Doctor Appointment System using LangGraph, FastAPI, and Streamlit

This project is a multi-agent, AI-powered Doctor Appointment
Booking System that simulates a realistic medical assistant
capable of handling user queries regarding doctor availability,
specialization, and appointment booking.

Tech Stack:

LangGraph for workflow automation between agents

LangChain for Model Loading, Prompt Creation and tool usage

FastAPI for serving the API endpoint /execute

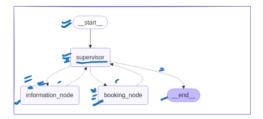
Streamlit for frontend interaction

Python + Pandas + CSV for data handling

Sample Use Case:

User: Can you check if a dermatologist is available on 5th August at 10:00

Agent Response: Dr. Smith (Dermatologist) is available at 10:00 AM on 05-08-2024. Do you want to book the appointment?



product

professional grade system design

tech stack data flow, deployment



a dog is warning sunglasses on beach

DOG

Features

Image upload

Natural language query input

Show results with similarity scores

Optional filters (category, color, etc.)

Endpoints:

/upload_image: Accept image, generate & store embedding /search: Accept text query, return top-k similar images

/images: Serve stored images (from S3 or local)

Embedding Layer (Multimodal Models)

 $Image\ Encoder:\ CLIP\ /\ BLIP\ /\ GIT\ (OpenAl\ or\ HF\ Transformers)$

Text Encoder: Same model (CLIP text encoder) or sentence-transformer

Tech: transformers (Hugging Face) sentence-transformers

open-clip-torch or Salesforce/BLIP

Vector Database

Stores image embeddings

Supports similarity search (text \rightarrow image)

Options: FAISS (local, fast) Qdrant/Weaviate (cloud/scale ready)

Embedding stored with: image_path meta (timestamp, uploader, etc.)

Image Storage

Local: ./data/images/

Or S3: s3://bucket-name/image-name.jpg

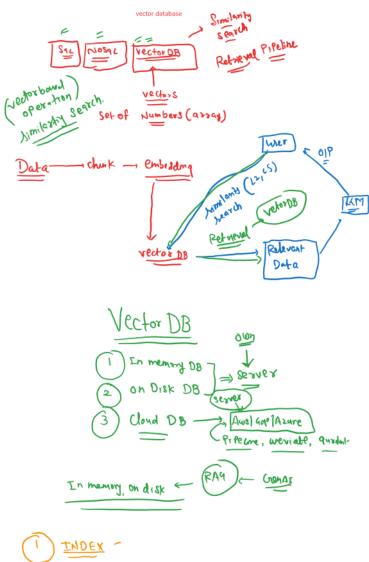
Serve via signed URL or FastAPI static

LLM (Optional Features)

Use-case: Query rewriting, semantic expansion "beach dog → a dog playing on a sunny beach"

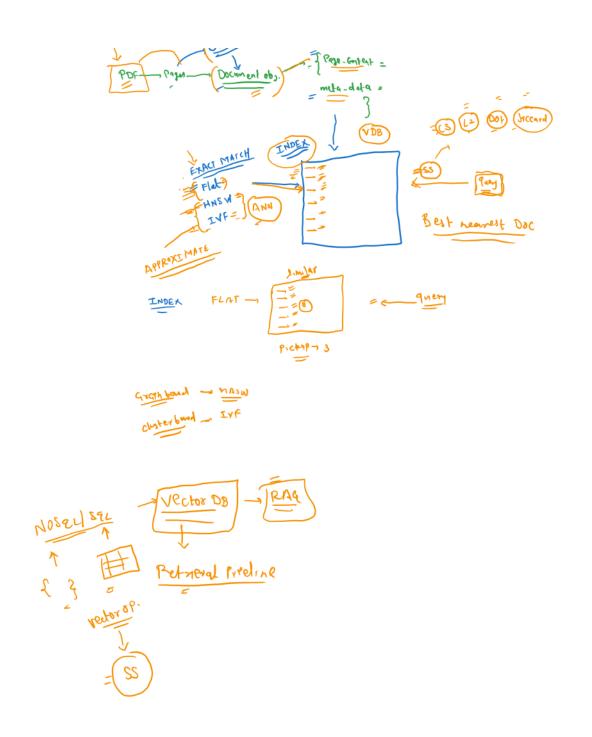
Model: GPT-3.5 (OpenAI) or phi-2 via Hugging Face

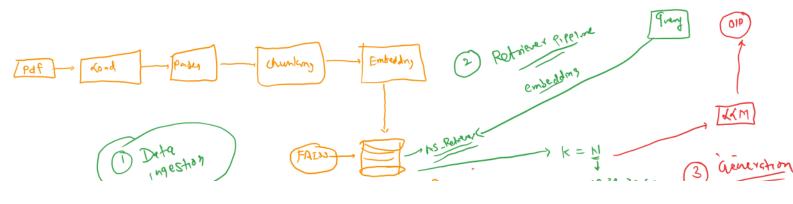
| Component | Tool/Platform |





Similarly Search =









we use VDB for Creating Retrievar Pipeline



Туре	Description	Example DBs
Туре	Description	Example Databases
Local	Runs on your own laptop or server, either on the file system or via Docker	FAISS, Qdrant (local), Weaviate
In-Memory	Stores data only in RAM — data is lost when the system restarts	FAISS, Qdrant (local), Weaviate
Cloud Hosted	Fully hosted by vendors — supports persistence, scaling, APIs, monitoring, etc.	Pinecone, Qdrant Cloud, Weaviate Cloud, Milvus (Zilliz Cloud)

Local: useful for experiments, dev, prototypes, or small deployments

Cloud: needed for prod systems with

- Scalability
- High availability
- Multi-user/team access
- · Real-time updates
- Monitoring & backups

Feature	In-Memory	Local Persistent	Cloud Hosted
Data survives restart	No	Yes	Yes
Scales across nodes	No	Limited	Yes
Team access (multi- user)	No	Manual	Yes
Easy integration (LangChain etc.)	SDKs	SDKs	SDKs/APIs ready
Use-case fit	Small, fast-only	Dev/PoC	Production scale

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Vendor	Local Support	Cloud Hosted
FAISS	Yes	No
Quadrant	Yes	Yes
Weaviate	Yes	Yes
Pinecone	No (API only)	Only
Milvus	Yes	(Zilliz)
Chroma	Yes	No

https://python.langchain.com/docs/integrations/vectorstores/ https://superlinked.com/vector-db-comparison

Term	Full Form	Meaning in Simple Terms
KNN	K Nearest Neighbors	Exact search : Return top K truly closest points
ANN	Approximate Nearest Neighbors	Faster search : Return top K almost closest points (approximate)

KNN (Exact)	ANN (Approximate)
Compares query with	Compares with a
all vectors	smart subset
Slow but 100% accurate	Fast but ~95-99% accurate
Works fine for small datasets	Best for large-scale datasets (millions)
Examples: brute-force FAISS flat	Examples: HNSW, IVF, PQ, ScaNN, etc.

Modalities = Different types of input data

Text → plain sentences, paragraphs

Image → photos, diagrams, charts

Audio → voice, speech, music

Video → moving visuals with audio

Documents \rightarrow combination of text + images

Use-case	Modalities Involved
ChatGPT Vision	Image + Text
YouTube caption search	Video + Audio + Text
Resume screening from PDF + profile	Text + Table + Image (logos etc.)
Search from screenshots	Image → OCR → Text → Semantic Search
Voice-commanded document search	Audio → Text + Query understanding
Medical Report Q&A	Document (PDF) = Text + Charts + Tables

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Text → BERT / GPT / MiniLM → vector
Image → CLIP / BLIP → vector
Audio → Whisper / Wav2Vec → vector
Video → Sampled frames + CLIP → vector
Table → LayoutLM / TAPAS → vector
The difference is not in creating the vector; the difference lies in the features and ecosystem of
the vector database.
Pinecone
 {"id": "doc-1", "values": [0.22, 0.88, ...]}
                                                                                   G
Weaviate
                                                                                   G
   "class": "Product",
   "properties": {
     "image": "path.jpg",
     "text": "Best shirt in market"
   }
 }
```

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Vector DB	Verdict	Use-case Fit
Milvus (Zilliz)	good for multimodal	Image + Text + Video + Audio, production scale
<u>Qdrant</u>	Excellent, very flexible	Text + Image + Metadata hybrid search
Weaviate	Good with Transformers + hybrid	RAG, text + image + metadata combo
Pinecone	Text-only focus (no native image)	Great for text-based RAG, not for images
Vespa.ai	Enterprise-heavy multimodal <u>infra</u>	Multimodal + advanced query logic
AWS OpenSearch (KNN)		
Azure AI Search		
GCP Vertex AI Vector Search		