LangChain Pipeline with OpenAl

Objective

We are building a **basic LangChain pipeline** using OpenAl's LLM (gpt-40) to:

- · Send prompts
- Process outputs
- Optionally use tracing with LangSmith
- (Optionally) serve the app via LangServe

Load Environment Variables

What?

Read API keys and config values from a .env file.

Why?

This keeps sensitive keys out of your code.

1. You write your API keys in a .env file

You create this ... file in the same folder as your Python script.

This file looks like:

OPENAI_API_KEY=your-real-openai-key LANGCHAIN_API_KEY=your-langchain-key LANGCHAIN_PROJECT=my-cool-project-name

import os
from dotenv import load_dotenv
load_dotenv()

```
# Set OpenAI and LangChain environment variables
os.environ['OPENAI_API_KEY'] = os.getenv("OPENAI_API_KEY")
os.environ["LANGCHAIN_API_KEY"] = os.getenv("LANGCHAIN_API_KEY")
os.environ["LANGCHAIN_TRACING_V2"] = "true"
os.environ["LANGCHAIN_PROJECT"] = os.getenv("LANGCHAIN_PROJECT")
```

load_dotenv() : This line reads the _.env file and makes the keys available in your script.

```
os.environ['OPENAI_API_KEY'] = os.getenv("OPENAI_API_KEY"): This tells Python:
```

Look inside the <u>.env</u> file, get the <u>OPENALAPLKEY</u>, and make it available for use."

(You do the same thing for LangChain (a tool to help with Al apps)

Load the OpenAl Chat Model

Purpose: Create an instance of OpenAl's latest model (gpt-40).

from langchain_openai import ChatOpenAl

Ilm = ChatOpenAI(model="gpt-40") # Uses GPT-4 Turbo (fast & powerful)
print(Ilm) # Prints model config (confirms setup)

Output:

```
client=<OpenAl Completions object>
model_name='gpt-4o'
openai_api_key=SecretStr('*********')
```

3 Send a Direct Prompt

Use .invoke() to send a single question and get a response.

result = Ilm.invoke("What is generative AI?")

print(result)

Key Notes:

- invoke() sends a single prompt to the model.
- Response includes metadata (token usage, model details).

Create a Prompt Template

What?

Use ChatPromptTemplate to define reusable prompt formats.

Why?

Makes it easy to dynamically insert user input and apply structure to conversations.

This code sets up a **chat prompt template** for an Al model — basically a way to tell the model:

- 1. What kind of role it should play (like an expert).
- 2. What the user will ask (using a placeholder).

```
from langchain_core.prompts import ChatPromptTemplate

prompt = ChatPromptTemplate.from_messages([
          ("system", "You are an expert Al Engineer. Provide me answers based on the e questions"),
          ("user", "{input}")
])
```

Key Notes:

system sets the Al's behavior (e.g., "expert Al Engineer").

• {input} is a placeholder for dynamic user queries.

ChatPromptTemplate.from_messages([...])

- This creates a conversation structure for the Al.
- ("system", "...")
 - Tells the Al what kind of role it should play.
 - Like: "Pretend you're an expert AI Engineer."
- ("user", "{input}")
 - This is a **placeholder**. Whatever you pass in later as "input" will go here.
 - For example, if you ask "What is a neural network?", that will replace {input}.
 - It tells LangChain:
 - "I'll give you the actual question later when I run this prompt."
- The **system** message sets the Al's tone and purpose.
- The **user** message is the question.



- If you skip the system role, the AI may behave in a generic way.
 - Using system helps customize behavior: teacher, coder, chef, therapist, etc.

Suild a Chain (Prompt → Model → Output Parser)

Purpose: Combine components into a **pipeline** for structured AI responses.

A Chain Without Parser (Raw Al Response):

```
chain = prompt | Ilm # Pipe: Prompt → Model
response = chain.invoke({"input": "What is LangSmith?"})
print(response)
```

Output:

• A detailed response about LangSmith (logging, testing, monitoring).

It's using the pipe operator () to connect two things:

- A **prompt** (your message template)
- An **LLM** (a language model like OpenAl's GPT)

So it's saying:

"Take this prompt, fill it in with input, then send it to the language model (LLM) to get a response."

6 Add an Output Parser

What?

Convert the LLM response into a plain string (if it's an object).

Why?

Standardizes the output format for further use in apps or logs.

```
from langchain_core.output_parsers import StrOutputParser

output_parser = StrOutputParser()

# Full chain: prompt → LLM → parse response
```

chain = prompt | Ilm | output_parser

response = chain.invoke({"input": "Can you tell me about Langsmith?"})
print(response)

Output:

"LangSmith is a tool for debugging and monitoring LLM applications..."

6 Key Takeaways

- ✓ LangChain simplifies Al workflows (prompts, models, parsing).
- ✓ LangSmith tracks interactions (useful for debugging).
- **Chaining components** makes Al apps modular & reusable.