

Lecture 08: Multi-step Autonomy and Looping

Learning Objectives

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

- Understand how agents maintain autonomy across multiple steps.
 - Implement the action-observation loop used in agent execution.
 - Handle intermediate results and dynamic decision-making.
 - Build a simple looped agent to solve compound tasks.
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Key Concepts

Agent Loop Architecture

- The agent performs a **thought–action–observation** cycle:
 1. **Thought**: Decide what needs to be done.
 2. **Action**: Call a tool or perform a step.
 3. **Observation**: Interpret the result of the action.
 4. **Repeat** until the goal is achieved or a stopping condition is met.
- This creates an iterative loop that allows agents to adapt as new data arrives.

Benefits of Looped Agents

- Can handle tasks with:
 - Uncertainty
 - Changing goals
 - External dependencies
- Capable of retrying or correcting failures

Example Use Cases

- Research assistant searching for multiple facts
 - Code generation agent refining its output
 - Task automation agent solving a multi-part goal
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Required Tools/Libraries

- Python
 - OpenAI API (or similar LLM)
 - Tool functions (calculator, search, etc.)
 - LangChain (optional for structured agent loops)
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Hands-on Exercise: Multi-step Research Agent

Goal: Build an agent that answers a research-style question by looping through multiple steps.

Step 1: Define tools

```
def search(query):  
    # Simulate a search tool (replace with real API later)  
    database = {  
        "Who is Elon Musk?": "Entrepreneur and CEO of Tesla and SpaceX.",  
        "What is SpaceX?": "A private aerospace company founded by Elon Musk.",  
    }  
    return database.get(query, "No results found.")
```

Step 2: Simulate a multi-step loop

```
question = "What company was founded by the person who leads Tesla?"  
  
# Step 1 - Thought: Need to find out who leads Tesla  
result_1 = search("Who is Elon Musk?")  
  
# Step 2 - Thought: Now find out what company he founded  
result_2 = search("What is SpaceX?")  
  
# Final Answer  
print("SpaceX is the company founded by Elon Musk.")
```

Step 3: Add LLM to generate each step

- Let the LLM generate intermediate questions.
- Use results to guide the next step.
- Track all steps in a log or memory buffer.

Bonus:

- Add stopping criteria (e.g., "Final Answer" in the response).
 - Limit the number of steps to avoid infinite loops.
 - Build a reusable agent loop function to handle any input goal.
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