

RAG CHATBOT FOR EFFICIENT RETRIEVAL OF INFORMATION



**PROBLEM STATEMENT TITLE - RETRIEVAL OF DATA FROM
KNOWLEDGE BASE OF COLLEGE WEBSITE AND STUDENT
RESOURCE BOOK.**

IDEA : AI –DRIVEN RAG CHATBOT FOR EFFICIENT & QUICK RETRIEVAL OF INFORMATION

What is the current problem?

- Multiple documents of policies, schemes & guidelines.
- Very scattered in different languages.
- Exists in multiple formats including scanned pdfs, & word documents.
- Manual, time consuming & inefficient search for information retrieval.

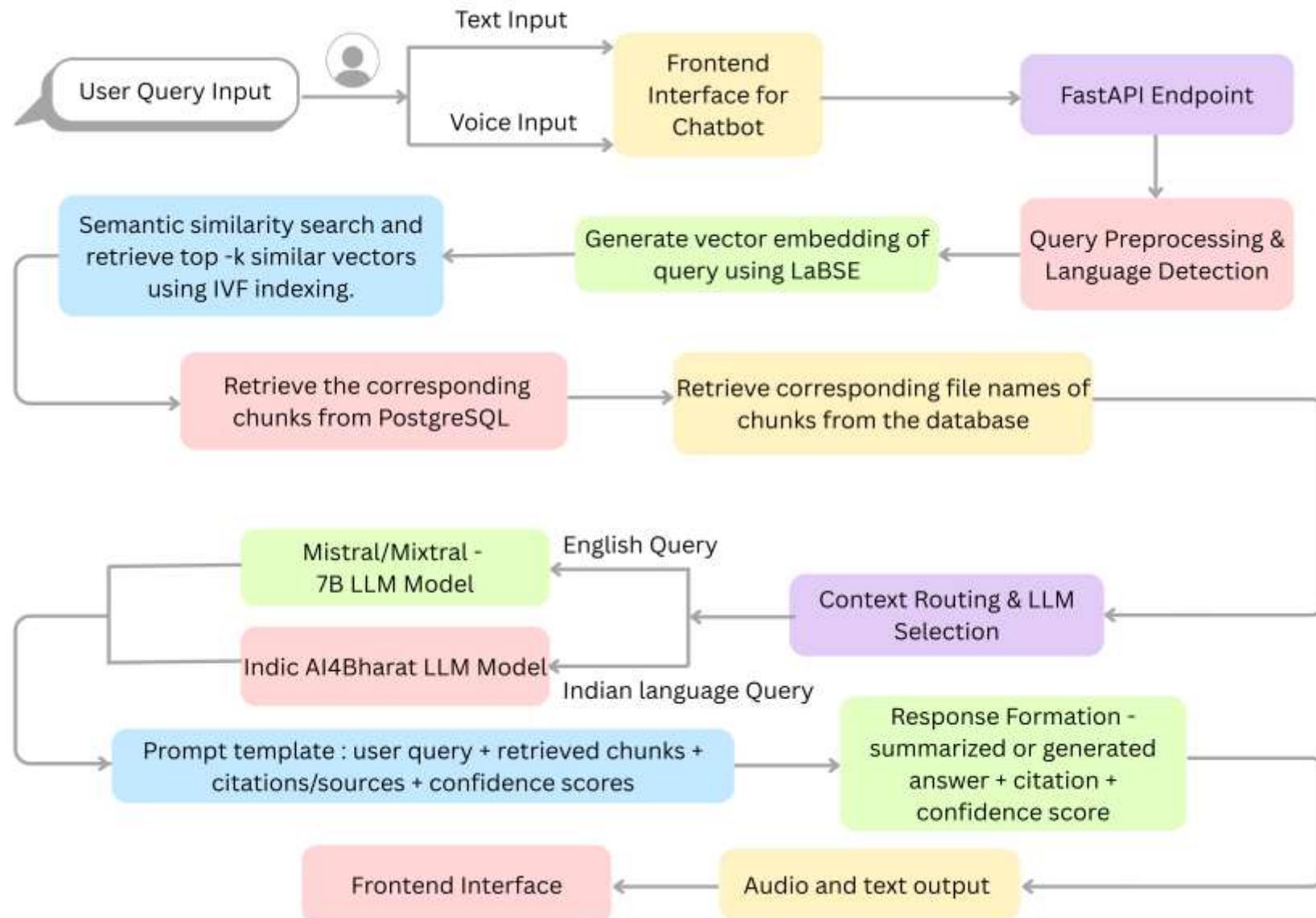
How we solve the problem?

- AI-powered Retrieval-Augmented Generation (RAG) System specifically designed for the MoE department.
- Supports multilingual queries & outputs including Indian languages.
- Gives summarized, citation-backed & confidence-rated answers to natural language queries, not just keywords.

Why our approach is unique & innovative?

- **Multilingual Understanding** for Indian languages.
- **Voice interaction** allowing users to interact hands-free.
- Can deal with **multiple document formats** with OCR.
- **Explainable AI** with confidence scores & traceability to source files.

TECHNICAL APPROACH



TOOLS & FRAMEWORKS



FEASIBILITY AND VIABILITY

HOW OUR SOLUTION IS DEPLOYABLE AND SCALABLE

No Licensing Costs

Uses all proven open - source frameworks and tools, making it free.



01



02

Operational Feasibility

Easy to use chatbot with simple interface, hence no learning curve



03

Easily scalable

Vertically and horizontally scalable with deployment to cloud services and GPUs.

04

Extendable to other departments

Can be fine tuned for other MoE departments without breaking core design

Problem



System



Challenges

Data scattered across multiple file formats (pdf, scanned pdfs, word files etc.).

Latency in LLM responses for complex queries.

May require GPU for large-scale use for more computational resources, maximizing speed.

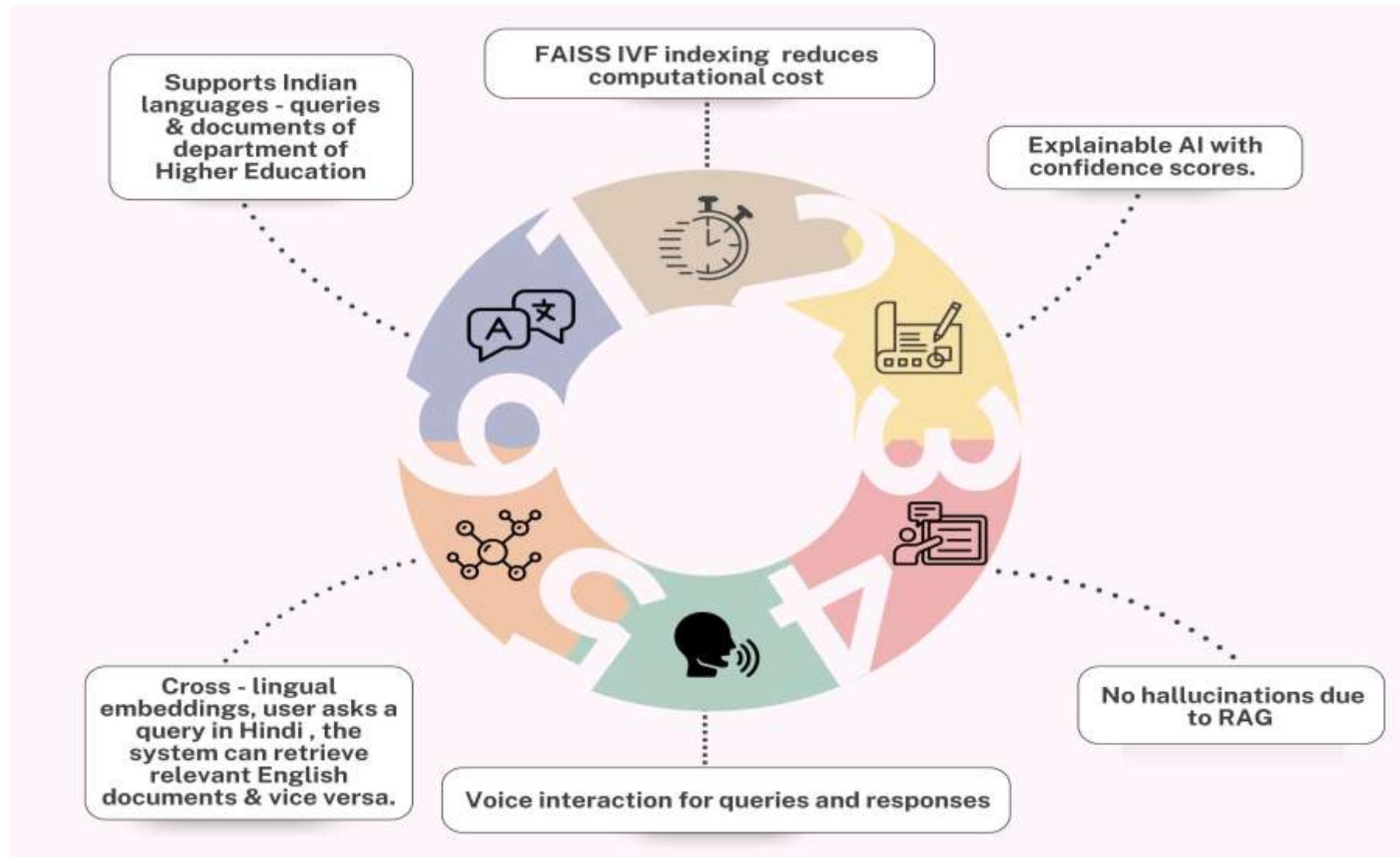
Strategies & Solution

Using pdf parsers and OCR for extracting text from these files.

Optimize query routing with RAG approach, cache frequently used embeddings & results.

Using containerized deployment via Docker, making it scalable for large scale production.

IMPACT AND BENEFITS



RESEARCH AND REFERENCES

- 1) <https://huggingface.co/sentence-transformers/LaBSE>
- 2) <https://huggingface.co/ai4bharat/Airavata?>
- 3) <https://python.langchain.com/docs/tutorials/rag/>
- 4) https://www.education.gov.in/en/higher_education
- 5) <https://towardsai.net/p/machine-learning/a-complete-guide-to-rag?>
- 6) https://huggingface.co/docs/transformers/model_doc/rag?
- 7) <https://github.com/stackitcloud/rag-template?>
- 8) <https://github.com/adithya-s-k/Indic-llm?>
- 9) <https://github.com/Zlash65/rag-bot-fastapi?>