

## DAY 8 – Planning Agents (Task Decomposition + Reasoning)

### What is Planning in Agentic AI?

Planning Agent takes a large user goal and intelligently breaks it into smaller, manageable steps before any action is taken.

### Why break down goals?

If you simply tell a chatbot:

“Research RAG → summarize → save notes → create PDF.”

It will likely fail because it tries to do everything at once without structure.

### What does a planner agent do?

- Splits the goal into small tasks
- Understands the order and dependencies between tasks
- Creates a clear, step-by-step plan
- Returns output in a structured format (like JSON)
- Leaves execution to a separate agent (Executor)

Think of planning as the brain that thinks and plans ahead. Execution is like the hands that carry out the plan.

### Why Planning is Needed?

Agents often fail because they:

- Skip important steps
- Hallucinate or invent wrong details
- Execute the wrong actions
- Call the wrong tools

Planning fixes these by:

- Creating a clear roadmap
- Avoiding errors through stepwise logic
- Making the agent's behavior consistent and structured

### 3 Major Planning Styles (Simplified)

Agents think in different ways depending on the task:

#### 1. Chain of Thought (CoT)

Think step-by-step. The LLM writes down each reasoning step before answering.

Example task:

“How to study RAG?”

CoT output:

- Learn embeddings
- Learn vector databases
- Learn document splitting
- Learn retrieval
- Learn evaluation

## 2. ReAct (Reason + Act)

The LLM alternates between:

- Thinking
- Deciding an action
- Using tools
- Thinking again

This helps it handle

- Web search
- File reading
- Calculations
- API calls

## 3. Tree of Thoughts (ToT)

The LLM explores multiple reasoning paths in parallel, then chooses the best one.  
Used in creative writing, coding, and complex planning.

### [How LangChain and LangGraph Do Planning](#)

LangChain

- Uses carefully designed prompts
- Uses templates and tools integration
- Planning happens through LLM reasoning in prompts

LangGraph

- More advanced graph-based connections
- Connect nodes like planner → executor → evaluator
- Highly suitable for production use

### [Build Your Own Planner Agent \(Python Code\)](#)

Create a simple, practical planning agent that takes a goal and returns steps in JSON.

Step 1 — Create `day6\_planner.py`

Step 2 — Add Imports

```
```python
from langchain_openai import ChatOpenAI
from langchain.prompts import PromptTemplate
import json
```

```

### Step 3 — Define Planner Prompt

```
```python
planner_prompt = """

```

You are a Planning Agent.

Your job:

- Understand the user goal.
- Break it into 3-7 clear steps.
- Do NOT execute the steps.
- Output in JSON only.

User Goal: {goal}

Return strictly this JSON:

```
{  
  "goal": "...",  
  "steps": ["step 1", "step 2", "..."]  
}
```

### Step 4 — Setup the LLM (OpenRouter Compatible)

```
```python  
llm = ChatOpenAI(  
    model="gpt-4o-mini",  
    temperature=0 # deterministic output  
)  
..
```

Why temperature=0?

So the plan is consistent and reproducible.

### Step 5 — Define Planner Function to Make Plans

```
```python  
def make_plan(goal):  
    prompt = planner_prompt.format(goal=goal)  
    response = llm.invoke(prompt)  
    return json.loads(response.content)  
..
```

### Test Your Planner

```
```python  
plan = make_plan("Research RAG techniques and write a 200-word  
summary.")  
print(plan)  
```
```

Expected output:

```
```json  
{  
  "goal": "Research RAG techniques and write a 200-word summary",  
  "steps": [  
    "Identify high-quality sources on RAG",  
    "Read 2-3 articles or documentation pages",  
    "Extract important key points",  
    "Write a rough summary",  
    "Edit it to exactly 200 words"  
  ]  
}```
```

This agent \*\*does not execute any task\*\*, it only \*plans\* the steps logically.

### Practice Tasks

Try these user goals:

1. "Plan a 7-day weight loss diet."

You should get 7 clearly defined steps.

2. "Search 5 AI tools, compare features, create a table."

Expected steps: collect tools, search, extract features, build table, summarize.

3. "Learn RAG in 7 days."

Expect daily topics like: Day 1 embeddings, Day 2 vector DB, etc.

### Simple Explanation of Key Concepts

#### **Task Decomposition**

Breaking a big task into small, achievable sub-tasks.

Example: Make a YouTube video

- Script
- Voiceover
- Editing
- Thumbnail
- Upload

#### Reasoning

Agent's ability to "think before acting":

- Choose the right tool
- Decide the correct order
- Avoid mistakes

#### **Why JSON Output?**

- Clean and well-structured
- Easy to parse
- Can be used by executors later
- Ensures predictability