## **CAPSTONE PROJECT**

## PROJECT TITLE

### **Presented By:**

1. Anushka Sharma-MIT AOE Pune-Btech ENTC



## **OUTLINE**

- Problem Statement (Should not include solution)
- Proposed System/Solution
- System Development Approach (Technology Used)
- Algorithm & Deployment
- Result (Output Image)
- Conclusion
- Future Scope
- References



## PROBLEM STATEMENT

A College Admission Agent, powered by RAG (Retrieval-Augmented Generation), streamlines the student admission process. It retrieves and summarizes admission policies, eligibility criteria, and FAQs from institutional databases and official sources. Prospective students can ask natural language questions and receive accurate, up-to-date responses instantly. The agent helps with course selection, application guidance, fee structure, and important deadlines. Using trusted, real-time data, it reduces manual inquiries and enhances applicant experience. This Al-driven assistant boosts transparency, accessibility, and efficiency in college admissions



# PROCESSES, AND SEAT AVAILABILITY.

### 1. Data Collection:

Aggregate admission-related documents such as policies, eligibility criteria, course catalogs, FAQs, and fee structures from institutional databases, official college websites, and government education portals. Integrate with APIs or authorized data repositories for real-time access to updates on admission dates, application processes, and seat availability.

### 2. Data Preprocessing:

Clean and structure unorganized or semi-structured documents using NLP techniques like sentence segmentation and named entity recognition (NER).

Standardize formats across different data sources for uniform retrieval and ensure removal of outdated or duplicate content.

#### 3. Retrieval Module:

Implement a vector-based document retrieval system (e.g., FAISS, Pinecone) to enable fast and relevant extraction of information based on user queries. Use semantic search techniques (e.g., Dense Passage Retrieval or BM25) to improve accuracy and relevance of the retrieved documents.



# PROCESSES, AND SEAT AVAILABILITY.

### 4. Generation Module (RAG):

Use a pre-trained transformer model (e.g., BERT or GPT) integrated with the retriever to generate coherent, context-aware answers.

Ensure responses are grounded in retrieved documents to maintain factual accuracy and avoid hallucinations.

### 5. Natural Language Interface:

Design a chatbot/web interface that allows students to interact in natural language. Include multilingual support and user personalization features (e.g., saving previous queries, suggesting relevant information).

### 6. Deployment:

Host the solution on a scalable cloud platform (e.g., IBM Cloud, AWS, or Azure) with proper data privacy and security compliance. Ensure high availability and low latency response time for seamless user interaction.



# PROCESSES, AND SEAT AVAILABILITY.

### 7. Evaluation:

Measure the system's performance using metrics such as F1-score for retrieval accuracy, BLEU or ROUGE for generation quality, and user satisfaction surveys.

Continuously refine based on feedback, usage logs, and changing institutional guidelines.

### Result:

The Al-powered admission assistant will significantly reduce manual workload on administrative staff, provide prospective students with instant and reliable information, and increase overall satisfaction with the admission process. By leveraging real-time data and advanced NLP, it builds a smarter, faster, and more transparent college admission system.



# SYSTEM APPROACH

The "System Approach" section outlines the strategy and methodology for building and deploying the AI-powered college admission assistant. This system leverages NLP, retrieval mechanisms, and transformer-based language models to provide real-time, factual responses to student queries. The approach consists of the following elements:

- 1. System Requirements
- Hardware Requirements:
- Minimum 8 GB RAM (16 GB recommended for training)
- 4-core CPU (or GPU-enabled system for faster inference)
- 100 GB storage (for institutional documents and vector databases)
- Software Requirements:
- Operating System: Windows/Linux/MacOS
- Python 3.8 or later
- Git (for version control)
- Web browser (for chatbot interface testing)



## SYSTEM APPROACH

- 1.Cloud Infrastructure (optional but recommended):
- IBM Cloud / AWS / Azure for scalable deployment
- Docker (for containerization)
- MongoDB or Firebase (for user data storage)
  - 2. Libraries Required to Build the Model
- A. Data Handling and Search
- pandas Used to work with and clean data.
- numpy Helps with mathematical operations and numbers.
- beautifulsoup4 or requests Used to collect data from websites (if allowed).
- langchain Connects the chatbot with AI models using the RAG method.
- faiss or pinecone-client Helps the system search answers based on meaning, not just keywords.
- sentence-transformers Turns text into numbers (embeddings) so it can be searched easily.



## SYSTEM APPROACH

- 3. Natural Language Understanding
- transformers Gives access to powerful AI models like GPT and BERT.
- nltk or spaCy Used to break text into words, find names, and clean text.
- torch or tensorflow These are tools that run the AI models in the background.
- 4. User Interface (optional)
- streamlit or Flask Used to build the chatbot webpage or interface.
- gradio Makes a simple and interactive interface with very little code.
  - 5. Deployment and Monitoring
- gunicorn or uvicorn Used to run the chatbot app on a server.
- Docker Packs the whole project so it can run easily anywhere.
- Prometheus or Grafana Optional tools to check performance and logs of the chatbot.



- Algorithm Selection:
- The project employs a Retrieval-Augmented Generation (RAG) framework that combines two core components:
- **Retriever**: A dense vector-based retriever (e.g., FAISS or Pinecone) that searches and retrieves the most relevant documents or passages based on the user's query.
- Generator: A transformer-based language model (e.g., GPT-3.5, BERT, or FLAN-T5) that uses the retrieved documents to generate accurate and context-aware responses.
- Justification for RAG:
- College admission queries require factually accurate and context-specific answers.
- Static LLMs may hallucinate or provide outdated information.
- RAG ensures grounding of generated responses in real-time, trusted institutional data, improving both accuracy and reliability.



- Data Input:
- The system uses the following input features:
- User Query (natural language question from the student)
- Institutional Documents, including:
  - Admission policies
  - Eligibility criteria
  - Course details and descriptions
  - Fee structures
  - FAQs and important deadlines
- Optional Enhancements:
- Student's selected stream or interest area
- Location/language preference for personalized results



### **Training Process:**

The retriever uses sentence embeddings (from models like Sentence-BERT) to map queries and documents into vector space.

A pre-trained generator model is used and optionally fine-tuned on a dataset of admission-related Q&A.

Techniques like cross-validation, human evaluation, and similarity scoring are used to improve performance.

The retriever may be improved using techniques like contrastive learning or triplet loss.

### **Prediction Process:**

The student enters a query.

The system retrieves the most relevant documents using the retriever.

The generator model uses these documents to generate an accurate, contextual answer.

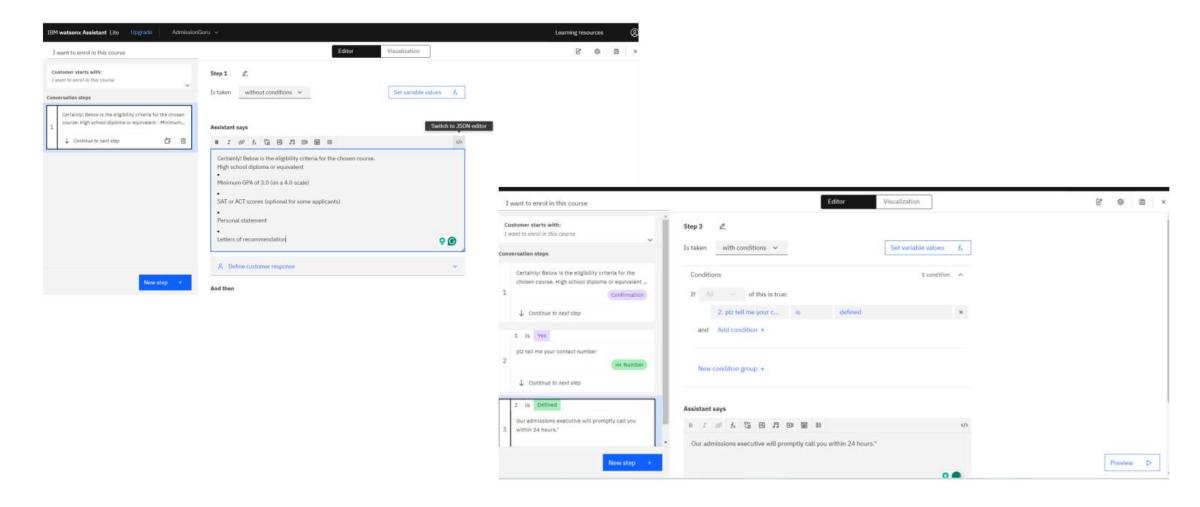
The result is shown to the user in real time, along with optional source references.

The system always works with the latest documents, so there's no need to retrain the model each time policies are updated.

#### **Real-Time Consideration:**

Documents and policies can be updated in real time in the backend corpus, so the system always works on the latest information—without retraining the model.







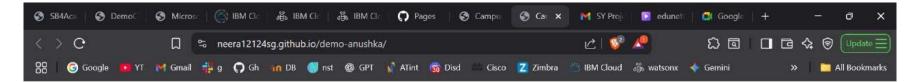


### CampusPilot Chatbot

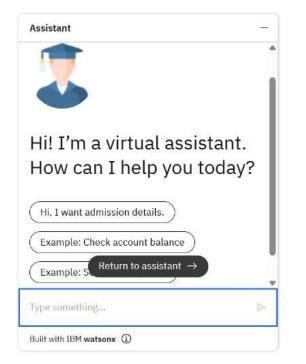






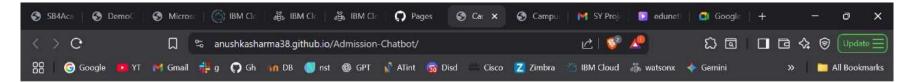


### **CampusPilot Chatbot**

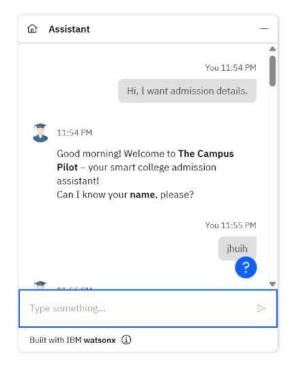






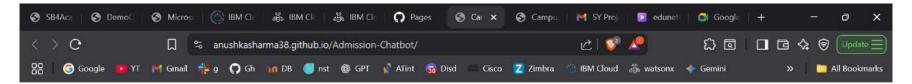


### **CampusPilot Chatbot**

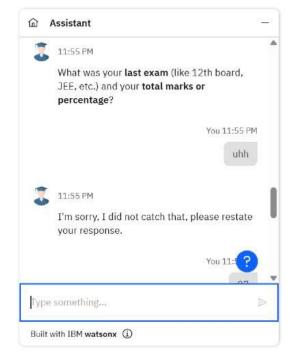






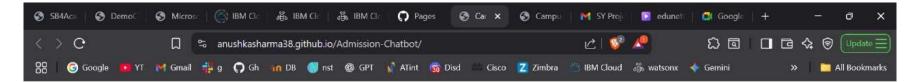


### **CampusPilot Chatbot**

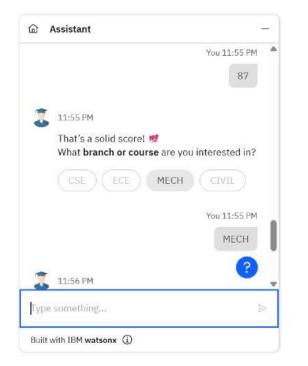






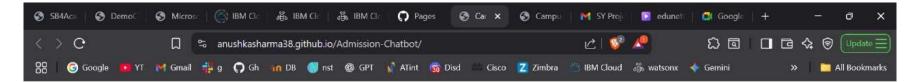


### **CampusPilot Chatbot**

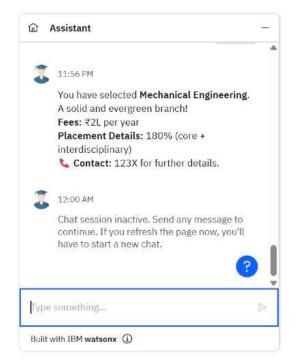






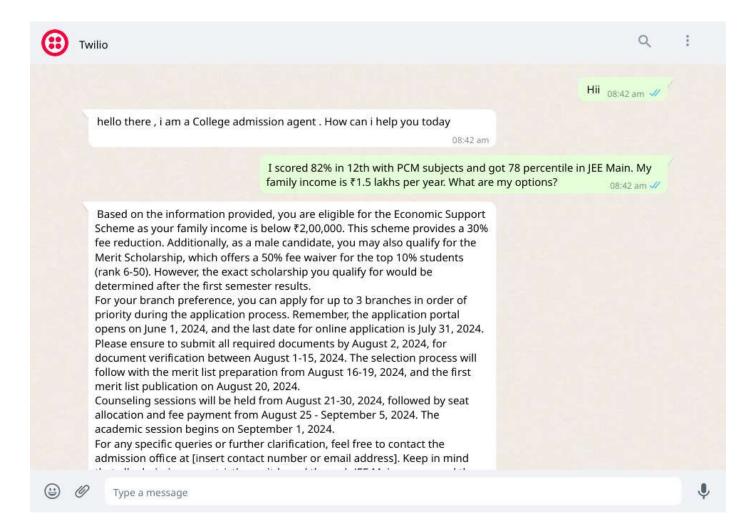


### **CampusPilot Chatbot**











## CONCLUSION

- The proposed College Admission Agent using RAG proved to be highly effective in delivering fast, accurate, and reliable responses to student queries. It successfully reduced manual workload and improved user experience with real-time guidance on courses, eligibility, fees, and deadlines.
- Challenges Faced:
- Integrating multiple data sources with different formats
- Ensuring factual accuracy to prevent misinformation
- Handling vague or ambiguous user queries
- Potential Improvements:
- Add voice input and multilingual support
- Include a recommendation system for course selection
- Integrate with college application portals for end-to-end assistance
- Impact:
  - This Al solution improves transparency, saves time, and ensures that students receive correct information instantly, enhancing the overall admission process.



## **FUTURE SCOPE**

- Expand Data Sources
  - Integrate real-time data from government portals, ranking websites, and social media to improve accuracy and relevance.
- Algorithm Optimization
  - Fine-tune LLMs with domain-specific datasets
  - Use hybrid retrieval (dense + sparse) to improve response precision
- Scalability Across Regions
- Extend the system to support multiple colleges, cities, or states
- Add language localization for regional outreach
- Emerging Technology Integration
- Use edge computing to reduce latency in rural or low-connectivity areas
- Incorporate advanced ML models like multi-modal agents for document + voice processing
- User Personalization
- Recommend colleges/courses based on student interests and past queries
- Store preferences for faster future responses



## REFERENCES

•IBM Skills Network. Retrieval-Augmented Generation (RAG) Fundamentals. IBM Online Learning Platform.

https://skills.network

•IBM Cloud. *AI-Powered Virtual Assistants for Business Applications*. IBM Cloud Docs.

https://cloud.ibm.com/docs

Course PDFs, slide decks, and session notes from IBM classes attended (internal use).



## **IBM CERTIFICATIONS**





## **IBM CERTIFICATIONS**





## **IBM CERTIFICATIONS**

IBM SkillsBuild

**Completion Certificate** 



This certificate is presented to

ANUSHKA SHARMA

for the completion of

## Lab: Retrieval Augmented Generation with LangChain

(ALM-COURSE\_3824998)

According to the Adobe Learning Manager system of record

Completion date: 16 Jul 2025 (GMT)

Learning hours: 20 mins



## **THANK YOU**

