Session - 4&5

Date: October 11, 2025

Detailed Course Notes: Agentic Al

Session 1: Intro & Agentic Frameworks

Initial Recap (N8N Flow)

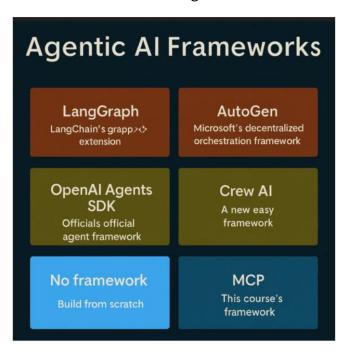
- We briefly looked at an N8N workflow diagram.
- It showed a flow: Gmail Trigger -> Code -> AI Agent -> Code1 -> Redis.
- The AI Agent used tools like OpenAI Chat Model, gmail label tool, and gmail draft tool. This was a good visual of what we're about to build manually.

Topic 0: What are our options? (Agentic Al Frameworks)

This is about *how* to build agents. There's a spectrum from easy/manual to complex/powerful.

- Complexity 1: (The "Do It Yourself" way)
 - No Framework: Just using raw API calls to LLMs (like OpenAI). This is what we did in Lab 1 & 2.
 - Pro: Full control.
 - Con: You have to build everything (state management, tool calling logic) yourself.
 - MCP (Model Context Protocol): From Anthropic. It's a standard for models to connect to each other without needing as much "glue code" from us.
- Complexity 2: (The "Managed" way)
 - OpenAl Agents SDK: The "official" toolkit from OpenAl for building agents.
 - Crew Al: A newer, "easy" framework. It's low-code and uses .yaml files for configuration. Seems good for simple agent teams.
- Complexity 3: (The "Heavy-Duty" way)
 - LangGraph: This is an extension of LangChain. It's graph-based, which means it's really good for complex, cyclical workflows where agents might need to loop or make decisions.

 AutoGen: Microsoft's big framework. It's for "decentralized multi-agent orchestration." This is for when you have many specialized agents all collaborating.



Lab Setup & Core Concepts

Step-by-Step: Getting the Labs Running

- 1. Clone Repo: git clone https://github.com/TEJAPS/agentic-notebooks.git
- 2. **Get IDE:** Download **Cursor** (https://cursor.com/). It's an AI-assisted VS Code fork.
- 3. **Install uv:** This is a new, super-fast Python package manager. We *must* use this.
 - Run this in PowerShell: powershell -ExecutionPolicy ByPass -c "irm https://astral.sh/uv/install.ps1 | iex"
- 4. Create Environment: cd into the repo folder and run:
 - o uv sync
 - This reads the pyproject.toml file, creates an isolated virtual environment (.venv folder), and installs all the packages.
- 5. **Get API Key:** Go to https://platform.openai.com/settings/organization/api-keys and create a new secret key.
- 6. **Set Environment Variable:** Create a .env file in the project folder and add your key:
 - OPENAI_API_KEY=sk-proj-YOUR_KEY_HERE

Key Definitions

This is the most important part for understanding the course.

Agent:

An **autonomous decision-maker**. It's a program where the LLM's output *controls the workflow*. It can **reason**, maintain **state** (memory, goals), and dynamically decide *what to do next* (e.g., call a tool, talk to another agent).

Example: A ResearchAgent that decides whether to use Google or search a database based on the query.

Tool:

A **non-intelligent function** or API that the agent can *call* to act in the world. It doesn't reason; it just *does one specific thing*.

Example: GoogleSearchTool, PythonREPLTool, DatabaseTool, gmail_draft_tool.

• Resource:

External data or a persistent store. It's *not* executable logic; it's just information that an agent or tool can **read from** or **write to**.

Example: A Vector DB (Pinecone), a SQL database, or just a summary.txt file (like in Lab 3).

• LLM Node (Non-Agent):

A simple, "dumb" LLM call. It performs a **fixed subtask** (like summarizing or extracting keywords) as part of a predefined path. It has *no decision-making power*.

Workflow Patterns & Lab 1-2

5 Agentic Workflow Design Patterns

These are the "plays" we can run to build complex systems.

1. Prompt Chaining:

- The simplest. Decompose a goal into fixed, sequential steps.
- IN -> LLM1 -> Gate -> LLM2 -> OUT

2. Routing:

- An "LLM Router" classifies the input and directs it to the right "expert" LLM or workflow.
- o IN -> LLM Router -> (Path A, Path B, Path C) -> OUT

3. Parallelization:

- A "Coordinator" breaks a task into independent parts, runs them at the same time (concurrently), and an "Aggregator" merges the results.
- IN -> Coordinator -> (LLM1, LLM2, LLM3) -> Aggregator -> OUT

4. Orchestrator-Worker:

- A "master" Orchestrator agent dynamically assigns complex subtasks to "worker" LLMs and then synthesizes their outputs. This is more flexible than Parallelization.
- o IN -> Orchestrator -> (LLM1, LLM2, LLM3) -> Synthesizer -> OUT

5. Evaluator-Optimizer (Used in Lab 3!)

- A quality-control loop. One LLM generates a solution, and a second LLM evaluates it. If it's rejected, it sends feedback to the generator to rerun and improve the answer.
- IN -> LLM Generator --(Solution)--> LLM Evaluator --(Accepted)--> OUT
- o ^ |--(Rejected w/ Feedback)--|

Lab 1: "No Framework" Agent

- Build an agentic workflow using only raw OpenAI API calls.
- Showed us how to manually construct the messages array, make the API call (openai.chat.completions.create(...)), and parse the response. We built the "agent" logic ourselves.

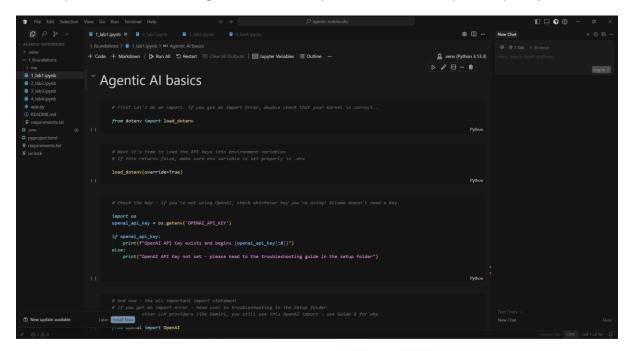
Lab 2: Multi-Model Agent

- Use different LLMs within the same workflow.
- Models We Used:
 - 1. OpenAl: gpt-4o-mini
 - 2. Google: gemini-2.0-flash
 - 3. **Ollama:** For running models *locally* (like llama3.1 or qwen).

Ollama Setup:

- Install from https://ollama.com/.
- o It runs locally at http://localhost:11434/.
- Need to restart Cursor/VS Code for it to detect the ollama app.

• **Key Takeaway:** Showed how to abstract the LLM call so we can easily switch providers. We saw big differences in speed, cost, and response quality.

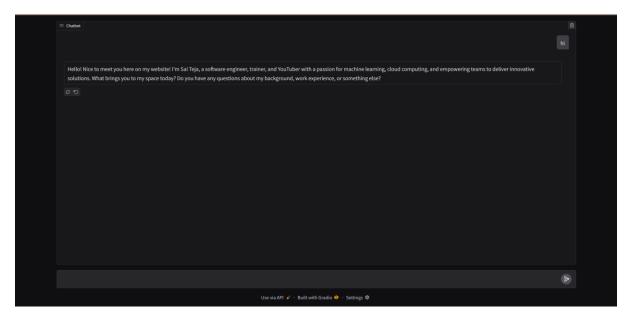


Resources, Tools, & Lab 3-4

Lab 3 - Resources & Evaluator Pattern

- Build a personalized chatbot using Resources.
- Resources Used:
 - 1. Our **LinkedIn profile** (downloaded as PDF and put in the /me folder).
 - 2. A summary.txt file we wrote about ourselves.
- The agent was prompted to read these files to get context before answering questions "as us."
- Pattern Used: Evaluator-Optimizer.
 - o We created a Pydantic model for the evaluation:
 - # Create a Pydantic model for the Evaluation
 - o from pydantic import BaseModel
 - class Evaluation(BaseModel):
 - o is_acceptable: bool
 - feedback: str

- An evaluator LLM would check the generator's answer. If is_acceptable ==
 False, the agent would re-run the request using the feedback.
- **UI:** We used **Gradio** to create a simple web chatbot interface.
 - gr.ChatInterface(chat, type="messages").launch()



Lab 4 - Tools & Notifications

- · Give our agent **Tools** to perform actions.
- Tools We Built:
 - 1. A tool to **record unknown questions** (to capture queries the LLM couldn't answer).
 - 2. A tool to **record user details** (for context/personalization).
 - 3. A **Pushover** integration tool.
- Pushover Setup (For Notifications):
 - 1. Go to https://pushover.net/ and create an account.
 - 2. Copy your **User Key** -> save as PUSHOVER_USER in .env.
 - Create a new "Application" -> get an API Token -> save as PUSHOVER_TOKEN in .env.
- **Key Takeaway:** This was a full end-to-end pipeline: **Personalization** (**Resources**) + **Action** (**Tools**) + **Notifications**.

Assignments

1. Complete all 4 labs.

2. Personalized LLM Use Case (Team Project):

- o Each team member must choose one (1) tool and one (1) resource.
- Design and implement a real-world solution using them with your personalized LLM.
- 3. Finish all previous pending assignments.