## **Objective**

Assess the candidate's experience in:

- Building LLM-based agentic bots (multi-agent pipeline)
- Implementing a RAG (Retrieval-Augmented Generation) flow
- Modular architecture and clean code
- Integration with external systems (webhooks/API)
- Prompt design—even if stubbed—and testing/documentation

## "Product-Query Bot via RAG Pipeline"

#### **Scenario**

Zubale needs a microservice that receives user questions about products (simulated WhatsApp messages), retrieves relevant information from a document corpus, and responds via a webhook callback.

### Requirements (Time-boxed, max 3 hours)

- 1. Incoming Endpoint
  - POST /query accepting JSON

```
JSON
{ "user_id": "string", "query": "string" }
```

Validate input and enqueue the request (e.g., using Python asyncio.Queue).

### 2. RAG Pipeline

- Index ~5–10 product description documents into an embedded vector store (in-memory or file-based).
- On receiving a query:
  - Retrieve top-k documents semantically similar to the query.
  - Send a prompt to a stubbed LLM function (e.g., ai\_generate(context, query)) that returns a generated answer grounded in retrieved context.

#### 3. Multi-Agent Structure

- o Implement at least two agents:
  - Retriever Agent: handles semantic retrieval.
  - Responder Agent: compiles generation logic.

• Either use a framework like <u>LangGraph</u> or a clean custom orchestration to emphasize modularity and "multi-agent" separation .

### 4. Callback Integration

 After generation, send the answer via POST to a configurable CALLBACK\_URL (log output is acceptable if external endpoint isn't available).

### 5. Code Quality

- Include unit tests (e.g., pytest or jest), at least for the retrieval logic or prompt flow
- Use environment variables for configuration (e.g., CALLBACK\_URL, top-k).
- o Provide clear documentation: setup, run instructions, time spent.
- o Containerization with Docker is a plus but not required.

## **Evaluation Criteria**

| Dimension               | What We Assess  |
|-------------------------|---|
| RAG Pipeline            | Understanding of retrieval + generation, avoiding hallucination |
| Agent Design            | Clean separation of concerns or use of agent frameworks         |
| Integration             | Webhook/API flow simulating real product use                    |
| Prompt & LLM Mock       | Evidence of thoughtful prompt structure                         |
| Testing & Documentation | Clarity, readability, minimal automated coverage                |
| Engineering Practices   | Modularity, configuration, Docker usage, code style             |
|                         |   |

# **Optional Enhancements (if time permits)**

- Real API call to an LLM (e.g., OpenAI, Claude).
- A memory or context agent to track conversation history.
- Use of Docker Compose to wire up the service.
- A frontend script to simulate POST requests.
- Basic evaluation with automated unit scoring or vetting.

## **Deliverables**

- 1. Git repo with:
  - a. Source code and test suite
  - b. README with: instructions to run/index, how to test flow, time taken (indicative of 1–2 h)
- 2. (Optional) Dockerfile and/or docker-compose.yml
- 3. Example POST commands or scripts demonstrating the exercise
- 4. Video sharing the screen and explaining the outputs and a demo, ideally with a duration of less than 6 minutes