

## Task Title:

Develop CPU-Based RAG LLM Inference Pipeline

## Objective:

Design and implement a Retrieval-Augmented Generation (RAG) pipeline optimized for inference on CPU-only environments. The solution should enable efficient document retrieval and response generation using a lightweight LLM and embedding model.

## Responsibilities:

### 1. Design the Architecture:

- Define a modular pipeline integrating document retrieval and generative response components.
- Ensure all components are optimized for CPU usage with minimal memory and latency footprint.

### 2. Embed and Store Documents:

- Use a CPU-compatible embedding model (e.g., `sentence-transformers`, `MiniLM`, `llamafile` etc.).
- Store embeddings in a vector database (e.g., FAISS with CPU backend, ChromaDB).

### 3. Implement Retrieval:

- Perform efficient similarity search using CPU-based vector index.
- Implement ranking logic if necessary to refine the top-k results.

### 4. Integrate LLM Inference:

- Use a lightweight CPU-optimized LLM (`llama.cpp`, `ggml`, `llamafile`, `vllm`).
- Implement prompt construction by combining query and retrieved documents.

### 5. Pipeline Integration:

- Build an inference API or script that accepts a user query, retrieves relevant documents, and returns a generated response.

- Ensure the pipeline runs within acceptable latency bounds on CPU.

#### **6. Testing and Benchmarking:**

- Evaluate response accuracy, retrieval quality, and latency.
- Compare performance across different CPU models if needed.

#### **7. Documentation:**

- Provide clear usage instructions and installation requirements.
- Document the pipeline structure, model choices, and optimization techniques.

### **Deliverables:**

- End-to-end CPU-based RAG inference pipeline (codebase).
- Sample script/API or streamlit for interactive use.
- Deployment guide and benchmark report.
- Documentation for extending or modifying the pipeline.