



Chatbot for College Website

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ABSTRACT—

The development of a Chatbot for the college website rcpit.ac.in aims to enhance user experience and streamline information access. Leveraging advanced technologies like the RAG-based system, Langchain model, and Ollama model, the project revolutionizes user interaction. It addresses existing system challenges through a defined problem statement and literature review. Detailed analysis informs design and development, with thorough testing ensuring reliability. Results and discussions reflect user feedback, emphasizing the project's significance in improving website interaction and setting a precedent for future innovations.

Keywords— Information encoding, document retrieval, Large Language Models, encoder-decoder architecture,

I. Introduction

The Chatbot project for rcpit.ac.in aims to enhance digital interaction within academia using advanced technologies like the RAG-based system, Langchain model, and Ollama model. It tackles issues such as fragmented information and lack of personalization to meet evolving user expectations for intuitive online experiences.

Motivated by the rapid growth of online resources, the project develops a robust chatbot system through thorough literature review and analysis of user requirements. It aligns with current technological trends to address present challenges while anticipating future needs, ensuring improved user engagement and satisfaction.

To concise, the project seeks to revolutionize academic website interaction by leveraging cutting-edge technologies and a user-centric approach. Through the implementation of a responsive and user-friendly chatbot, it aims to redefine accessibility, efficiency, and empowerment for users of rcpit.ac.in.

A. Motivation of Work

The motivation behind the Chatbot project for rcpit.ac.in is rooted in a commitment to enhancing user experience, improving information accessibility and maintaining technological competitiveness in academia. In response to the rapid digitization and evolving user expectations, the project aims to innovate and adapt digital platforms to meet diverse user needs efficiently.[1] Central to this motivation is the goal of streamlining information access and retrieval on the college website, addressing challenges users face in navigating the growing digital content landscape.

Driven by a desire to improve user engagement and satisfaction, the project leverages advanced technologies such as the RAG-based system, Langchain model, and Ollama model to provide a seamless and intuitive browsing experience. Through natural language processing and machine learning techniques, the chatbot offers personalized assistance, fostering a sense of community within the college ecosystem. Moreover, by embracing innovative technologies early on, the college demonstrates its commitment to delivering cutting-edge digital services and maintaining leadership in educational technology.

B. Proposed System



Fig. 1 This is a figure caption. It appears directly underneath the figure.

The proposed system for the Chatbot project at rcpit.ac.in is designed to revolutionize digital interaction within the academic website. By leveraging state-of-the-art technologies such as the Retrieval-Augmented Generation (RAG) system, Langchain model and Ollama model, the chatbot aims to provide a personalized, efficient, and intuitive user experience.[2] The system will serve as a comprehensive solution to address the challenges of fragmented information and lack of personalization, aligning with the growing expectations for advanced digital interfaces in academia.

The Retrieval-Augmented Generation system integrates a retrieval mechanism with a generative model, enabling the chatbot to provide accurate and contextually relevant responses. This system enhances the chatbot's ability to fetch precise information from the college's database and online resources. This model will be utilized to manage the flow of conversation and context.[4] Langchain's ability to handle complex dialogues ensures that the chatbot maintains coherent and meaningful interactions over extended conversations.

A specialized model for natural language processing (NLP) and understanding, Ollama will help the chatbot interpret user queries with high accuracy, facilitating nuanced and personalized responses.

II. Literature Review

A. Customized learning: Personalized attention to students enhances their academic performance as tutors can identify and address the areas where students struggle. The presence of personal educators for students with varying abilities can lead to the development of a greater number of skilled professionals. This approach allows students to gain a deeper understanding of their areas of interest.[6] Technology Mediated Learning (TML) refers to an environment where the learner's interactions with educational materials (such as readings, assignments, and exercises), peers, and instructors are facilitated through advanced information technologies. One aspect of TML is chatbot-mediated learning, which personalizes education by allowing students to interact dynamically with these bots for their studies. According to Thomas (2020), chatbots evaluate the students' understanding and provide tailored lectures accordingly. For example, the Summit Learning Project employs chatbots to identify students' weak points, adapt to their learning styles, and help them manage their modules. Additionally, chatbots conduct quizzes and report the results to tutors, who then offer immediate feedback to the students.[3] This process is supported through digital platforms. Spaced Interval Learning aids students in reinforcing what they have learned. It enables students to recall information they have memorized effectively. Super Memo is an application that assists students by reminding them of concepts they are about to forget. It uses an algorithm to track the frequency of learning sessions and revisits topics that have been previously covered.

Currently, educators often use multiple-choice questionnaires to assess students, simplifying the evaluation process. However, students can be better assessed based on their writing and composition skills, which can be developed through essay writing.[5] This method has been explored using an automated evaluation system where researchers implemented unsupervised machine learning for robotic assessment. They analyzed the performance of the system using a combination of term frequency-inverse document frequency (tf-idf) and cosine Euclidean distance metrics. A real-time study conducted on a group of medical students demonstrated that a web tutoring program significantly improved their test scores and cognitive efficacy, increasing them threefold as measured by Cohen's d effect size (95%) and confidence interval (CI)1.

It is a misconception among teachers that chatbots might replace them and lead to job losses. Instead, chatbots simplify tasks by addressing students' frequent queries and providing personalized assessments. This enables teachers to utilize the supplementary time for keeping up with the latest research. One of the pioneering educators in this field, Ashok Goel, developed a chatbot named Jill Watson. Jill was designed to respond to students' questions on an online forum, providing comprehensive information, including answers to technical doubts.

Apart from standalone chatbots, there has been a rise in the integration of chatbots into social platforms such as Facebook and Google Classroom. Chatbots used in educational contexts on Facebook, for example, have been studied based on their category, language, and development platform, with their efficacy evaluated using the Analytic Hierarchy Process (AHP). The effectiveness of educational chatbots also depends on their appealing design. Reeves and Nass (1996) demonstrated that many people perceive social platforms like televisions, computers, and the internet as peers, treating them with respect. This insight inspired some researchers to creatively impart knowledge by embodying influential historical figures. For instance, a 2014 study developed a talkbot named Freudbot using non-proprietary software such as AIML (Artificial Intelligence Markup Language) and ELIZA-like control features.

Freudbot interacted with learners as a famous historical personality. Although its results were neutral, it was considered a promising approach for the future of online education.

III. Project Design

The Chatbot project for rcpit.ac.in is envisioned as a transformative tool to enhance digital interactions within the academic environment, providing users with a streamlined, efficient, and personalized experience. The frontend interface is meticulously designed to be user-friendly, accessible through both the college website and a dedicated mobile application, accommodating the needs of students, faculty, and staff. By supporting text and voice inputs, the interface ensures inclusivity and flexibility in how users interact with the chatbot. The interactive chat window embedded within the site will allow seamless communication, enabling users to obtain information, ask questions, and receive support in real-time.

At the core of the system is the RAG-based architecture, which combines the strengths of retrieval and generative models to deliver precise and contextually relevant responses.[7] This dual approach allows the chatbot to fetch specific information from the college's extensive database and generate responses that are not only accurate but also engaging and conversational. The Langchain model plays a crucial role in managing the flow of conversation, ensuring that the chatbot maintains coherence and relevance even during complex or lengthy interactions. This capability is essential for addressing a wide range of user queries, from simple factual questions to more complex, context-dependent inquiries.

The Ollama model further enhances the chatbot's functionality by providing advanced natural language processing (NLP) capabilities. This model enables the chatbot to accurately interpret user queries, understanding nuances and contextual subtleties that are critical for providing meaningful assistance. The integration of the Ollama model ensures that the chatbot can handle a diverse array of questions and scenarios, delivering responses that are tailored to the individual needs of each user. This level of personalization is reinforced by the personalization engine, which creates user profiles based on past interactions and preferences, offering customized content and recommendations that enhance user engagement and satisfaction.

A dynamic and comprehensive knowledge base forms the backbone of the chatbot's information repository. Managed by an efficient Content Management System (CMS), this knowledge base includes a wide array of structured and unstructured data such as academic resources, frequently asked questions (FAQs), administrative information, and user-generated content. Regular updates to the CMS ensure that the chatbot has access to the most current information, which is essential for maintaining the relevance and accuracy of its responses. Additionally, the system's integration with existing academic management systems and third-party APIs allows it to pull relevant data, such as course schedules, faculty information, and event calendars, further enriching the user experience.

The project design also incorporates a robust feedback and improvement loop, where user feedback is continuously collected and analyzed to identify areas for improvement. This iterative process is supported by machine learning algorithms that enable the chatbot to learn from interactions and refine its performance over time. By continuously updating and improving based on real-world usage data, the chatbot remains responsive to evolving user needs and expectations. Security and privacy are also prioritized, with measures such as data encryption and adherence to data protection regulations ensuring that user information is safeguarded at all times.

The implementation plan is structured into several phases to ensure a thorough and systematic approach to project execution.[2] The initial phase involves gathering detailed requirements and conducting stakeholder consultations to inform the system design. This is followed by the development and testing phase, where the various components of the system are built and rigorously tested to ensure functionality and reliability. Deployment involves integrating the chatbot with the college's digital infrastructure and conducting user training sessions to facilitate smooth adoption. Finally, the maintenance phase involves continuous monitoring and updating of the system to address any emerging issues and incorporate new technological advancements.

By leveraging cutting-edge technologies and a user-centric approach, the Chatbot project for rcpit.ac.in aims to set a new standard for academic digital interaction. The system's design not only addresses current challenges but also anticipates future needs, ensuring that the college remains at the forefront of educational technology.[8] The project promises to significantly enhance user engagement, improve information accessibility, and foster a stronger sense of community within the college, ultimately empowering users and demonstrating the institution's commitment to technological innovation and excellence

IV. Operating System Selection:

Choose a suitable operating system for development and deployment. Common choices include Ubuntu (Linux) for its robustness and wide support for development tools, or Windows if it aligns better with institutional preferences. The selection of an operating system is critical at this point, since OS such as MacOS are not supporting custom training of models on there hardware since there ecosystem is dependent on Apple's support it maybe possible in future to select them. Current choices are WindowsOS+ Linux subsystem or entire use of linux system for latest models and there implementation.

| Ollama Zephyr 7B | |
|------------------------|--|
| Specification | Details |
| Number of Parameters | 7.2 billion |
| Model Architecture | Decoder-Encoder |
| Quantization | Q4_0 |
| File Size | 4.1 GB |
| License | MIT License |
| Platform Compatibility | macOS, Linux, Windows |
| Model Type | Language Model |
| Training Data | Mixed corpus (details not specified) |
| Inference Speed | Optimized for local execution |
| Primary Use Case | Language assistance and generation |
| API Support | Python, Javascript, REST API |
| Developer Tools | Ollama CLI, Integration with AutoGen and other libraries |

Fig2 Ollama Zephyr 7B Technical info

| BERT | |
|----------------------|---|
| Specification | Details |
| Number of Parameters | 110 million (BERT Base) / 340 million (BERT Large) |
| Model Architecture | Transformer-based Encoder-Decoder (Masked Language Modeling) |
| Layers | 12 (BERT Base) / 24 (BERT Large) |
| Hidden Size | 768 (BERT Base) / 1024 (BERT Large) |
| Vocabulary Size | 30,000 (BERT Base) / 30,000 (BERT Large) |
| Training Data | Wikipedia, Books, Blogs, News, etc. (English) |
| Pre-training Task | Masked Language Modeling (MLM), Next Sentence Prediction (NSP) |
| Applications | Text classification, Named Entity Recognition, Question Answering, etc. |
| API Support | Python (via Hugging Face Transformers), Java, etc. |
| Developer Tools | Hugging Face Transformers library, TensorFlow, PyTorch |

Fig3 BERT model info

1. Software Installation:

Install essential development software. This includes Python (a popular programming language for machine learning), Git (version control), and Docker (containerization). Set up

Integrated Development Environment (IDE) such as PyCharm, VS Code, or Jupyter Notebook for code development and testing.

2. Environment Setup:

Create a virtual environment using tools like virtualenv or conda to manage dependencies and ensure consistency across different systems.

3. Data Collection and Preprocessing:

Gather data from various sources including the college's databases, existing documents, and online resources. Clean and preprocess the data handling missing values, normalizing text, and converting data into suitable formats for training.

4. Model Selection:

Choose appropriate models for different components of the chatbot. For instance, the RAG-based model for retrieval and generation, Langchain for conversation flow, and Ollama for natural language understanding.

5. Training the Models:

Split the data into training and validation sets. Train each model component separately, using appropriate machine learning libraries like TensorFlow, PyTorch, or Hugging Face Transformers. Fine-tune pre-trained models (e.g., BERT, GPT) on the collected data to improve performance.

6. Model Integration:

Integrate the trained models into a cohesive system. Ensure that the RAG-based model interacts effectively with Langchain and Ollama models to manage conversation flow and understand user queries. Develop APIs or microservices for model components to facilitate communication between them.

7. Backend Development:

Develop the backend infrastructure to support the chatbot, including database management systems (e.g., MySQL, MongoDB) for storing user interactions and retrieved information. Implement a Content Management System (CMS) for maintaining and updating the knowledge base.

8. Frontend Development:

Design and develop the user interface, ensuring it is intuitive and user-friendly. Use web development frameworks like React.js or Angular for dynamic and responsive design. Embed the chat interface within the college website and develop a corresponding mobile application if necessary.

9. Testing and Validation:

Conduct rigorous testing of the integrated system. Perform unit tests, integration tests, and user acceptance tests to ensure all components work seamlessly together.

Collect feedback from a small group of users to identify any usability issues and refine the system accordingly.

10. Deployment:

Deploy the system using containerization tools like Docker to ensure it runs consistently across different environments.

Use cloud services such as AWS, Google Cloud, or Azure for scalable and reliable hosting.

11. Monitoring and Maintenance:

Set up monitoring tools to track system performance and user interactions. Tools like Prometheus and Grafana can be used for real-time monitoring and alerts.

Implement a continuous feedback loop, allowing the chatbot to learn from interactions and improve over time. Regularly update the system based on user feedback and emerging technologies to maintain its effectiveness and relevance

V. Implementation

The implementation of the Chatbot project for rcpit.ac.in is a meticulously planned process designed to revolutionize digital interaction within the academic environment. The first phase focuses on requirements gathering and system design. This involves extensive consultations with key stakeholders, including students, faculty, and administrative staff, to understand their needs and expectations from the chatbot. These insights inform the design of the system architecture, which outlines the integration of frontend and backend components, data flow, and security measures. Detailed specifications are drafted for each component to ensure clarity and coherence in the development process.

The development and testing phase begins with setting up the development environment on a suitable operating system, such as Ubuntu or Windows. Essential software installations include Python, Git, Docker, and machine learning libraries like TensorFlow and PyTorch. A virtual environment is created to manage dependencies, ensuring consistency across different systems. Data collection and preprocessing follow, where data is gathered from college databases and online resources, then cleaned and preprocessed using Python libraries like pandas and nltk. Model development involves training the RAG-based model for information retrieval and generation, the Langchain model for managing conversation flow, and the Ollama model for natural language understanding. Backend services are developed using frameworks like Flask or Django, while a responsive frontend interface is built using React.js or Angular. Rigorous testing, including unit, integration, and load tests, ensures all components function seamlessly together.

Deployment and user training constitute the third phase. The application is containerized using Docker to maintain consistency across environments and deployed on cloud platforms such as AWS or Google Cloud for scalability and reliability. CI/CD pipelines are set up using tools like Jenkins or GitHub Actions to automate deployment and updates. Comprehensive user manuals and training materials are developed to facilitate user adoption. Training sessions are conducted to familiarize students, faculty, and staff with the chatbot's functionalities, and ongoing support is provided to address any issues and collect feedback for continuous improvement.

The final phase focuses on monitoring and maintenance. Monitoring tools like Prometheus and Grafana are implemented to track system performance and user interactions in real-time, with logging and alerting mechanisms to quickly identify and resolve issues. A continuous improvement loop is established, analyzing user interactions and feedback to refine the chatbot's performance. Regular updates to the knowledge base and retraining of models with new data ensure the chatbot remains accurate and relevant. Security patches and software updates are applied regularly to maintain the system's robustness and security. This comprehensive approach ensures that the Chatbot project for rcpit.ac.in not only meets current user needs but is also adaptable to future technological advancements and user expectations.

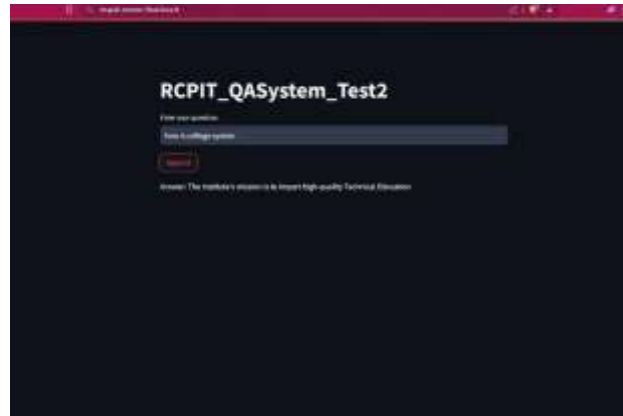


Fig 4. Testing the chatbot

Outcomes:

The outcome of the Chatbot project for rcpit.ac.in is a state-of-the-art digital interaction platform that significantly enhances user engagement and satisfaction by providing personalized, efficient, and intuitive access to information. Leveraging advanced technologies like the RAG-based system, Langchain model, and Ollama model, the chatbot delivers accurate and contextually relevant responses, streamlines information retrieval processes, and supports both text and voice inputs. This innovative solution not only addresses the challenges of fragmented information and lack of personalization but also positions the college at the forefront of educational technology, fostering a stronger sense of community and demonstrating a commitment to technological excellence and leadership.

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