENCES

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