

# Implementation of an Agentic RAG Framework for Comparative Analysis of Classical Islamic Corpora

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GitHub: <https://github.com/shafaeihossin-hub/Classical-Shia-AI-RAG>

**Abstract**—This research introduces a localized Retrieval-Augmented Generation (RAG) system specialized for theological discourse. Using an Agentic Query Expansion mechanism with DeepSeek-R1, we developed a system capable of cross-referencing the Holy Quran, Al-Kafi, Nahj al-Balaghah, and Musnad Ahmad ibn Hanbal. The architecture ensures data privacy via Docker and Qdrant.

## I. INTRODUCTION

Navigating classical theological texts requires high precision. Traditional LLMs often suffer from hallucination. This project implements a private RAG pipeline to ensure strict adherence to verified local contexts.

## II. SYSTEM ARCHITECTURE

### A. Infrastructure and Storage

The vector storage is handled by Qdrant, running within a Docker container to ensure environment isolation.

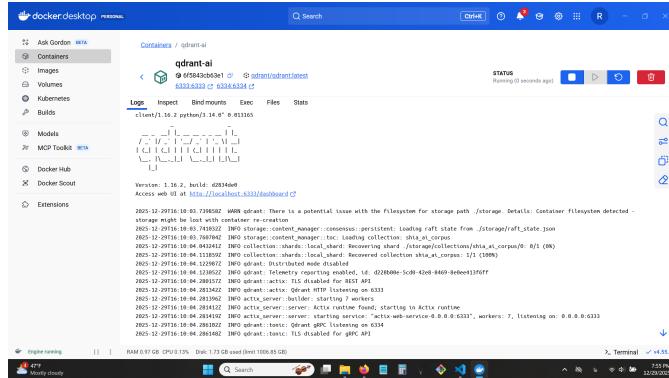


Fig. 1. Containerized Qdrant-AI environment in Docker.

### B. Agentic Search Logic

The system uses an "Agentic" approach, generating additional search keywords to improve retrieval hit rate before querying the database.

## III. IMPLEMENTATION DETAILS

Listing 1 demonstrates the core logic for query expansion and the RAG loop.

```
1 def get_smart_queries(user_question):
2     prompt = f"Generate 2 short search keywords in Persian related to: {user_question}."
3     res = ollama.generate(model=MODEL_NAME, prompt=prompt)
4     keywords = res['response'].strip().split(',')
5     return [user_question] + [k.strip() for k in keywords]
6
7 def shia_ai_rag_query(user_question: str):
8     queries = get_smart_queries(user_question)
9     combined_context = ""
10    for q in queries:
11        combined_context += query_database(q) + "\n---\n"
12    return generate_response(combined_context)
```

Listing 1. Agentic Query Expansion Logic

A screenshot of the Visual Studio Code (VS Code) interface. The top bar shows 'File', 'Edit', 'Selection', 'View', 'Go', 'Run', 'Terminal', 'Help', and a search bar. The main area has a code editor with Python code. The code defines two functions: 'get\_smart\_queries' and 'shia\_ai\_rag\_query'. The 'shia\_ai\_rag\_query' function calls 'get\_smart\_queries' and then iterates over the returned queries to call 'query\_database' for each, concatenating the results with '\n---\n' between them. The code also includes imports for 'ollama', 'httpx', 'requests', and 'OLLAMA\_API\_URL'. Below the code editor, there's a terminal window showing log output. The bottom right corner shows the date and time as '12/29/2023 10:51:40 AM'.

Fig. 2. Implementation of the RAG pipeline in VS Code.

## IV. PROJECT MANAGEMENT

The project is open-sourced under the MIT License and hosted on GitHub.

## V. CONCLUSION

This study validates that Small Language Models (SLMs) can achieve professional-grade performance in specialized fields when paired with intelligent retrieval mechanisms.

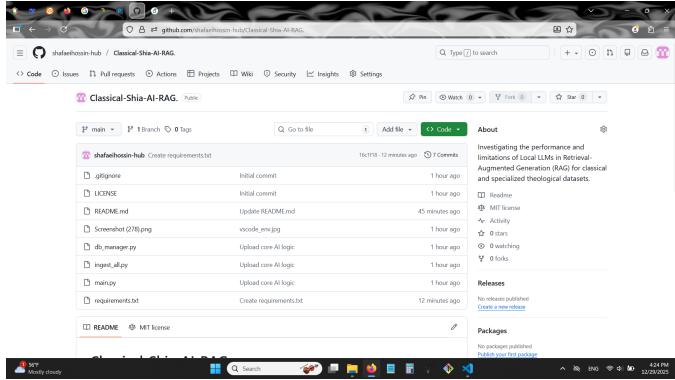


Fig. 3. Project structure and version control on GitHub.

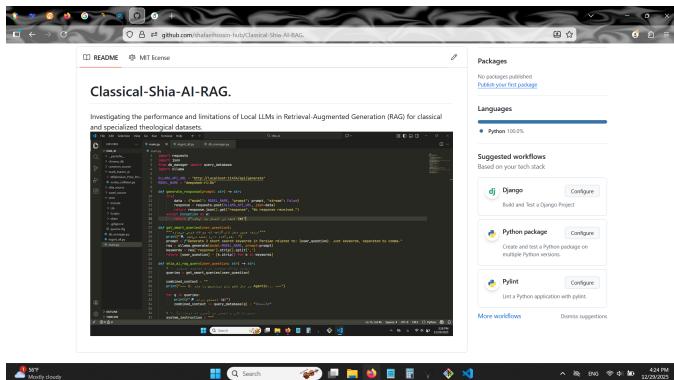


Fig. 4. Project Documentation and README details.