

Evvo Technology Solutions Pvt Ltd

Technical Challenge – LLM Engineers

Assumptions and Design considered to set up the H-RAG

Assumptions

1. **Document Availability:** Relevant, structured documents are available and can be broken into meaningful chunks.
2. **Clear Queries:** Users will ask clear, specific questions that can be answered using the document data or language model knowledge.
3. **Embedding Quality:** Generated embeddings accurately represent document content, enabling effective search in the vector store.
4. **System Scalability:** The system can handle large document volumes and high query frequency.
5. **Relevance of Data:** Documents in the system are pertinent to user queries.

Design Considerations

1. **Two-Layer Hierarchy:**
 - **Layer 1:** Stores summaries or keywords to quickly narrow down relevant sections.
 - **Layer 2:** Holds detailed document chunks for in-depth answers.
2. **Efficient Chunking:** Optimize chunk sizes to balance context preservation and search efficiency.
3. **Model Selection:** Use Mistral for embedding and rephrasing; fallback to web search if the answer isn't in the vector store.
4. **Performance Optimization:**
 - Use FAISS for fast embedding search.
 - Add a re-ranking mechanism for accuracy.
5. **Relevance Feedback:** Allow user feedback to improve system accuracy.
6. **Scalability:** Design for easy updates as documents grow, with minimal impact on system performance.
7. **Fallback to Internet Search:** If vector store search fails, use web search as a last resort.

Embedding used for the vector store

Embedding Used

- **Model:** Mistral
- **Type:** Dense vector embeddings
- **Source:** Generated using the Mistral model in Ollama.

Rationale for Selection

1. **Accuracy and Contextual Understanding:** Mistral embeddings effectively capture the semantic context, making them well-suited for retrieving relevant content based on user queries.
2. **Consistency in Representation:** Dense vectors offer a uniform structure, which is beneficial for calculating similarity metrics in FAISS-based vector stores.
3. **High Retrieval Performance:** Mistral embeddings provide reliable search accuracy for complex queries, ensuring high relevance of retrieved results.
4. **Scalability:** Dense embeddings maintain consistent performance even as the dataset grows, supporting scalability in a multi-layered system.
5. **Flexibility for Hierarchical RAG:** The embeddings work well in a two-layer system, supporting both keyword-based and detailed retrieval as needed.

Test questions used to evaluate the performance of the H-RAG system

To evaluate the effectiveness and relevance of the H-RAG system, the following test questions were

used:

1. Basic Definition Queries:

- Example: "What is the minimum effective tax rate?"
- Purpose: Assess the system's ability to retrieve direct answers from the vector store.

2. Contextual Information Queries:

- Example: "Explain BEPS in the context of Singapore."
- Purpose: Evaluate the system's understanding of nuanced questions and its ability to provide context-specific information.

3. Comparative and Analytical Queries:

- Example: "How does the Singapore tax policy compare to other regions under BEPS guidelines?"
- Purpose: Test the H-RAG's capability to retrieve and combine multiple relevant chunks to form a detailed response.

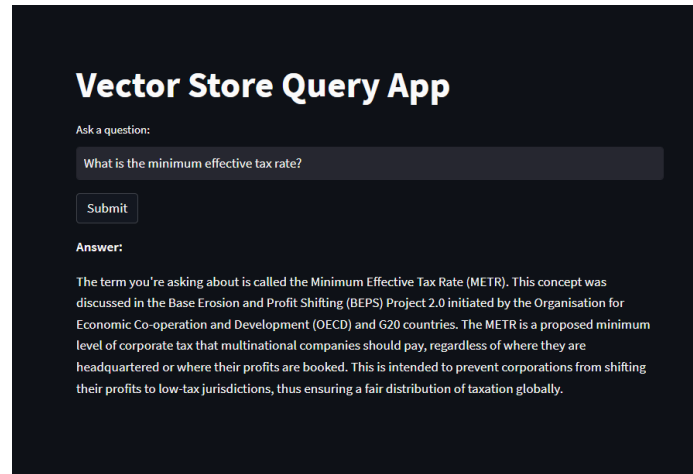
4. Keyword and Summary-Based Queries:

- Example: "List the key provisions under Pillar Two of BEPS."
- Purpose: Assess the effectiveness of the keyword-based layer in the hierarchical structure.

5. Complex, Multi-Part Queries:

- Example: "How does the H-RAG system enhance tax transparency and prevent base erosion?"
- Purpose: Test the system's ability to understand and respond accurately to multi-faceted queries.

For example image:



Agent Design and Test Questions for Evaluation

Agent Design: The decision-making agent is structured to intelligently choose the best method for responding to user queries based on query type and relevance:

1. RAG System Response:

- The agent first queries the vector store to retrieve relevant information based on the embeddings from the hierarchical RAG system.
- If relevant content is found, it formulates a comprehensive answer using retrieved chunks.

2. Direct Answer from Knowledge Model (Mistral):

- If the query's relevance to stored content is low, the agent uses a language model (Mistral) to generate a response based on its built-in knowledge.

3. Internet Search for Unavailable Information:

- For queries outside the scope of the vector store and language model, the agent performs an internet search to retrieve up-to-date and detailed information.