



GENAI RAG PROJECT: TOPIC STUDY ASSISTANT [POC]

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# GENE EXPRESSION ACTIVATION VIA MICRO RNA

# USER-FOCUS AND TOPIC CHOICE

► Developer's goal:

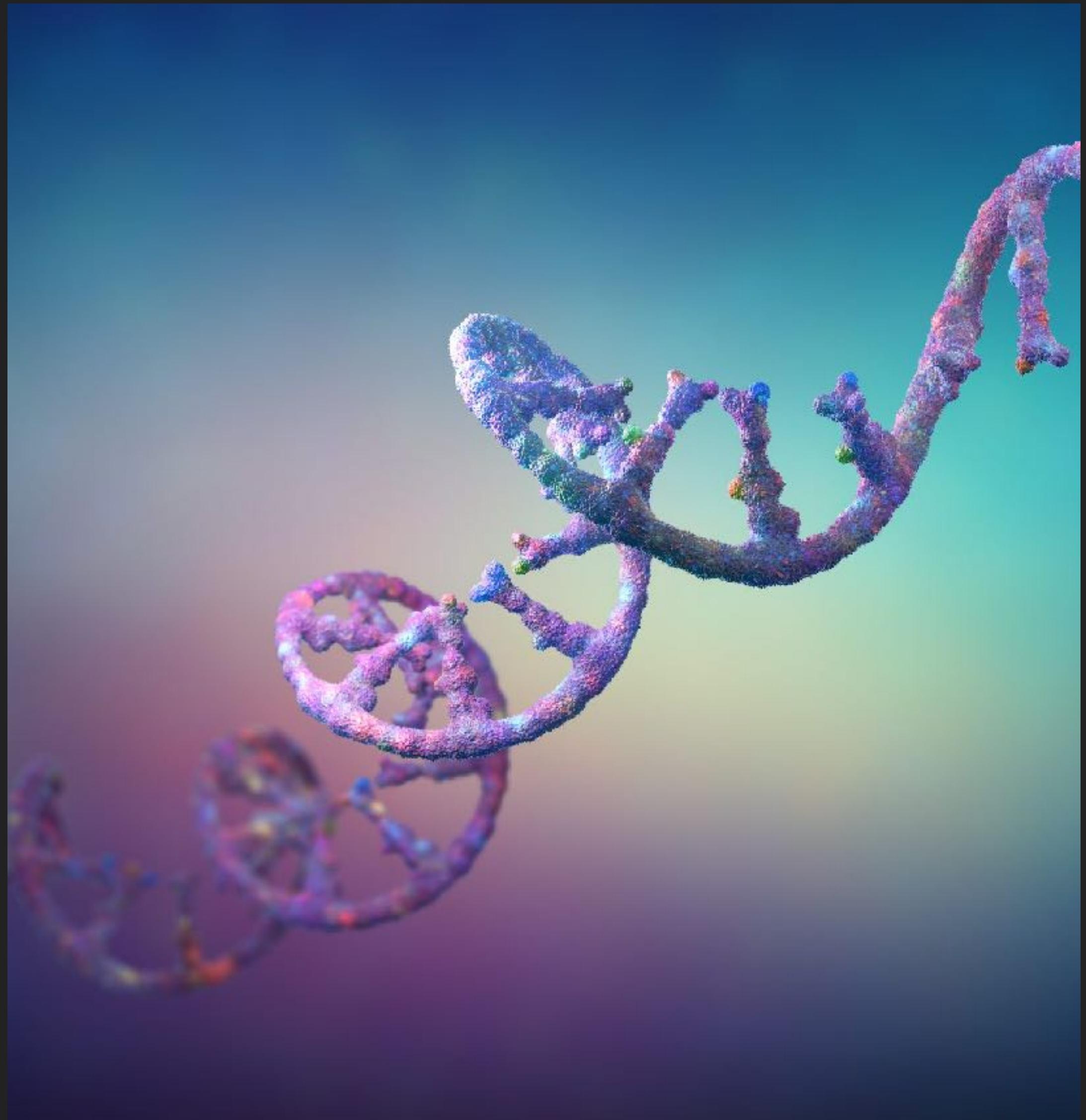
- ▶ Learn and practice implementing cutting-edge technologies.
- ▶ Create a **pet-project** for portfolio and upscale.
- ▶ Practice user-oriented project design.

► User's goal:

- ▶ Get assistance for writing **Master's thesis** in Biology on the topic:

## "GENE EXPRESSION ACTIVATION VIA MICRORNA".

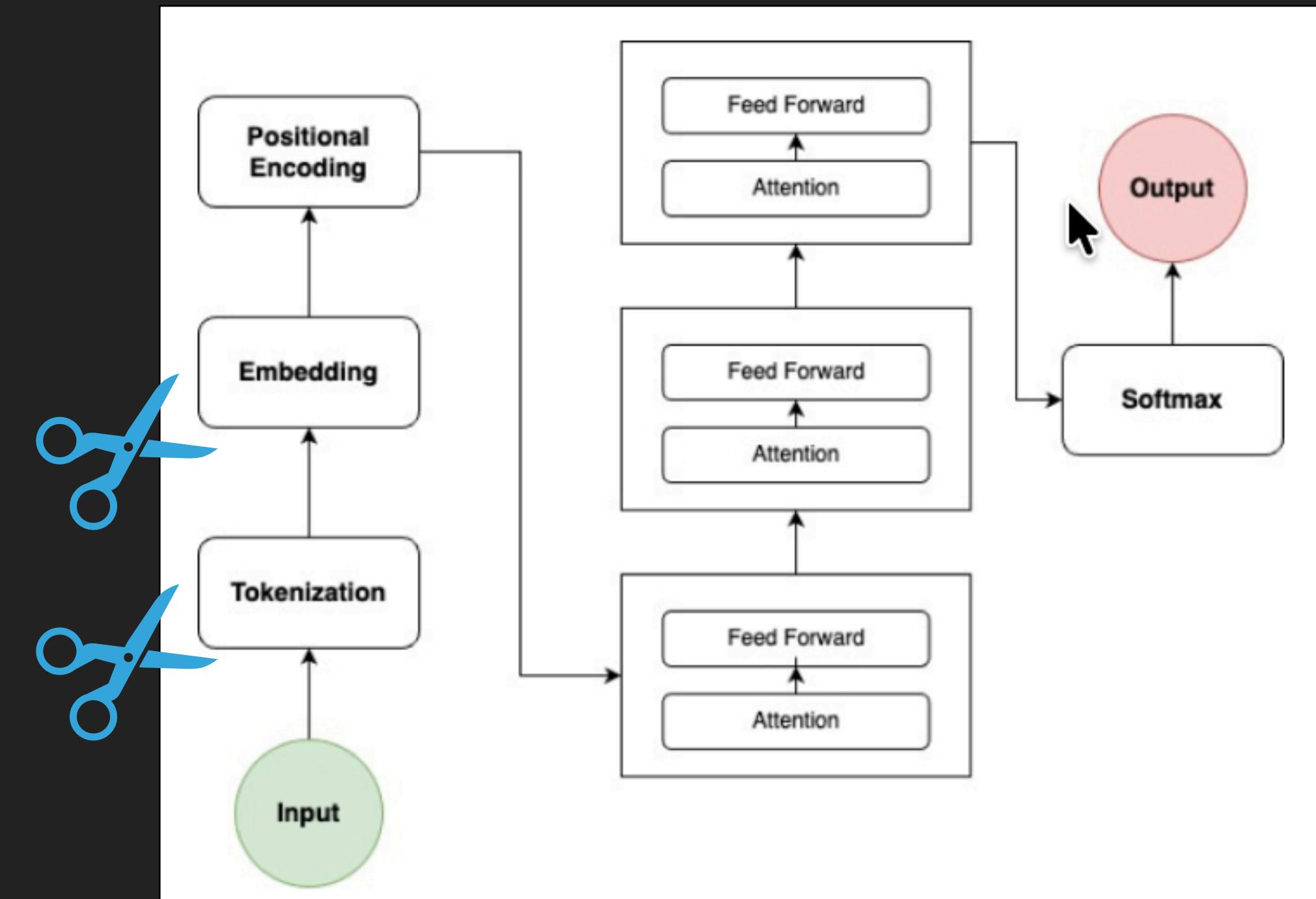
- ▶ Deepen knowledge in the research topic.
- ▶ Start collaborating with data scientists for further steps in Bioinformatics.



# TERM DISAMBIGUATION

- ▶ **Generative AI** - AI that generates new data by learning patterns from existing data.
- ▶ **GPT** - Generative Pre-trained Transformer
- ▶ **LLM** - Large Language Model - generates output.
- ▶ **RAG** - Retrieval Augmented Generation
- ▶ **Embedding model** - converts data chunks into a numeric vector.

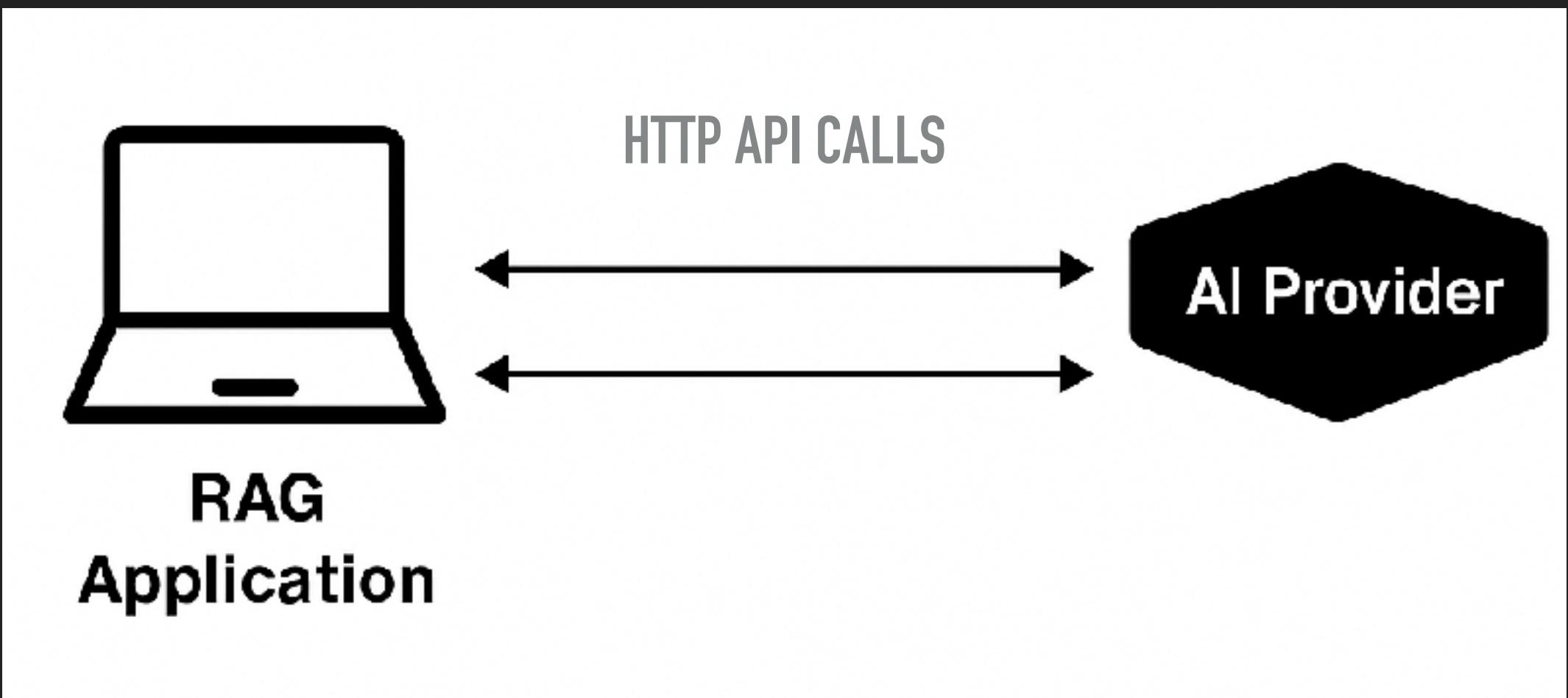
## TRANSFORMER / GPT WORKFLOW



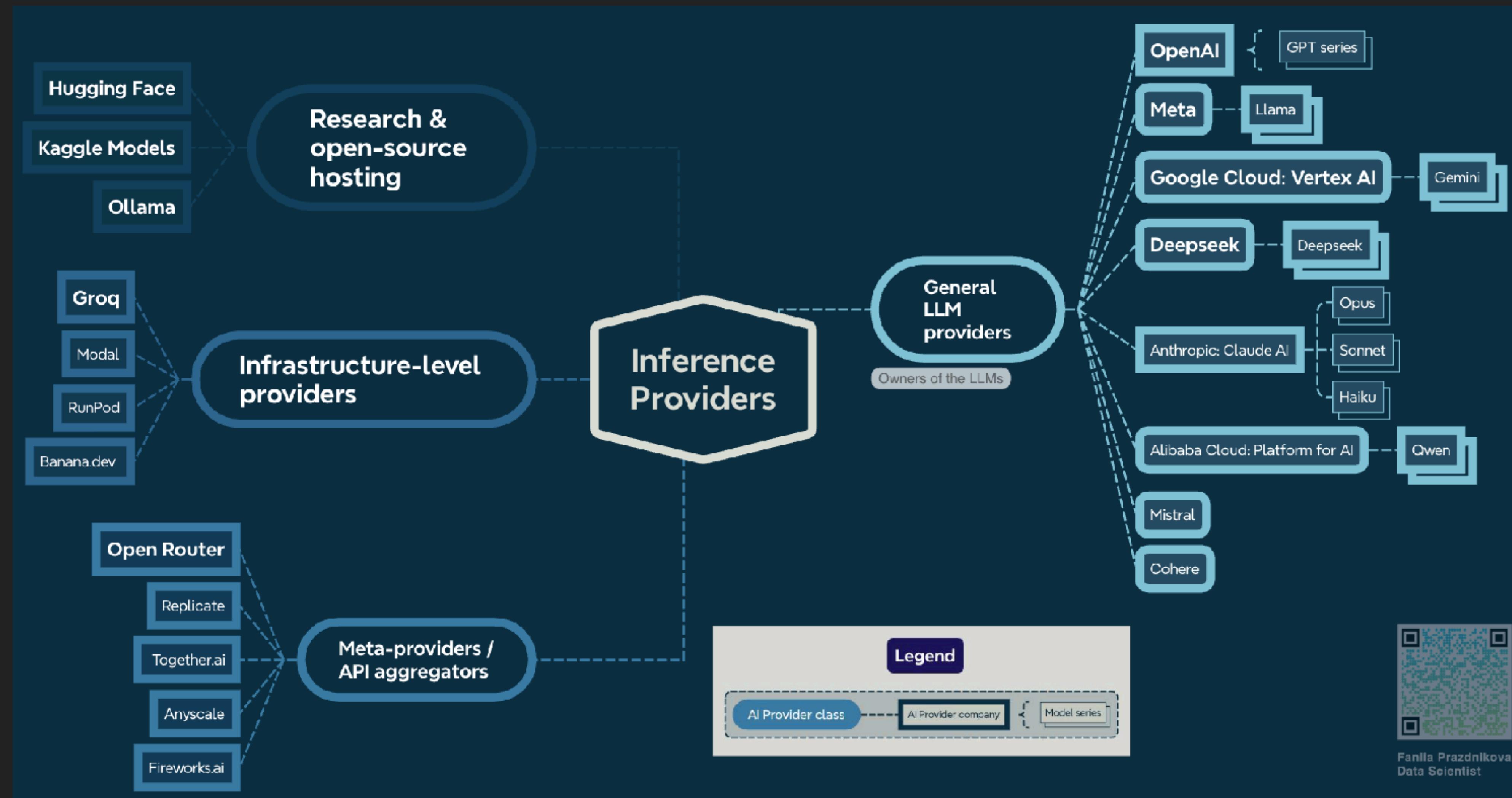
# WHAT ARE INFERENCE PROVIDERS

- ▶ Platforms that host, run, and expose AI models via APIs.
- ▶ Handle the **inference** - the actual computation that generates answers, embeddings, images, etc.
- ▶ Provide:
  - ▶ model access (LLMs, embedding models, etc.)
  - ▶ APIs / SDKs for querying these models
  - ▶ Compute infrastructure (GPU clusters for inference)
  - ▶ Model hosting for custom deployments
  - ▶ Monitoring, billing, and authentication

## SYSTEM ARCHITECTURE



## CHOOSING INFERENCE PROVIDER



# RAG PIPELINE

## 1 Data collection

- ▶ Data types: PDF documents with text, images, formulas
- ▶ POC: data is collected manually by the user
- ▶ MVP: web-scraping or GPT-5 for finding relevant scientific articles

## 2 Data retrieval and tokenization

- ▶ Extract data from PDF files
- ▶ Preserve text, images, scientific formulas
- ▶ Tokenization: split into chunks

## 3 Embedding and indexing

- ▶ Embedding model: convert data to vectors
- ▶ Vector DB: store the vectors in DB and build semantic index

## 4 Data retrieval for answer

- ▶ Using GUI for getting query from a user
- ▶ Query tokenize/encode/embedding using the same model and methods
- ▶ MVP: top-k similarity search, reranking

## 5 Generating answer

- ▶ Prompt LLM with the retrieved context
- ▶ Return the answer to the user in GUI

# CHOOSING AI FRAMEWORK

## LangChain

tool- and workflow-centric

for various LLM projects

more diverse functional modules for architecture flexibility and scalability

open-source and free



## Llamaindex

data-centric

specific for RAG projects

more efficient in document indexing, data retrieval, semantic search

**commercial**, price depends on usage

migration to it can be considered after MVP and efficiency evaluation + donations / monetization prospects

# RAG TOOLCHAIN FOR POC PHASE

User Interaction: Type: Web site

- Back-End: none for POC
- Front-End: StreamLit with caching on localhost

AI framework: LangChain

PDF extractor: PyPDFLoader

Vector DB: Chroma DB on localhost

Chunker: RecursiveCharacterTextSplitter

Retriever: Chroma.as\_retriever

Embedding model: text-embedding-3-small

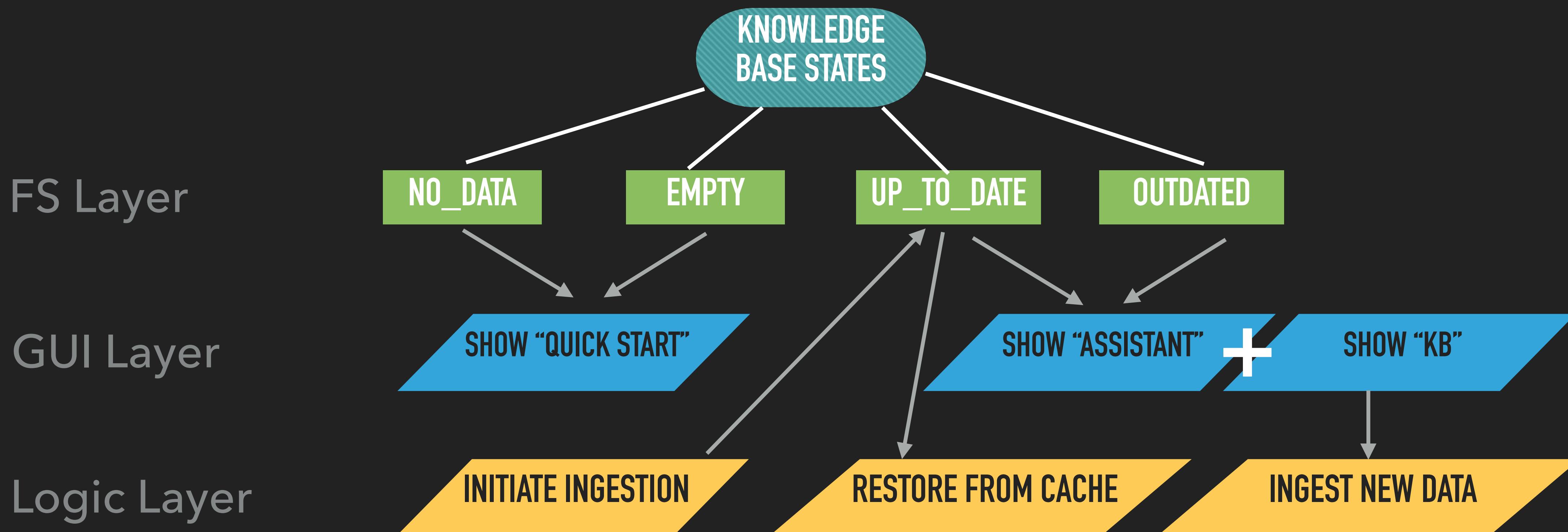
LLM: gpt-5-mini (primary), gpt-5-nano (for drafts/tests)

Inference Providers: 

- Embedding model: OpenAI

- LLM: OpenAI

## STATE MACHINE





**THANK YOU FOR  
ATTENTION**

**Faniia Prazdnikova**