

# Lecture 15: Building Your Own Agentic Framework

---

## Learning Objectives

By the end of this lecture, you should be able to:

- Understand the core architectural components of an agentic framework.
  - Design a modular and extensible agent framework from scratch.
  - Implement a reusable loop for reasoning, memory, and tool use.
  - Build a lightweight, pluggable system tailored to your use case.
- 

## Key Concepts

Why Build Your Own Framework?

- Greater control over execution, memory, and integration.
- Tailored to domain-specific constraints (e.g., latency, safety).
- Avoids over-engineering or dependency overload.

Core Building Blocks

- **Controller:** Orchestrates the reasoning-action loop.
  - **Reasoner:** Generates thoughts, actions, and decisions using the LLM.
  - **Toolset:** Modular tools callable by the agent.
  - **Memory:** Stores relevant history or retrieved knowledge.
  - **Goal Tracker:** Optional state machine to monitor progress.
- 

## Required Tools/Libraries

- Python
  - OpenAI / Hugging Face API
  - Optional: FAISS or Chroma for memory, JSON schema for tool definition
- 

## Hands-on Exercise: Build a Mini Agent Framework

**Goal:** Create a simple Python-based agent framework with reasoning, tool use, and memory support.

Step 1: Define tool functions

```
tools = {  
    "search": lambda q: f"[Search result for '{q}']",  
    "math": lambda expr: str(eval(expr))  
}
```

## Step 2: Define LLM reasoner function

```
def reasoner(prompt):  
    # Placeholder for actual LLM call  
    print("LLM receives prompt:\n", prompt)  
    return input("LLM response: ")
```

## Step 3: Implement the agent loop

```
memory = []  
goal = "Find the square root of 144 and explain it."  
  
while True:  
    context = "\n".join(memory[-3:])  
    prompt = f"Goal: {goal}\nContext:\n{context}\nWhat should I do next?"  
    response = reasoner(prompt)  
  
    memory.append(response)  
  
    if "Final Answer:" in response:  
        break
```

## Step 4: Review and extend

- Add regex to extract tool calls.
- Log each step for transparency.
- Modularize components into classes: Agent, Tool, Memory, GoalTracker.

---

### Bonus:

- Add a configuration file (e.g., YAML) to define available tools and agent personality.
  - Support multiple agents working collaboratively.
  - Integrate vector memory for semantic recall.
-