

# Intro to Agentic AI

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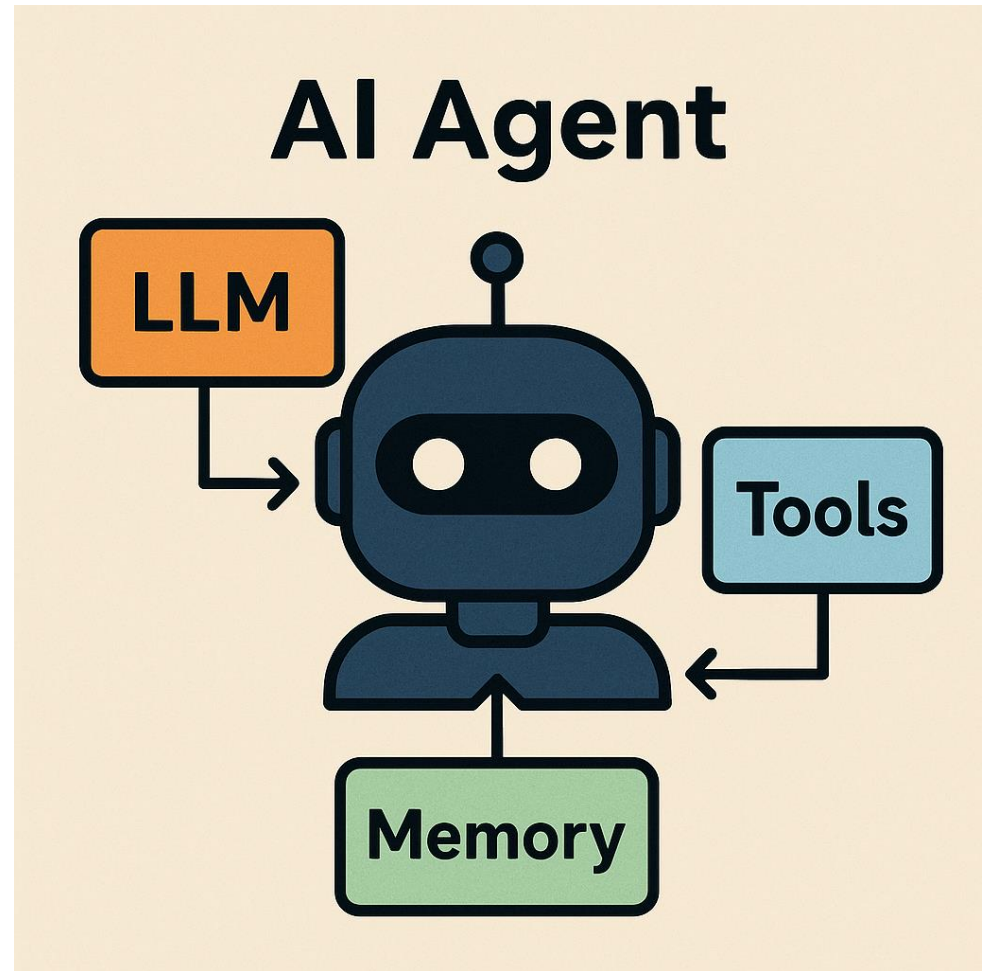


# What We Will Learn Today

- What is an AI agent?
- Why agents matter now
- Simple LangChain agents
- Tool calling & ReAct
- Hands-on mini-exercise

# What Is an AI Agent?

- LLM that can plan, act, use tools
- Moves beyond static prompting
- Acts in multi-step loops



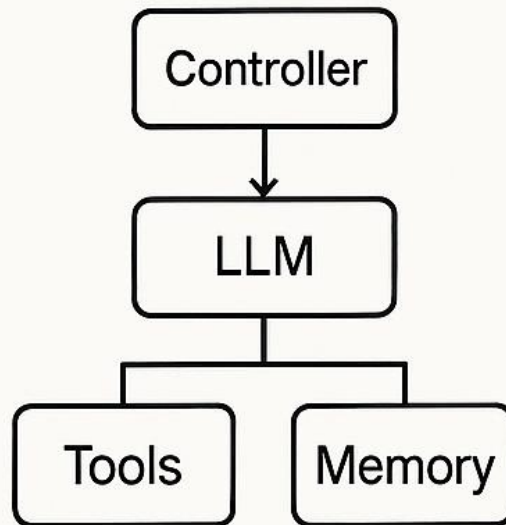
# Why Agents Now?

- Better reasoning models
- Tool calling standardization
- Library support (LangChain)

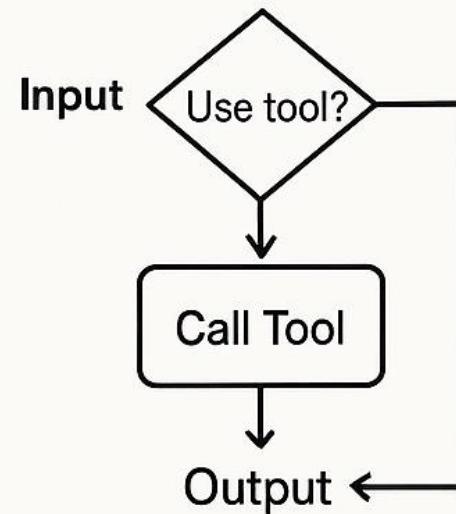
# Agent Architecture (Simple)

- LLM core
- Tools
- Memory (optional)
- Controller / Executor

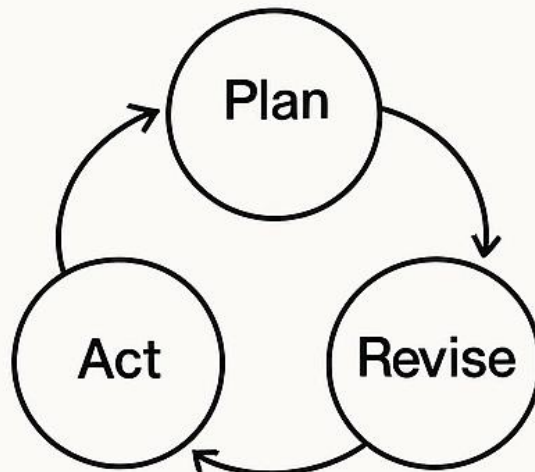
## Agent Architecture



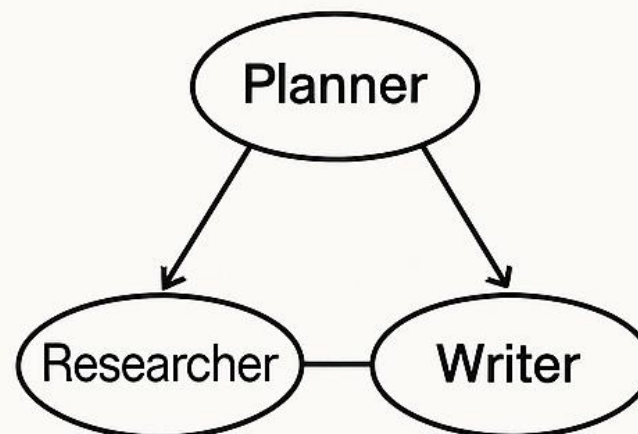
## Tool Calling Flow



## Planning Loop



## Multi-Agent System

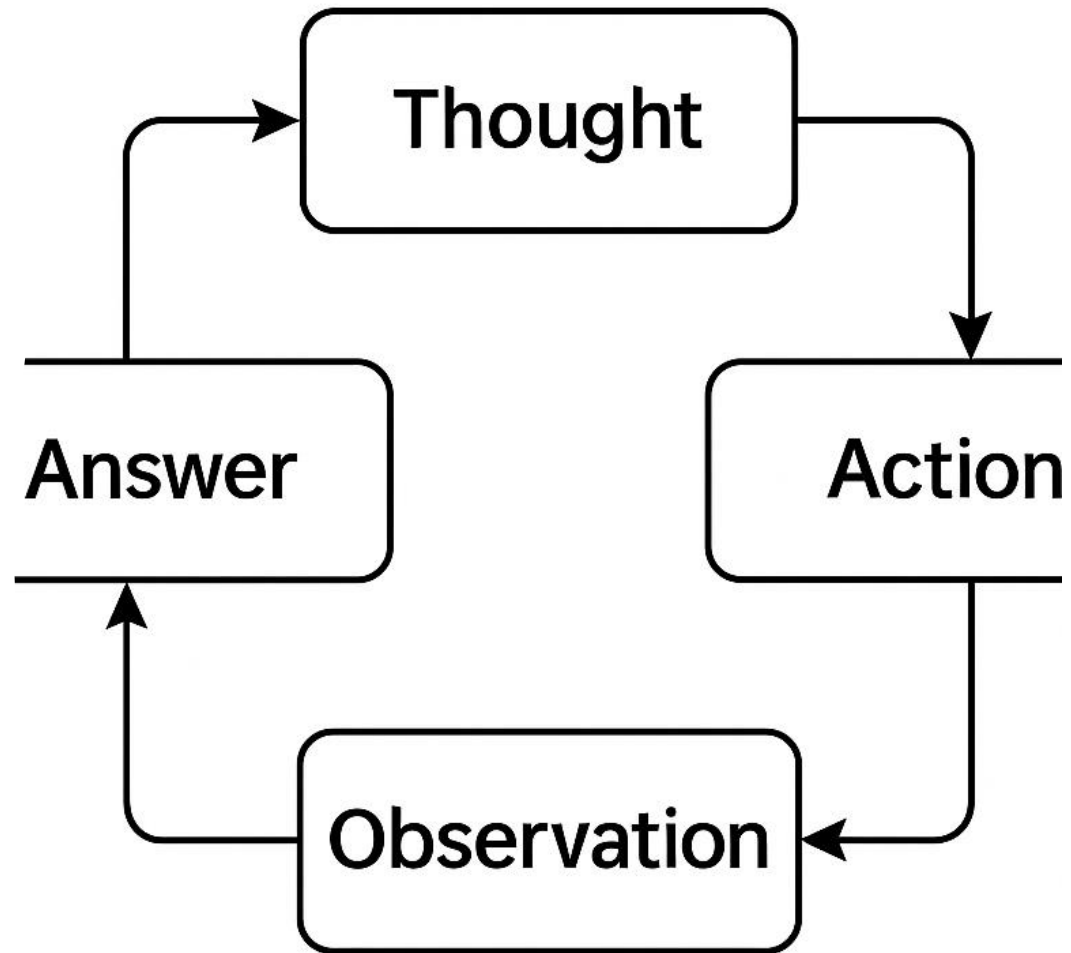


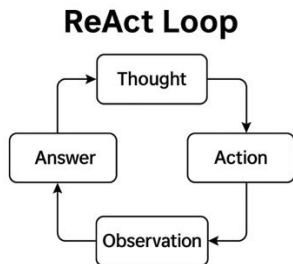
# ReAct Loop

## LLM Core

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- Thinks
- Plans
- Decides next action





# ReACT Loop Example

Best coffee shop nearby?

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**Thought:** I should search the user's query.

**Action:** search["best coffee shops in Denver"]

**Observation:** 1) Huckleberry Roasters ... 2) Corvus Coffee ...

**Thought:** Now I can answer the user's question.

**Final Answer:** Here are the best coffee shops in Denver...



# Tools

- Calculator
- Search
- Python
- APIs

# Memory

- Optional
- Conversation history
- Vector memory

# Controller

- Routes between model, tools, memory

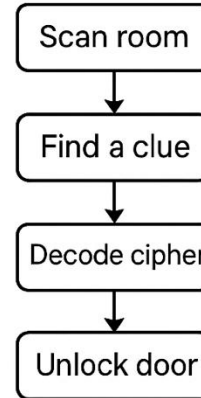
# ReAct Loop (Thought→Action→Obs)

- Reason step
- Take action
- Observe result
- Repeat

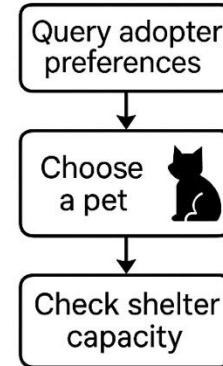
# ReAct Loop (Lab)

- Reason step
- Take action
- Observe result
- Repeat

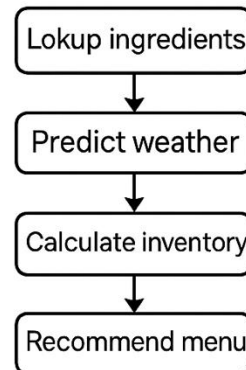
## AI Escape Room Agent



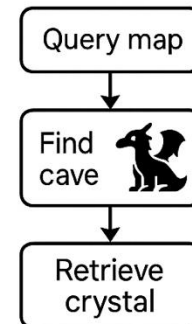
## Pet Adoption Agent



## AI Food Truck Manager



## Wizard's Quest Agent



# AI Escape Room Agent

Goal: Escape the locked room using clues and tools.

Tools:

- scan\_room(clue)
- decode\_cipher(text)
- check\_inventory(item)
- unlock(door)

Instructions:

1. Identify visible clues.
2. Use scan\_room to gather details.
3. Decode any symbols or text.
4. Check your inventory.
5. Determine the sequence to unlock the door.

# AI Food Truck Manager

Goal: Plan the most profitable lunch service.

Tools:

- lookup\_ingredients(item)
- predict\_weather(location)
- calculate\_inventory(levels)
- recommend\_menu(day)

Instructions:

1. Check weather to predict customer turnout.
2. Review inventory levels.
3. Build a menu based on demand + supplies.
4. Use calculate\_inventory to estimate costs.
5. Finalize menu and explain reasoning.

# Pet Adoption Agent

Goal: Find the best match between adopter and pet.

Tools:

- lookup\_pet\_traits(pet)
- query\_adopter\_preferences(name)
- check\_shelter\_capacity()

Instructions:

1. Gather adopter preferences.
2. Compare preferences to available pets.
3. Evaluate traits and capacity constraints.
4. Recommend a pet and justify your match.



# AI Travel Buddy Agent

Goal: Plan a weekend trip on budget.

Tools:

- search\_flights(city)
- hotel\_prices(date\_range)
- weather(city)
- walking\_distance(origin, destination)

Instructions:

1. Pick 2–3 destination candidates.
2. Find flight + hotel options.
3. Evaluate weather conditions.
4. Design a simple itinerary.
5. Stay within budget and explain trade-offs.

# Wizard's Quest Agent

Goal: Complete a magical quest safely.

Tools:

- spell\_lookup(spell)
- creature\_weakness(creature)
- map\_query(location)
- inventory\_check(item)

Instructions:

1. Map the route to your goal location.
2. Identify threats (creatures, barriers).
3. Look up spells or weaknesses.
4. Check inventory for magical items.
5. Produce a step-by-step quest plan.

# Detective Agent

Goal: Solve the mystery using evidence.

Tools:

- fingerprint\_scan(item)
- interrogate(character)
- analyze\_bloodtype(sample)
- lookup\_suspect\_history(name)

Instructions:

1. Identify key evidence at the scene.
2. Run scientific analyses.
3. Interrogate suspects.
4. Cross-check histories + clues.
5. Present your suspect and reasoning.

# Logistics Optimization Agent

Goal: Deliver all packages efficiently.

Tools:

- `route_distance(start, end)`
- `truck_capacity_query(truck)`
- `package_priority(package_id)`

Instructions:

1. Check package priority levels.
2. Assign packages to trucks based on capacity.
3. Map shortest delivery route.
4. Resolve conflicts (capacity, distance).
5. Deliver optimized plan + justification.

# When to Use Agents

- Multi-step tasks
- Data access
- Planning workflows

# When NOT to Use Agents

- Simple Q&A
- Single prompt tasks

# LangChain: Minimal Agent

- LLM + Calculator tool

```
from langchain_openai import ChatOpenAI
from langchain.agents import Tool,
AgentExecutor, create_react_agent
from langchain.tools import Calculator
```

```
llm = ChatOpenAI(model="gpt-4o-mini")
```

```
tools = [Calculator()]
```

```
prompt = """
You are a math agent. Use tools when needed.
"""
```

```
agent = create_react_agent(llm, tools, prompt)
executor = AgentExecutor(agent=agent,
tools=tools)
```

```
executor.invoke({"input": "What is 22 * 47?"})
```

# Code Demo: Math Agent

- LLM
- Calculator tool
- Agent executor



# Understanding Tool Calls

- LLM generates JSON/tool schema
- Controller executes call

# Exercise Setup

- Modify a simple agent
- Add or change one tool

# Exercise Instructions

- 1. Run starter agent
- 2. Add tool
- 3. Test

# Live Example: Adding a Search Tool

- Define fake search()
- Register tool
- Run agent

# Plan → Act → Review Agents

- Planning first
- Executing steps
- Refining

# Mini Workflow Agent

- Clarify
- Plan
- Execute

# Multi-Agent Systems

- Planner
- Researcher
- Writer

# Failure Cases

- Hallucinated tools
- Infinite loops
- Bad assumptions



# Safety & Guardrails

- Stop conditions
- Tool whitelists
- Retries

# Final Demo

- Agent solves a small problem
- Shows multi-step reasoning

# Summary

- Agents = LLM + tools + loops
- Great for multi-step tasks

# Assignment

- Build a two-tool agent
- Explain design choices