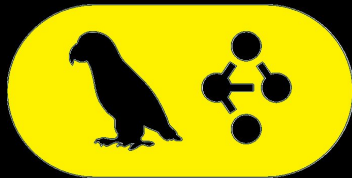


# Code Generation-Explanation Agent using LangGraph



# The Flow of the App:

1. Get user query.
2. Classify the intent of the user whether: **explain code**, **generate code**, or **other**.
3. If the user needed to **generate code**, then we use the **retriever** to get similar code snippets from our **vector database**, in which there are 164 code examples from the human eval dataset. Then we use the top 3 relevant code snippets as a context in the prompt to generate new code.
4. If the user needed to **explain code**, we check whether the user typed a python function or class to explain, otherwise we explain the last generated code.
5. If the user query was neither a code generation or explanation we handle it with a different node in the graph.

## The openai/openai\_humaneval dataset:

The HumanEval dataset released by OpenAI includes 164 programming problems with a function signature, docstring, body, and several unit tests. They were handwritten to ensure not to be included in the training set of code generation models.

Link: [https://huggingface.co/datasets/openai/openai\\_humaneval](https://huggingface.co/datasets/openai/openai_humaneval)

# The Vector Database, and the Embeddings:

I used **FAISS** (Facebook AI Similarity Search) Database in this project.

And for the embedding model, I used: **BAAI/bge-code-v1**, which is specialized in creating vector embeddings for code samples and retrieving them.

## The LLM Backbone:

The main LLM that I used in comprehending the retrieved codes and creating new ones, and explaining, and the general replies is the same LLM which was: `mistralai/mistral-7b-instruct:free` from OpenRouter.

The reason for choosing this model is that it worked well for my task (Compared to Llama) and it is free and available most of the time.

# The AgentState:

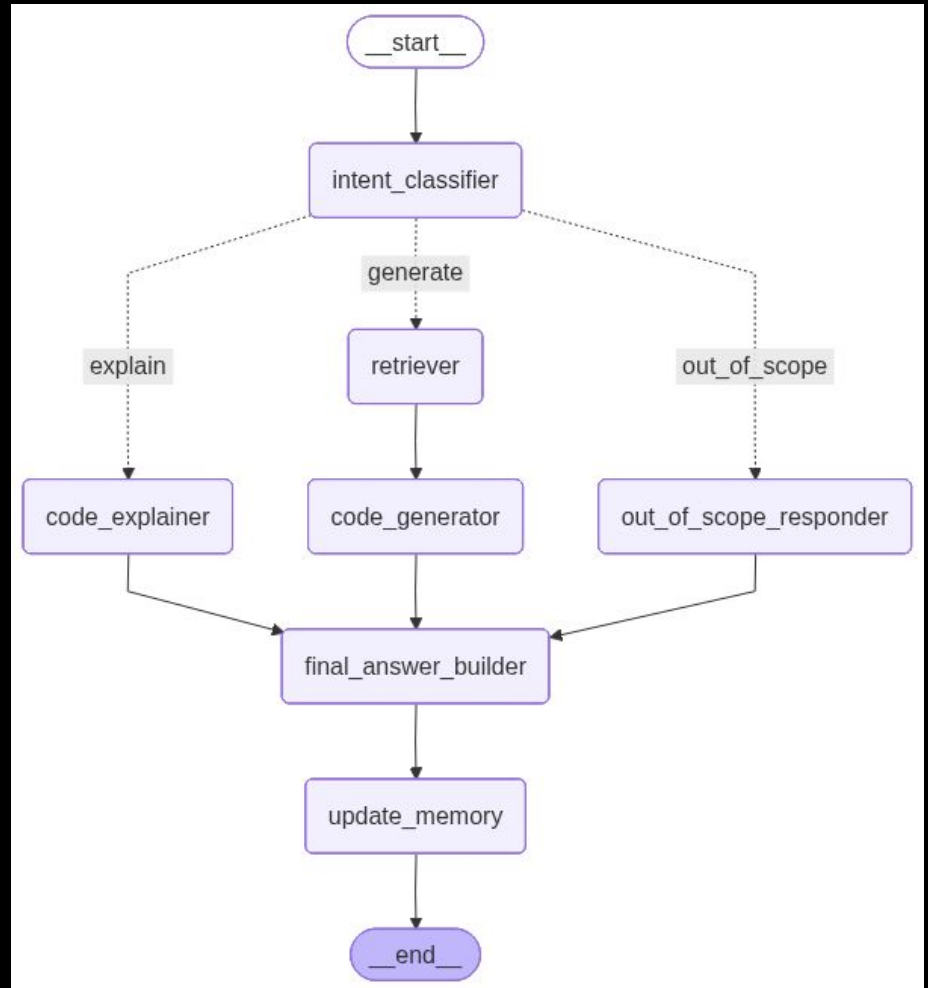
The Langgraph requires a state to flow between the nodes of the graph. I chose to implement the state to include the following data:

```
class AgentState(TypedDict):  
    user_input: str  
    intent: Optional[str]  
    retrieved_docs: Optional[List[Document]]  
    generated_code: Optional[str]  
    explanation: Optional[str]  
    final_answer: Optional[str]  
    memory: List[Dict[str, str]]
```

## The Graph:

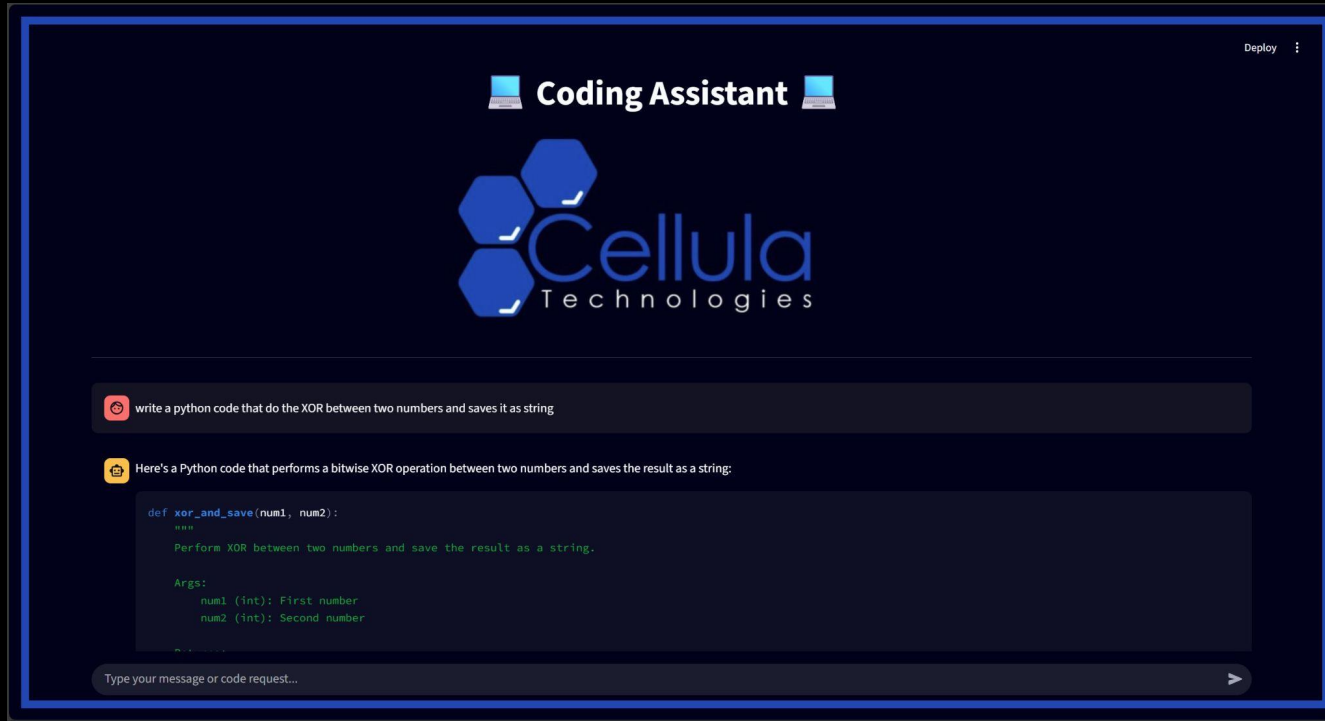
I created 7 nodes in my graph and made each one specialized in one task.

The intent\_classifier has two versions, one with simple hard coded rules and another one that uses LLM to classify.



# The Interface:

I made the interface using the Streamlit Library.





THANK YOU.