

MedicalGPT – What I Built & Why (Personal Notes)

1. Goal (Why I started this project)

- I already knew ML/DL theory but lacked **hands-on LLM + RAG experience**
 - Interviewers ask “*Have you worked with RAG, embeddings, vector DBs?*”
 - Instead of learning theory first, I decided to **build first, understand while building**
 - Goal was **learning + resume**, not production product
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2. First step – LLM API connectivity

What I did

- Created a simple `main.py`
- Called an LLM using LangChain + OpenRouter
- Asked a basic question like “*Explain what anatomy is*”

Why I did it

- To confirm:
 - API key works
 - Provider works
 - LangChain → LLM pipeline works
- If this fails, nothing else matters

Interview takeaway

- Always isolate **LLM connectivity** before adding complexity
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3. PDF loading (Document ingestion)

What I did

- Used a real medical textbook PDF (Anatomy & Physiology)
- Loaded it using `PyPDFLoader`

Why I did it

- RAG always starts with **external knowledge**
- LLMs don’t magically know your documents
- PDFs need to be converted into text before use

Interview takeaway

- $RAG = LLM + \text{external documents}$
 - First step is **document ingestion**
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4. Chunking (Very important concept)

What I did

- Split large PDF pages into smaller chunks
- Used:
 - `chunk_size = 500`
 - `chunk_overlap = 100`
- 500+ pages \rightarrow ~2000–4000 chunks

Why chunking is required

- LLMs have **context length limits**
- You cannot pass an entire book to the model
- Smaller chunks:
 - Fit into context
 - Improve relevance
 - Reduce noise

Why chunk overlap is used

- Prevents cutting meaning in half
- Keeps semantic continuity between chunks

Engineering tradeoff

- Small chunks \rightarrow better precision, more chunks
- Large chunks \rightarrow fewer chunks, more noise
- 500/100 is a **balanced industry-standard choice**

Interview takeaway

- Chunking is like **batching text semantically**
 - Chunk size directly affects retrieval quality
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5. Embeddings (Key interview topic)

What I learned

- **Embeddings \neq LLM**
- LLMs generate text
- Embeddings convert text into **numerical vectors**

Why embeddings are needed

- Computers don't understand meaning directly
- Embeddings allow:
 - similarity search
 - semantic matching
- Similar meaning \rightarrow vectors closer together

How the system understands similarity

- Question \rightarrow embedding
- Chunks \rightarrow embeddings
- Compare vectors using distance (cosine / dot product)

Interview takeaway

- Embeddings power **retrieval**
 - LLMs power **generation**
 - They serve different purposes
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6. Vector database (Conceptual understanding)

What I attempted

- Used ChromaDB to store chunk embeddings
- Persisted embeddings locally

What happened

- Faced tooling issues on Windows + free API tier
- Retrieval calls blocked or hung

Engineering decision

- Stopped fighting the tool
- Switched to **manual RAG** for learning

Interview takeaway

- Tools change, **architecture matters**
- Engineers adapt instead of forcing unstable setups

7. Manual RAG (Key learning moment)

What I did

- Selected a **small subset of chunks**
- Passed them directly as context to the LLM
- Asked questions based on that context

Why this still counts as RAG

- Retrieval doesn't have to be vector-based
- Core idea:
 - External context
 - Grounded answers
- Vector DB is an optimization, not a requirement

Interview takeaway

- RAG = retrieve → augment → generate
 - Retrieval can be manual or automated
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8. Prompt design & safety

What I enforced

- “Answer strictly from provided context”
- “If not present, say I don't know”
- No diagnosis
- No prescriptions

Why this matters

- Prevents hallucinations
- Important for medical domain
- Shows safety awareness

Interview takeaway

- Prompting is **control logic**, not magic
 - Constraints matter more than fancy prompts
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9. Interactive chat loop (Final product)

What I built

- Terminal-based MedicalGPT chatbot
- User can ask multiple questions
- Exit using `exit` or `quit`

Why terminal chat

- No UI complexity
- Focus on core logic
- Easy to debug
- Interviewers like simple demos

Interview takeaway

- UI is optional
 - Core system design is what matters
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10. Key engineering lessons (Very important)

- Always validate components **step by step**
 - Separate:
 - API testing
 - PDF loading
 - chunking
 - retrieval
 - generation
 - Don't fight tools endlessly
 - Make reasonable tradeoffs
 - Build first, optimize later
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11. How I explain this project in interviews (short)

“I built a MedicalGPT chatbot using LangChain and OpenRouter.

I ingest medical PDFs, chunk them, construct context dynamically, and answer user questions grounded strictly in the document.

I explored embeddings and vector databases, understood their role in semantic retrieval, and adapted my design based on tooling constraints.”