Trent Limited FAQ Chatbot – Report

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Introduction

Trent Limited operates brands such as Westside, Zudio, Samoh, and Star. Stakeholders frequently seek concise, accurate answers about the company and its brands. This project delivers a Streamlit chatbot that uses only FAQs_questions.pdf as its knowledge base, ensuring responses remain within the document's scope.

Objectives

- Build a chatbot that replicates the existing app's flow and API usage but restricts context to FAQs_questions.pdf only.
- Provide transparent citations of the PDF snippets used for each answer.
- Maintain a simple, dependable retrieval pipeline for fast responses and easy deployment.

Dataset/Source Description

- Source: FAQs_questions.pdf (bundled with the app)
- Structure: Free-form text extracted per page using PyPDF2.PdfReader.
- Preprocessing: Minimal text cleaning; pages are treated as atomic context units.

Methodology

Retrieval-augmented generation (RAG) with:

- Vectorization: CountVectorizer(max features=5000, stop words='english')
- Weighting: TfidfTransformer(norm='12', use idf=True)
- Similarity: cosine similarity between query vector and page TF–IDF vectors
- Generation: OpenRouter Chat Completions (openai/gpt-3.5-turbo by default)

Rationale: TF–IDF is light-weight, explainable, and effective for short FAQs; it avoids external indexes and reduces operational complexity for coursework and small deployments.

System Design & Architecture

- Ingestion: PyPDF2 extracts text per page from FAQs_questions.pdf.
- Indexing: Pages are vectorized and transformed into a TF–IDF matrix (cached via st.cache_resource).
- Retrieval: Top-k pages selected by cosine similarity to the user query.
- Prompting: Retrieved snippets are concatenated into the user prompt; the system prompt instructs grounding in the PDF.

- Generation: The LLM produces an answer with the provided context.
- Transparency: The UI displays the snippets (page numbers and similarity scores) used.

Component summary:

- load faq artifacts(pdf path): PDF ingestion and TF-IDF matrix build
- retrieve_context(query, vectorizer, tfidf_matrix, pages, k): top-k retrieval
- build prompt(question, snippets): system and user prompts
- call openrouter(api key, model, system prompt, user prompt): API call

Implementation Details

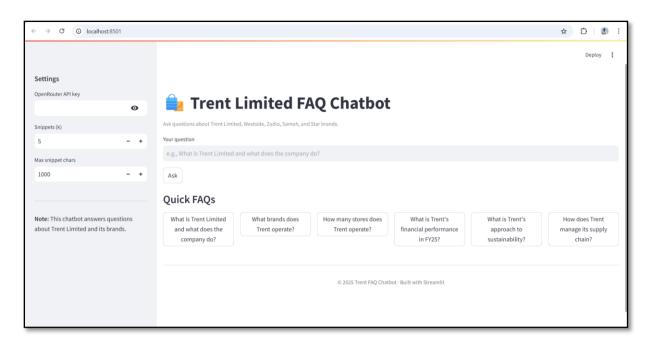
- Language & Frameworks: Python, Streamlit
- Dependencies: streamlit, scikit-learn, numpy, requests, PyPDF2
- Model API: OpenRouter (https://openrouter.ai/api/v1/chat/completions)
- Model: openai/gpt-3.5-turbo (configurable)
- Key Management: sidebar input, environment variable OPENROUTER_API_KEY, or Streamlit secrets
- Performance: k and snippet length tunable in sidebar; TF-IDF artifacts cached

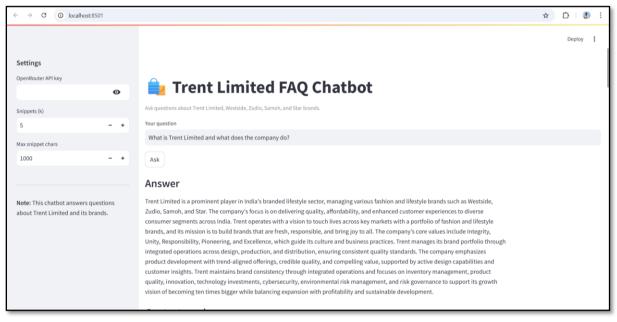
Key constraints enforced in code:

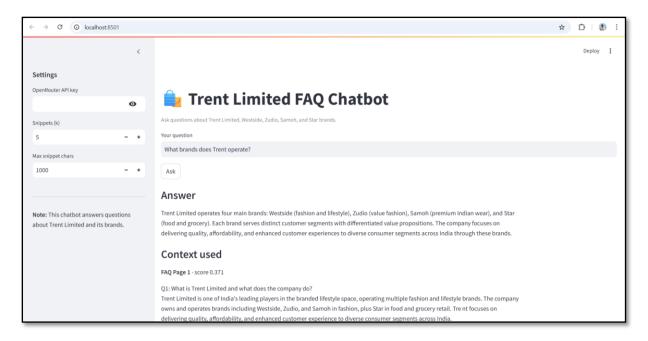
- Only FAQs questions.pdf is read for knowledge.
- If the answer is not in the PDF, the bot states this explicitly and suggests relevant sections.

Results & Observations

- The TF-IDF retriever reliably surfaced on-topic pages for most FAQ-style questions.
- Concise, specific queries improved retrieval precision compared to very broad prompts.
- The transparency panel (snippets + scores) helped verify that answers were grounded in the PDF.







Setup & Configuration

python -m venv .venv

.venv\Scripts\activate

pip install -r requirements.txt

API key options:

- Sidebar input (masked)
- Environment variable: OPENROUTER API KEY
- Streamlit secrets: create .streamlit/secrets.toml with:

OPENROUTER API KEY = "sk-or-v1-..."

How to Run

streamlit run trent_faq_pdf_chatbot.py

Open the provided local URL. If needed, paste your API key in the sidebar. Ask a question and review the "Context used" panel for the supporting PDF snippets.

Conclusion

This project delivers a focused, PDF-grounded chatbot for Trent/Zudio that mirrors the original app's API and UX while strictly constraining knowledge to FAQs_questions.pdf. The lightweight TF-IDF retriever provides dependable performance for FAQ-style questions and clear transparency of sources, satisfying the assignment's requirements.

References

- Streamlit documentation https://docs.streamlit.io
- scikit-learn: CountVectorizer, TfidfTransformer, cosine similarity https://scikit-learn.org
- OpenRouter API https://openrouter.ai

Conclus	on			
API and IDF retri	ect delivers a focused, PDF-gr UX while strictly constraining ever provides dependable perf s, satisfying the assignment's	knowledge to FAQ formance for FAQ-st	s_questions.pdf. The li	ghtweight TF-