

# Introduction to

 LangChain &  LangGraph

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10/8/2025

# References Used Throughout the Slides

1. [A Hands-On Guide to Building Intelligent Systems](#), Antonio Gulli
2. [LangChain Academy](#) – curated courses & tutorials.  
GitHub: <https://github.com/langchain-ai/langchain-academy.git>
3. [LangGraph Cookbook](#)
4. [LangChain Tutorials](#)
5. [Interrupt](#) – the AI Agent Conference by LangChain
6. [Prof. Ghassemi Lectures and Tutorials](#), AI Agents lectures

# Agenda

0. Introduction
  - LLM vs Agent
  - LangChain vs LangGraph
1. Setup for Google Colab
2. Basic Usage
  - LangChain Expression Language (LCEL)
3. Routing vs Chaining
4. Tool Use
5. Reflection
6. Retrieval-Augmented Generation (RAG)
7. Multi-Agent Collaboration
8. Human in the Loop

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## 0. Introduction

- **LLM vs Agent**
- **LangChain vs LangGraph**

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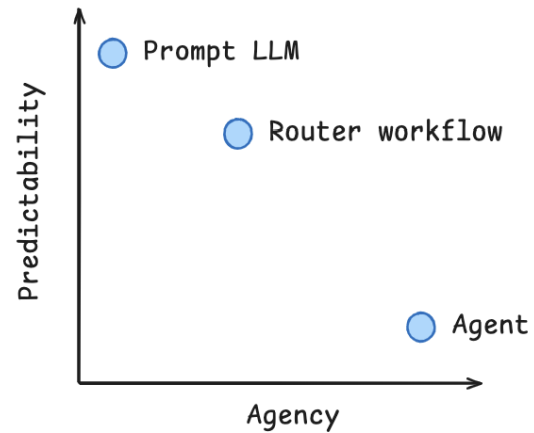
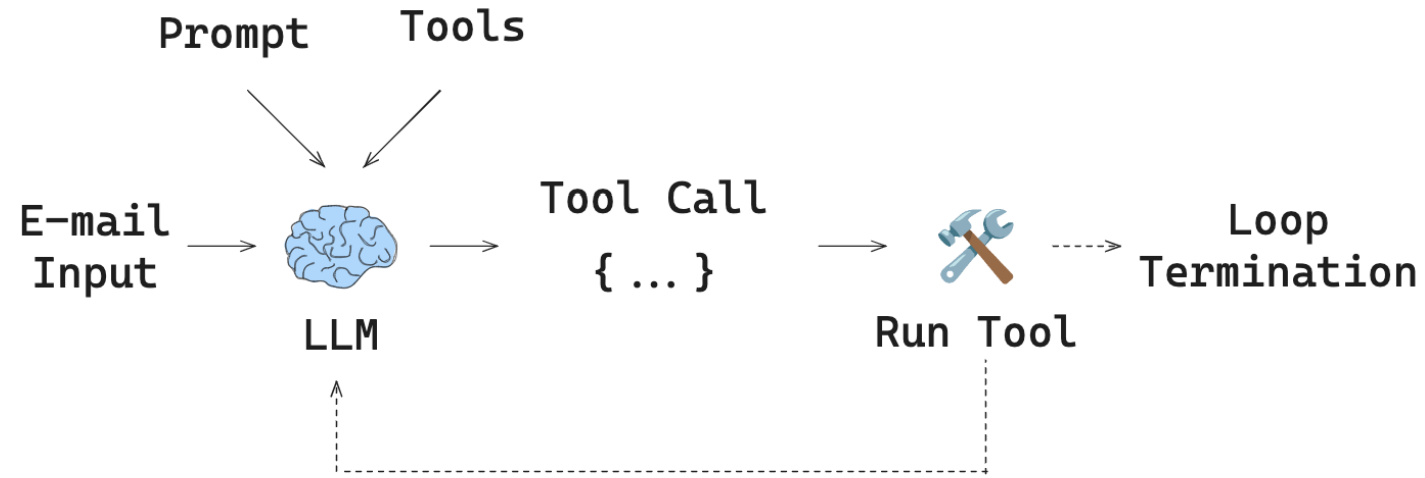
# 0) Introduction

- LLMs make it possible to embed intelligence into a new class of applications.
- Chat models are the foundation of LLM applications. They are typically accessed through a chat interface that takes a list of messages as input and returns a message as output.

# LLM vs Agent

- LLM: a text generator (no goals, tools, memory by itself).
- Consider loop: *Observe* → *Decide* → *Act* → *Reflect*.
- It is incredibly difficult to build systems that reliably execute on these tasks.
- Agents can automate a wide range of tasks that were previously impossible.
- Agent: LLM + tools + memory + policy that decides what to do next (answer, search, call a function, delegate, clarify).

# Agent Example



# LangChain vs LangGraph



## LangChain

- [LangChain](#) is a framework to help developers build applications using power of LLMs.
- Provides abstractions and components to easily create complex workflows, or "chains," by combining different parts.
- Provides [a standardized interface for chat models](#), making it easy to [access many different providers](#).



# LangChain vs LangGraph



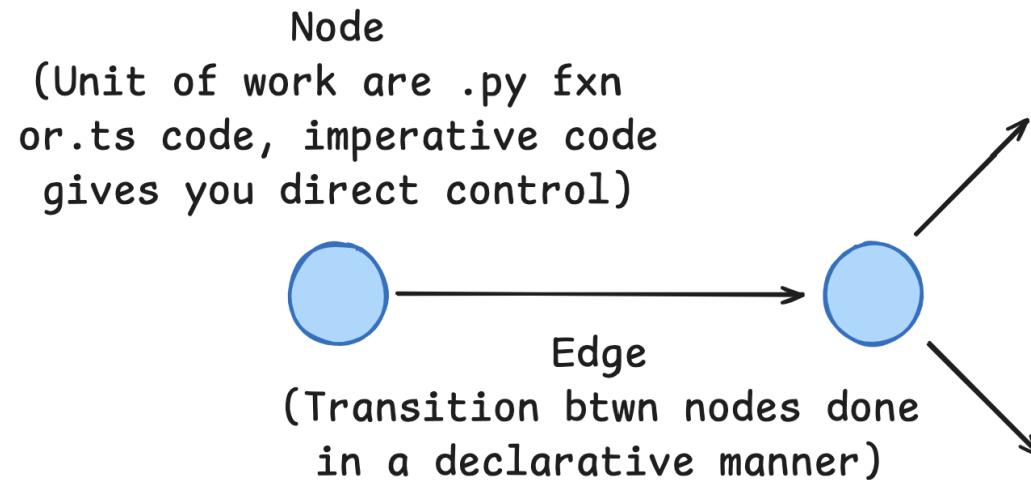
## LangGraph

- [LangGraph](#): an extension of LangChain
- Enables creation of multi-agent applications by representing the workflow as a graph.
- Helps developers add better precision and control into agent workflows
- Suitable for complexity of real-world systems.

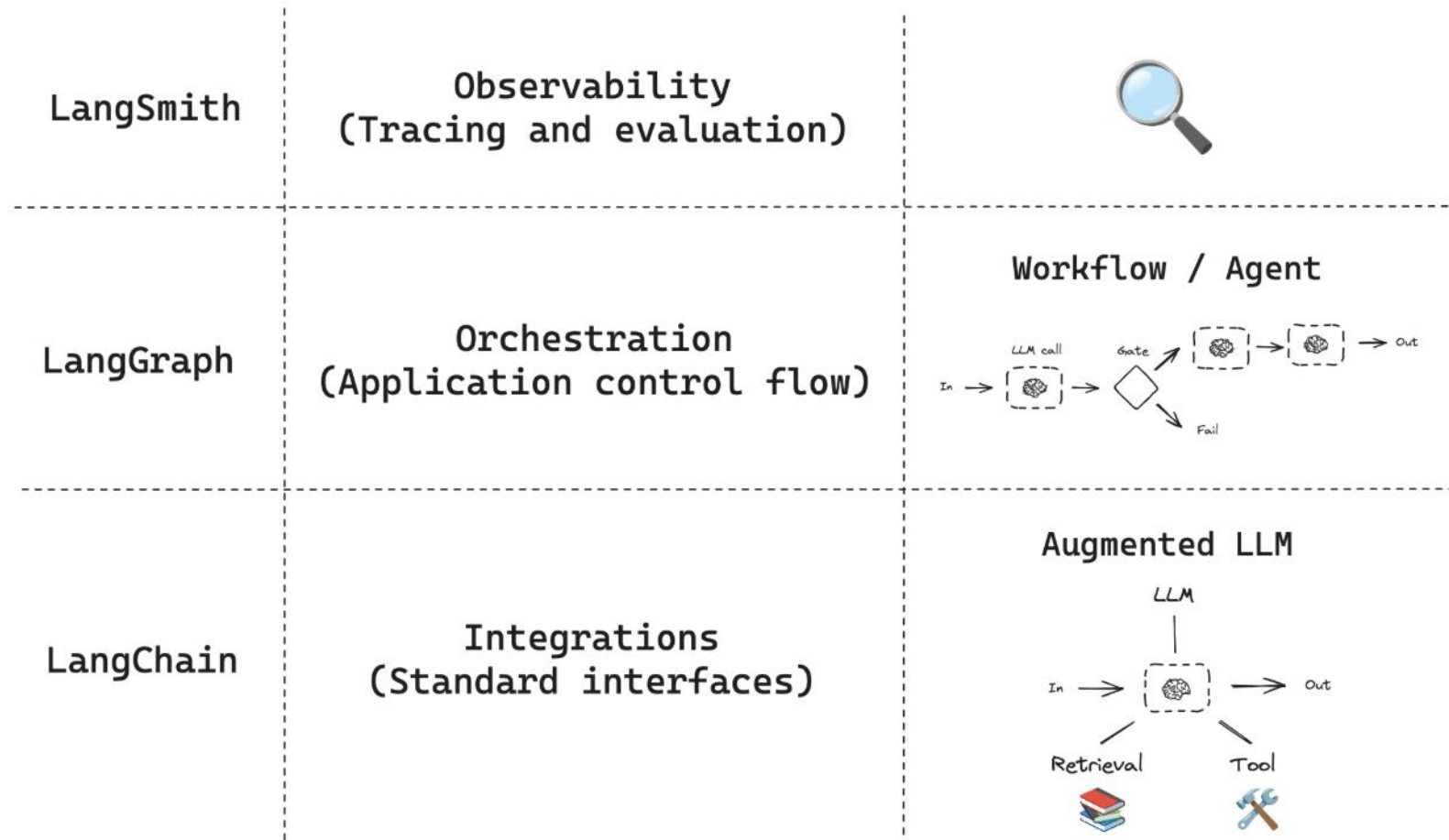
**Tip:** Start with LangChain for simple flows; move to LangGraph for collaboration/routing/retries/long-lived state.

# Node and Edge

LangGraph allows you to define nodes (which can be LangChain runnables or other Python functions) and edges (which define the transitions between nodes).



# Ecosystem



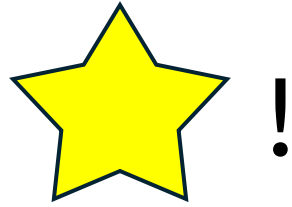
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# 1) Setup for Google Colab

- Requirements for today:
  - **HuggingFace account (for LLMs)**
  - **Personal Google account (for Google Colab)**
- Google Colab:
  - Free tier: usually NVIDIA T4 GPU. It's not the fastest GPU, but it's available for free.
  - Paid subscription (Colab Pro): faster GPUs, more memory, etc.
  - [Google Colab Pro](https://colab.research.google.com/signup) is free for students and teachers! Link: <https://colab.research.google.com/signup>

Link to GitHub  
Please Leave a



!



<https://github.com/maryambrij/langchain-langgraph-tutorials.git>

# 1) Setup for Google Colab

  **Please save a copy before starting to edit the current notebook!** File -> Save a copy in ...

- Running this entire workshop **without paid API keys**.
  - Using models from **HuggingFace**.
- Steps for authenticating with HuggingFace Hub :
  - Create a token in settings: <https://huggingface.co/settings/tokens>
  - Choose “write” option
  - Set it as a secret in Google Colab ("🔑" icon, left panel)
  - Name the secret key `HF_TOKEN`
  - Restart the session + select a GPU

# 1) Setup for Google Colab

- Install:

- ``transformers``
- ``accelerate``
- ``langchain``
- ``langgraph``
- ``langchain-huggingface``
- ``langchain-community``
- ``sentence-transformers``
- ``faiss-cpu``

- Load a small chat model **TinyLlama 1.1B Chat**



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## 2) Basic Usage

### LangChain Expression Language (LCEL)

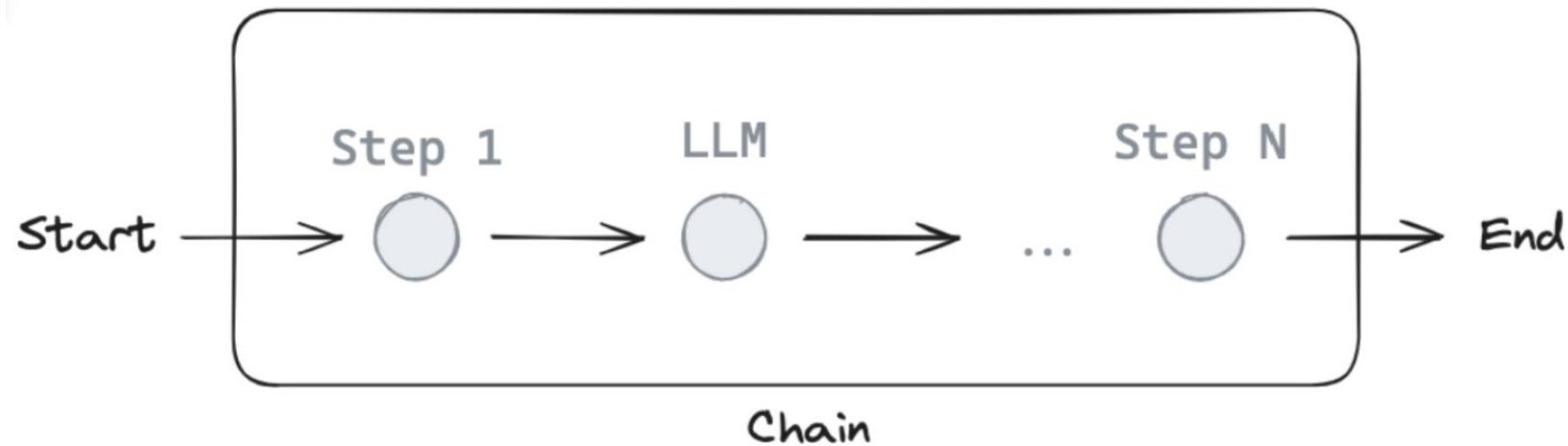
- A declarative way to compose chains.
- Combine various LangChain components, like prompts, models, and output parsers, into complex workflows.
- Build pipelines with features like streaming, async support, and parallel execution.
- Creation of chains by **piping components together using the `|` operator**.

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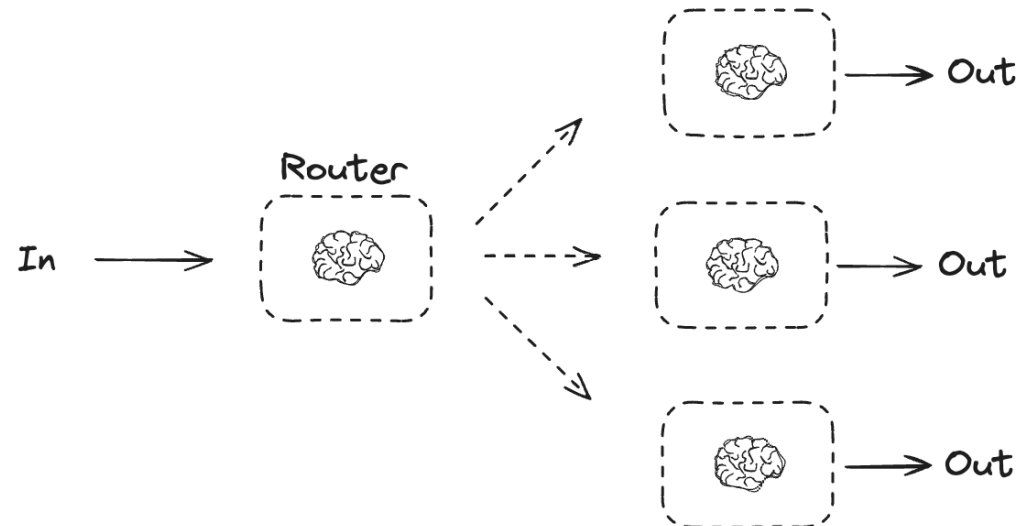
### 3) Routing vs Chaining

- **Chaining** can be thought of as a linear path:
  - `Step A` (Planner) -> `Step B` (Researcher) -> `Step C` (Writer).
- A fixed sequence of steps

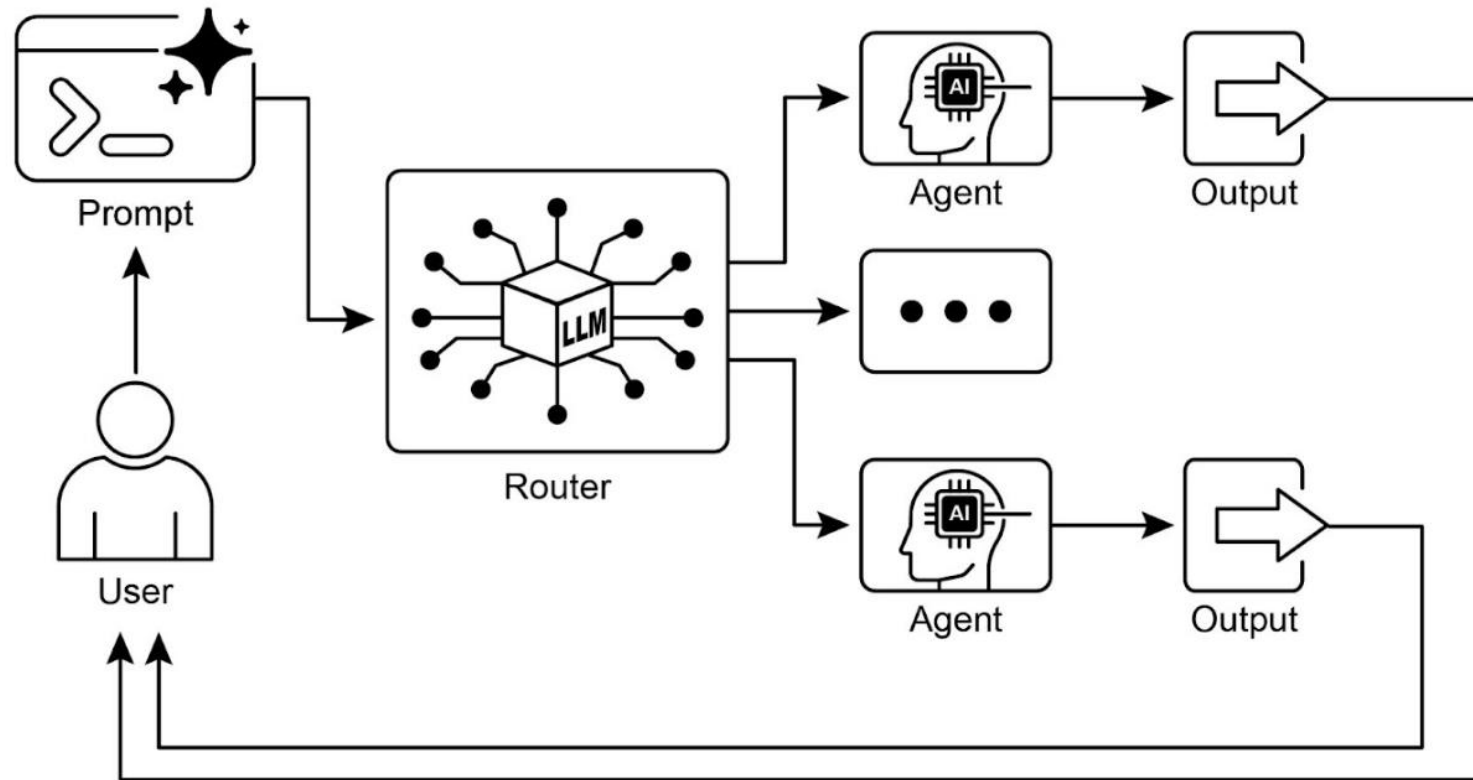


### 3) Routing vs Chaining

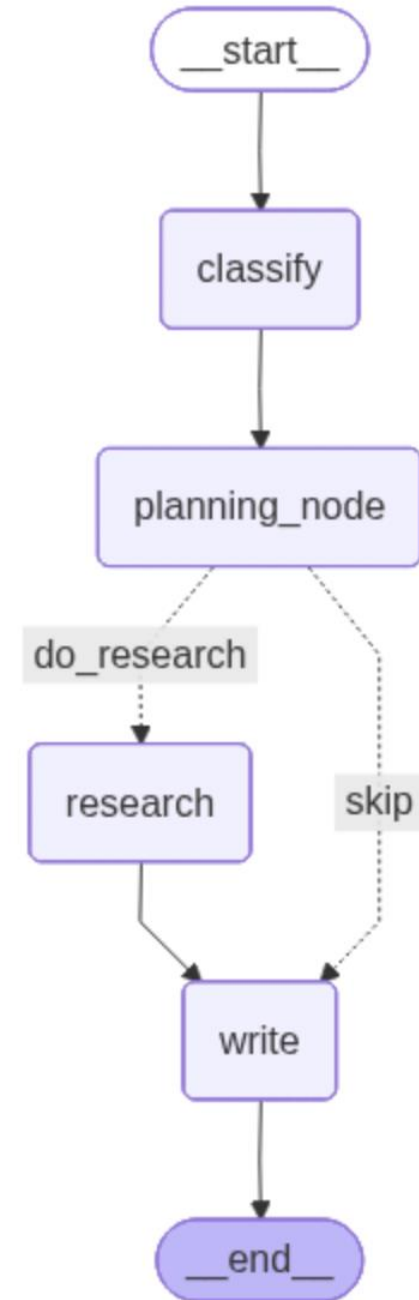
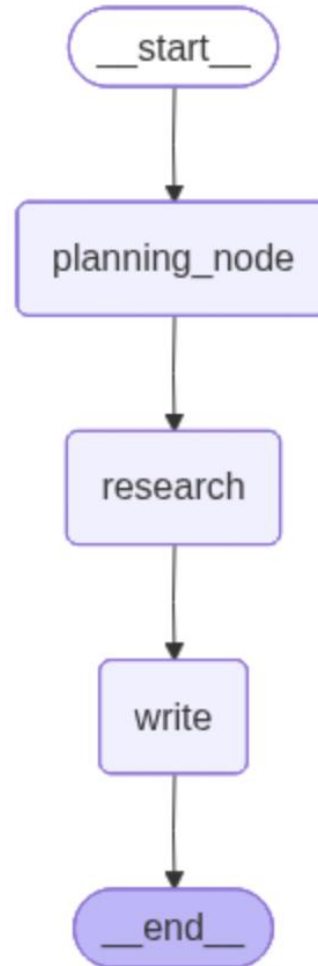
- **Routing:** More complex applications require dynamic workflows where the next step depends on the output of the previous one or on some condition.
  - Workflows with **loops** and **dynamic** paths
- Allows for **conditional** paths:
  - After `Step A` (Planner), based on the result, the workflow might go to `Step B` or `Step C`
- **LangGraph** supports conditional edges and routing logic



# LLM as a Router



# Example: Research Agent



# Agenda

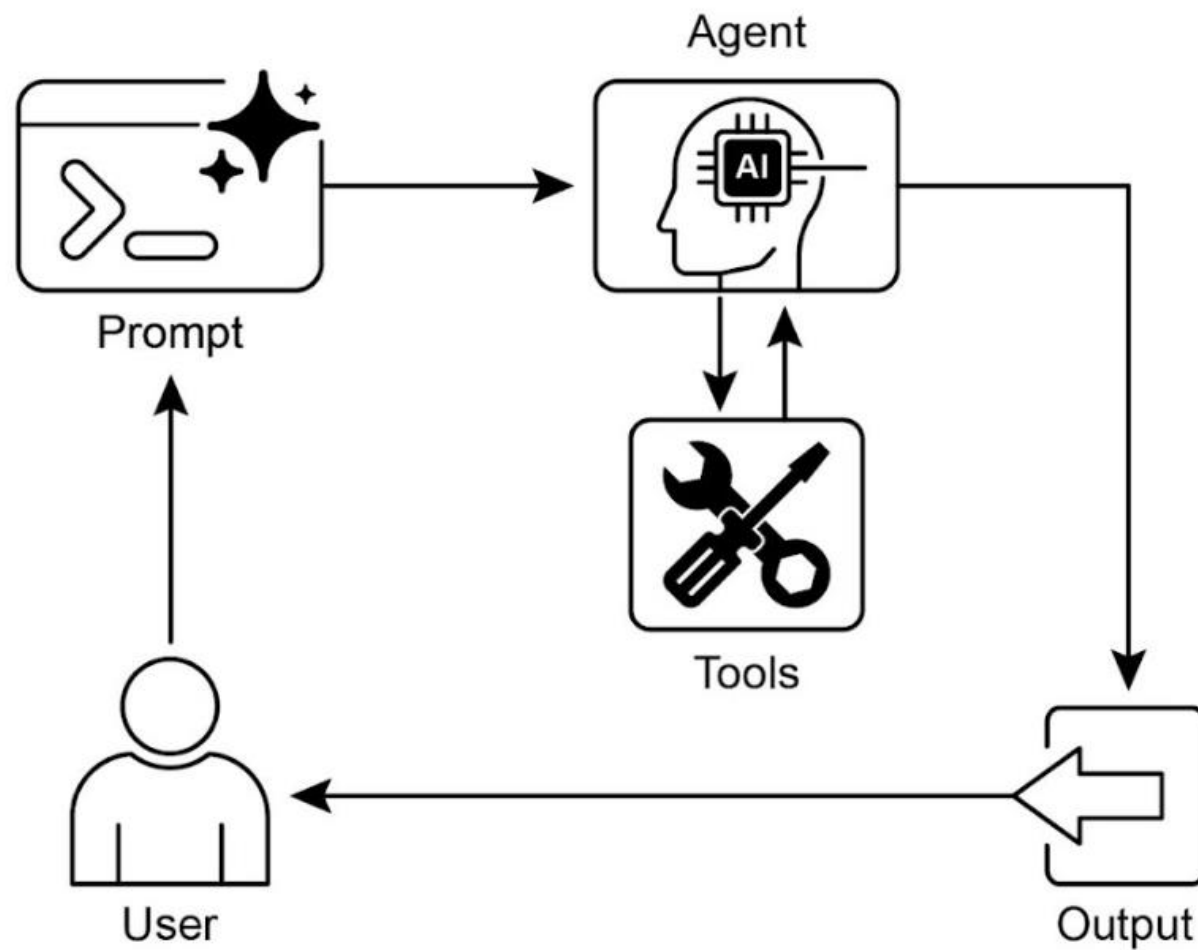
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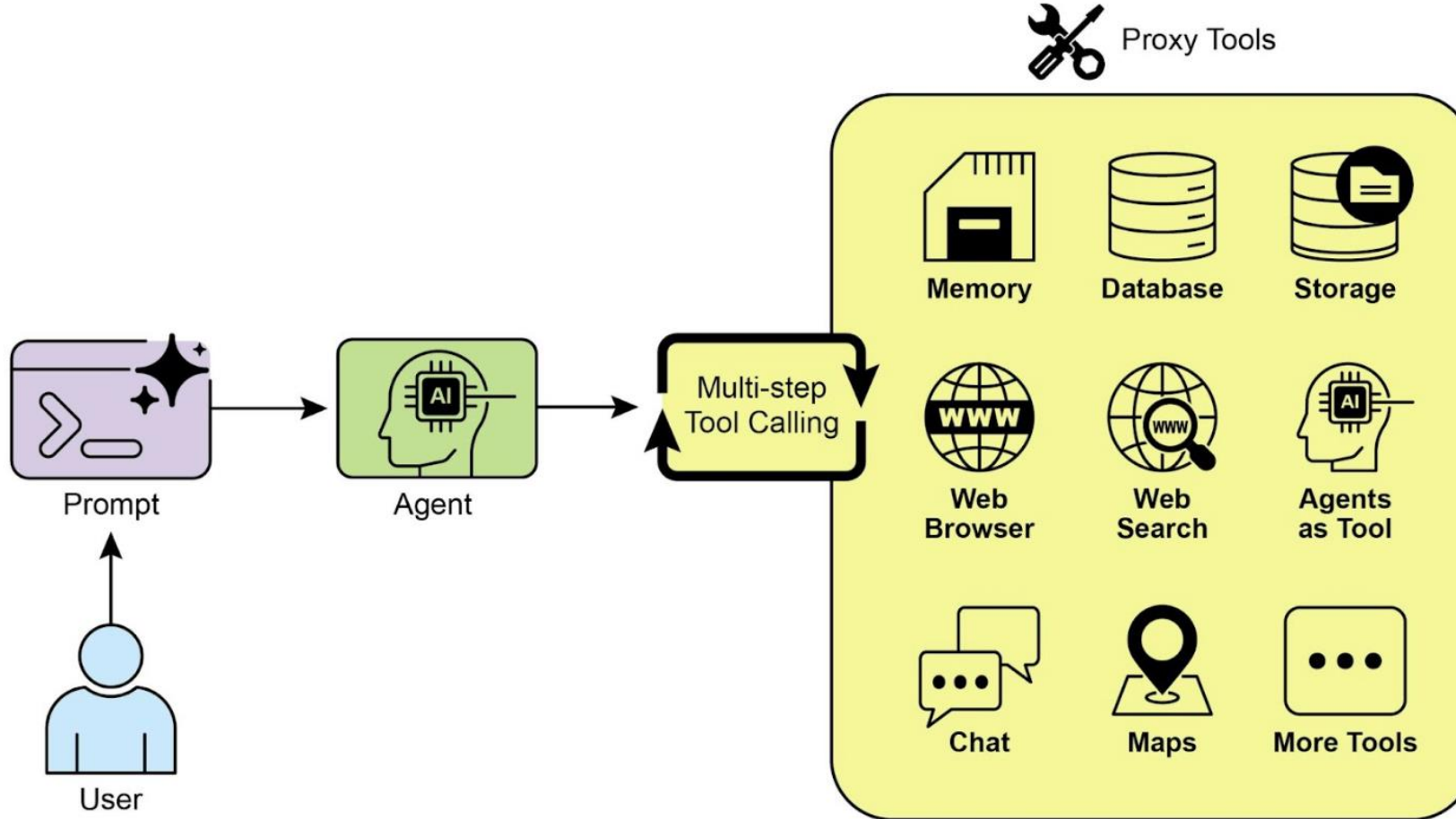
## 4) Tool Use

- Tools: external functionalities agents can use to extend their capabilities
- Can be anything: from a calculator or a search engine stub to more complex integrations with databases, APIs, or other services.
- Creating tools: can be done using the `@tool` decorator, which transforms Python functions into callable tools.

## 4) Tool Use



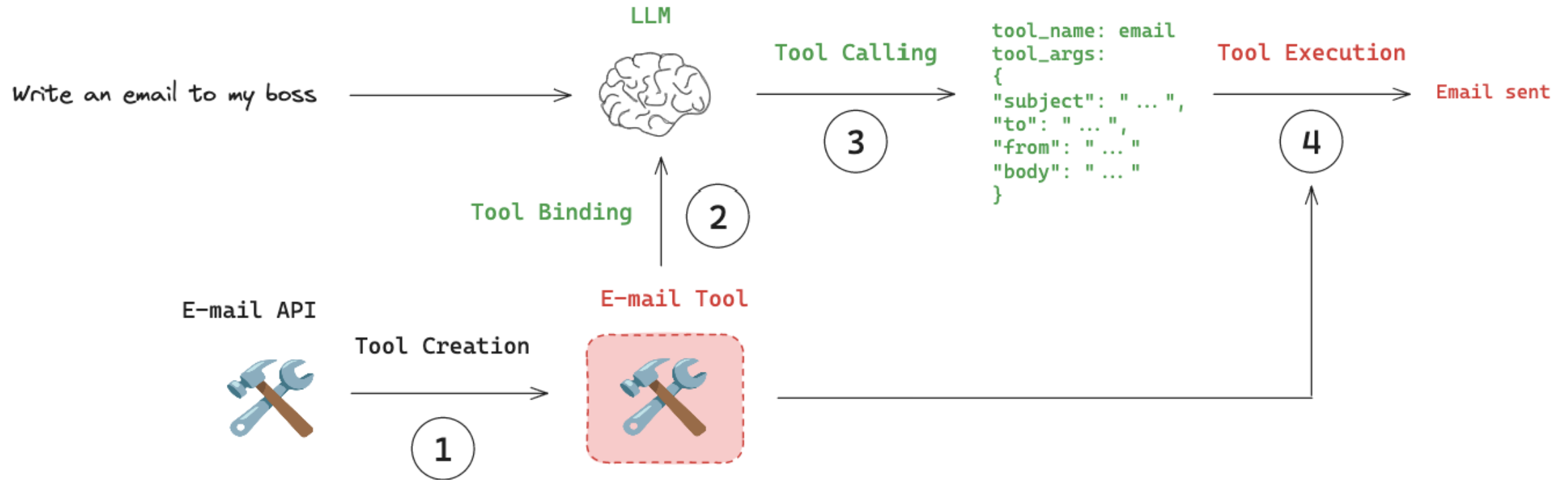
## 4) Tool Use



## 4) Tool Use

- LangChain provides a framework for defining and integrating various types of tools
- Agents can decide when and how to use these tools to achieve their goals.
- They automatically infer the tool's name, description, and expected arguments from the function definition.

## 4) Tool Use



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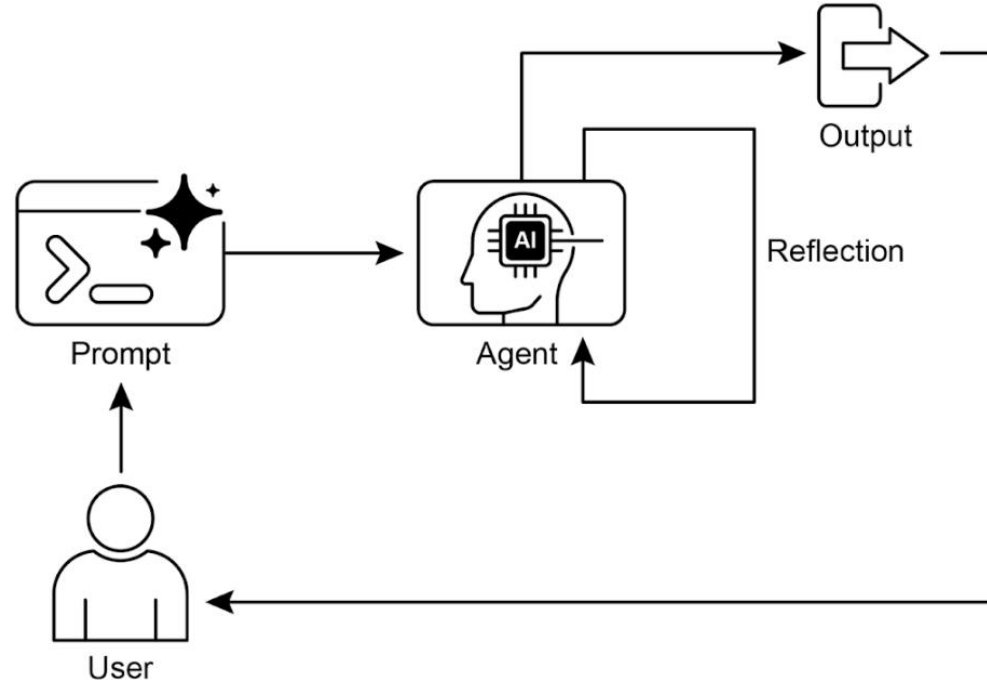
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# 5) Reflection

- Agents can improve their outputs through reflection, a process where they evaluate and refine their work.
- This allows the agent to iteratively improve its response.

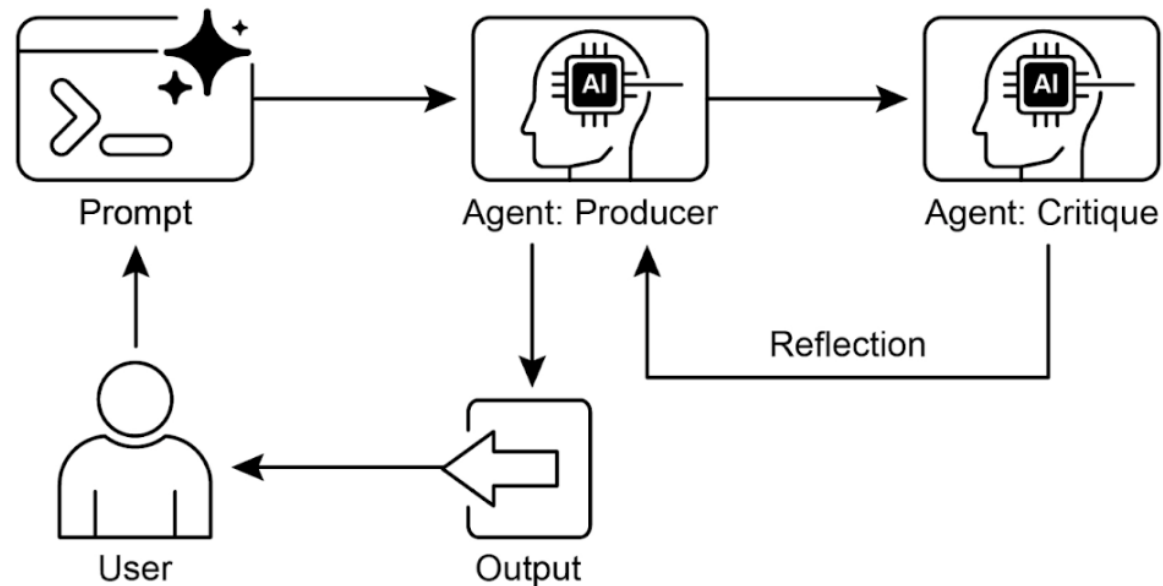
Two main types of reflection:

- **Self-reflection:** agent critiques **its own** generated output.



## 5) Reflection

- **External reflection:** A *separate agent* or component acts as a critic, providing feedback on the initial draft generated by the primary agent.
- External feedback can offer a different perspective.





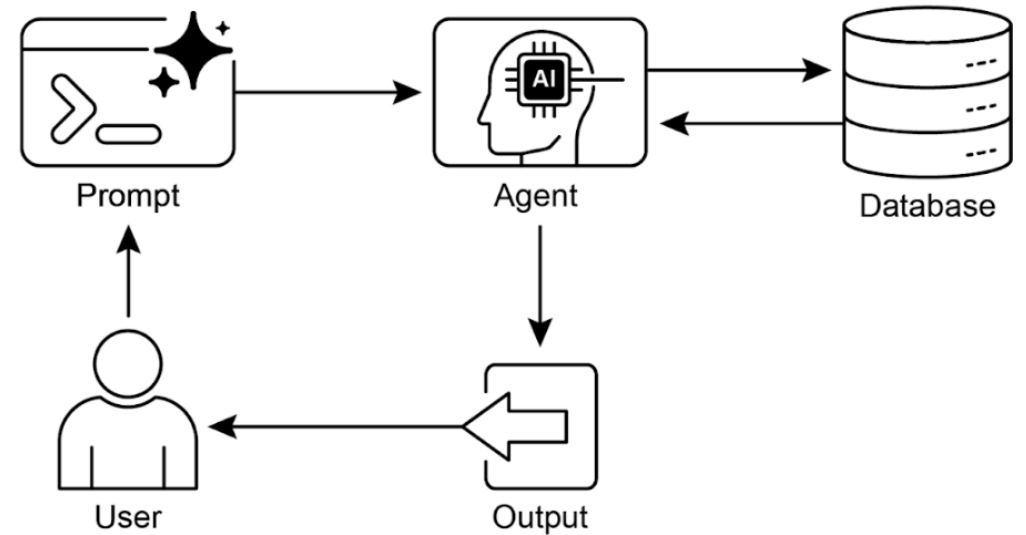
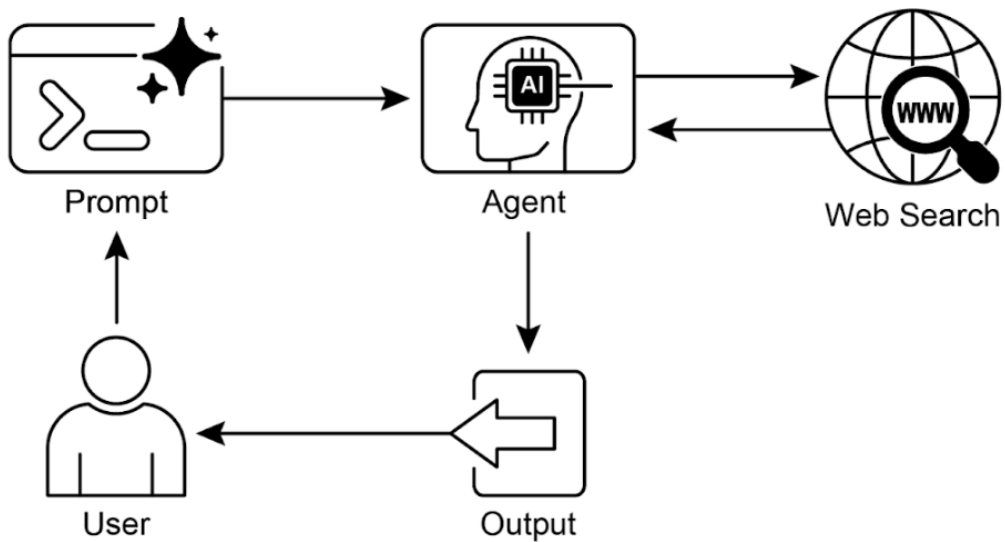
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# 6) Retrieval-Augmented Generation (RAG)

- A technique that combines the power of LLMs with external knowledge sources to generate more accurate and informed responses.
- Instead of relying solely on the knowledge encoded in the model's parameters during training, RAG systems retrieve relevant information from a separate knowledge base
  - Database
  - Collection of documents
  - Internet
- Use this retrieved information as context when generating an answer.
- Reduce the likelihood of:
  - Hallucination
  - Providing outdated information

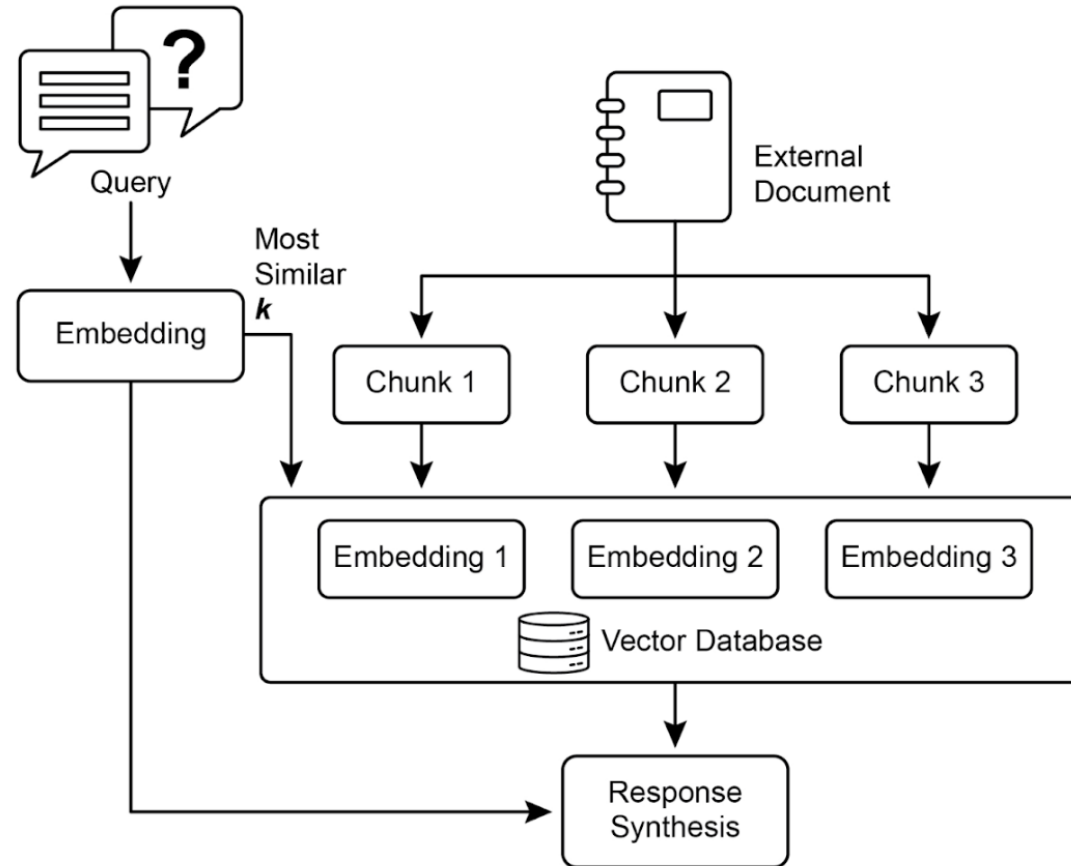
## 6) Retrieval-Augmented Generation (RAG)



# 6) Retrieval-Augmented Generation (RAG)

## RAG Core Concepts:

- Chunking
- Embeddings
- Vector database



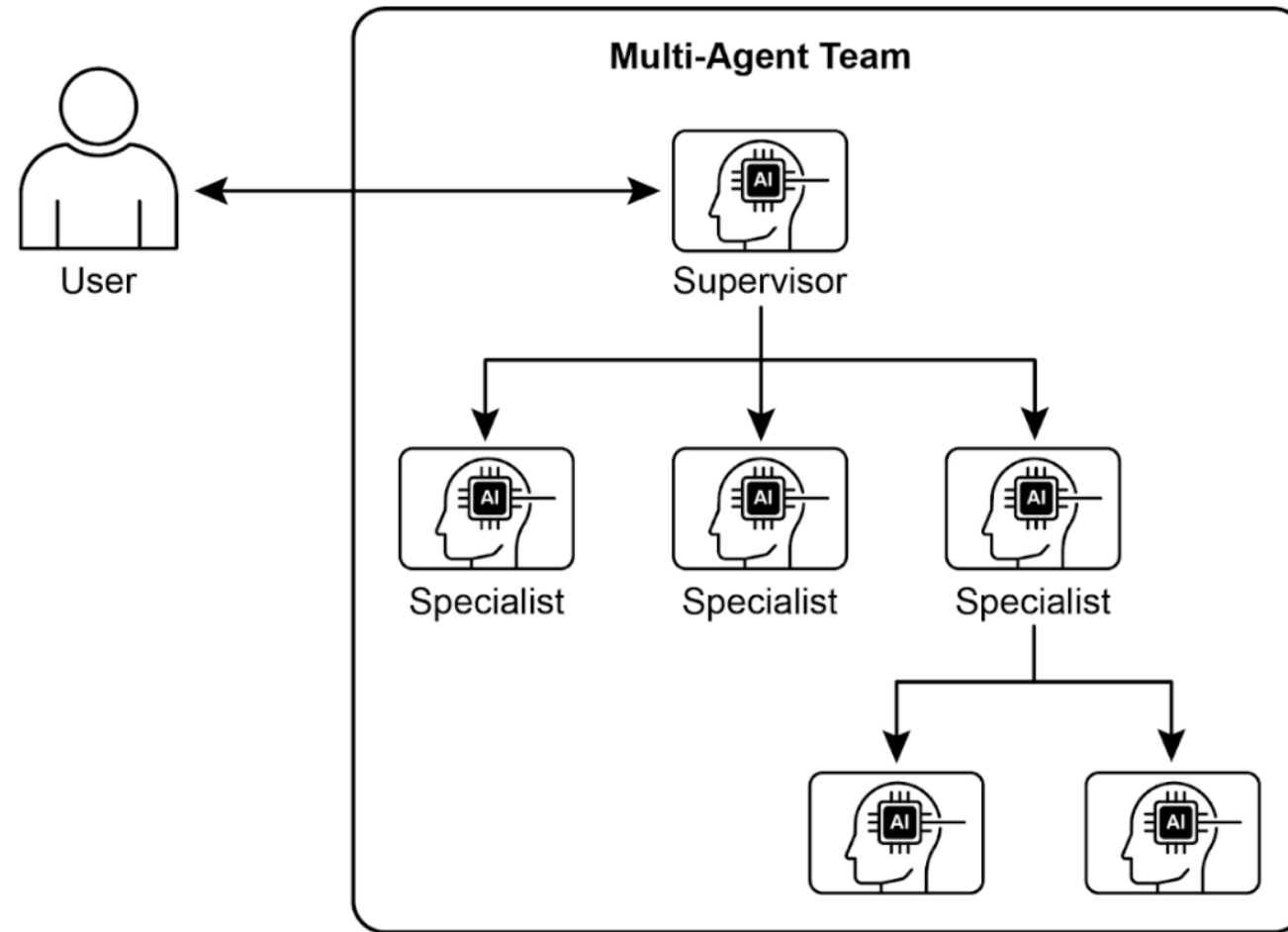
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## 7) Multi-Agent Collaboration

- Multiple independent or semi-independent agents work together to achieve a common goal.
- Interaction and synergy between agents.
- Each agent has a defined role, specific goals, and access to different tools or knowledge bases.

## 7) Multi-Agent Collaboration



# 7) Multi-Agent Collaboration

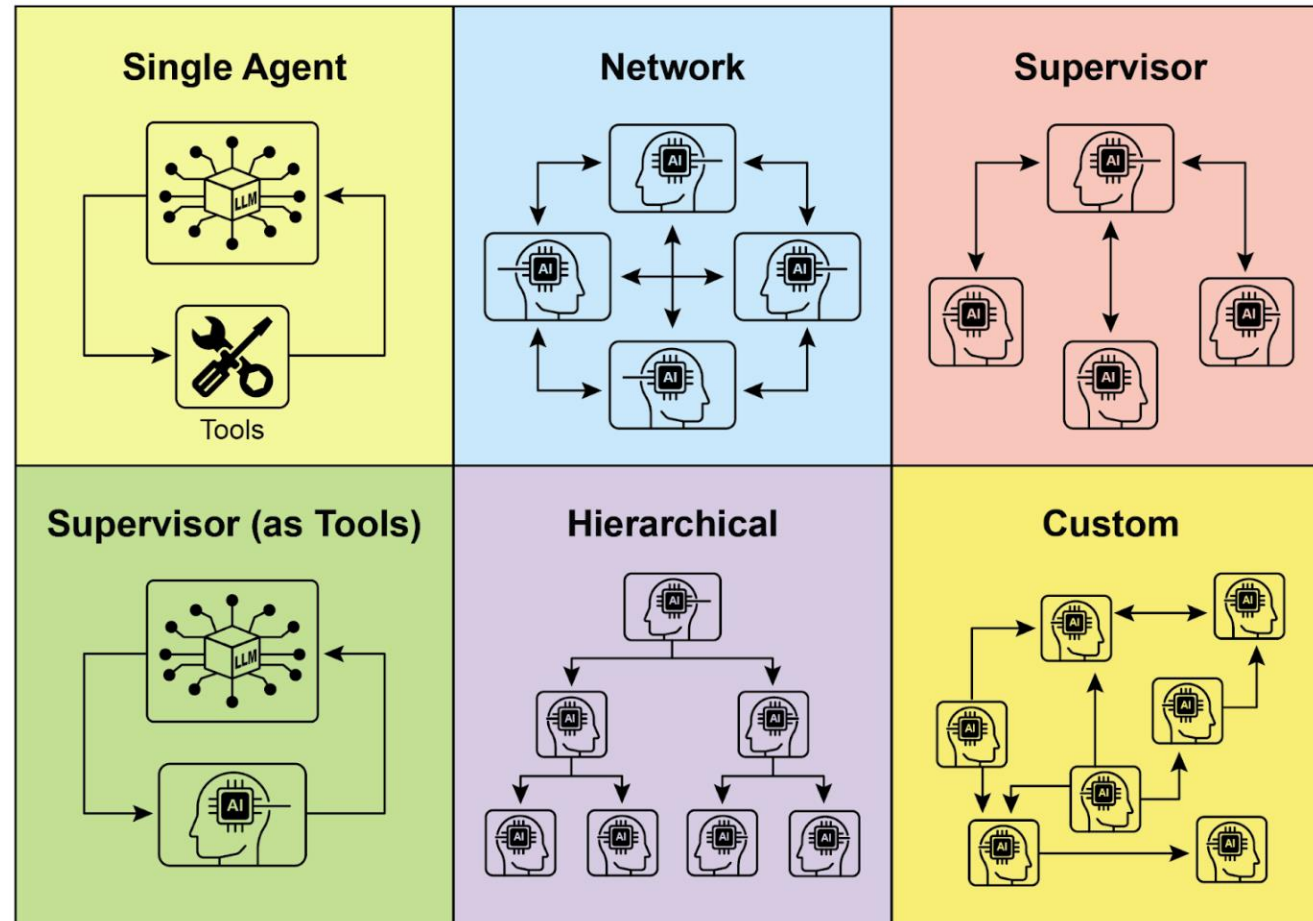


Fig. 2: Agents communicate and interact in various ways.



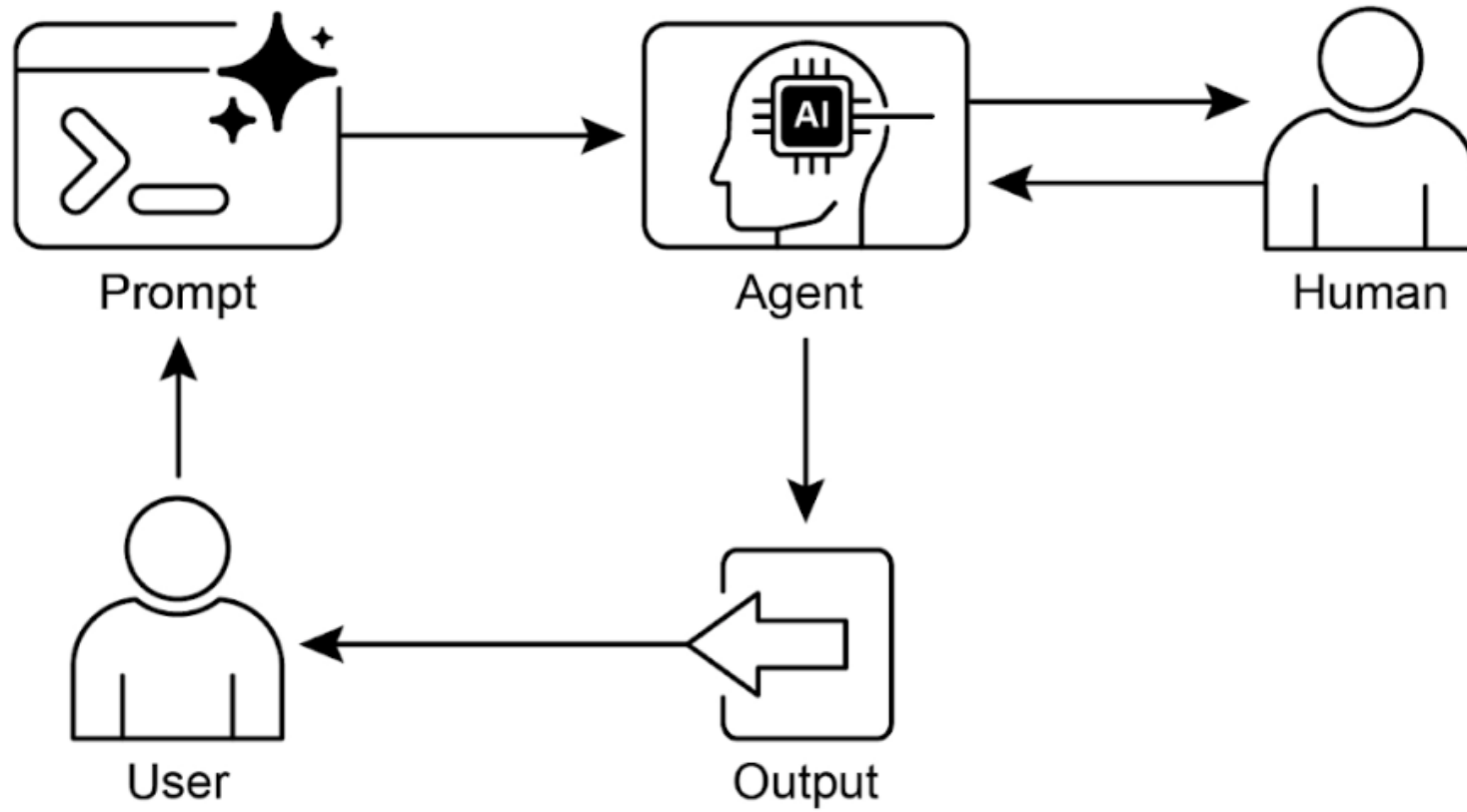
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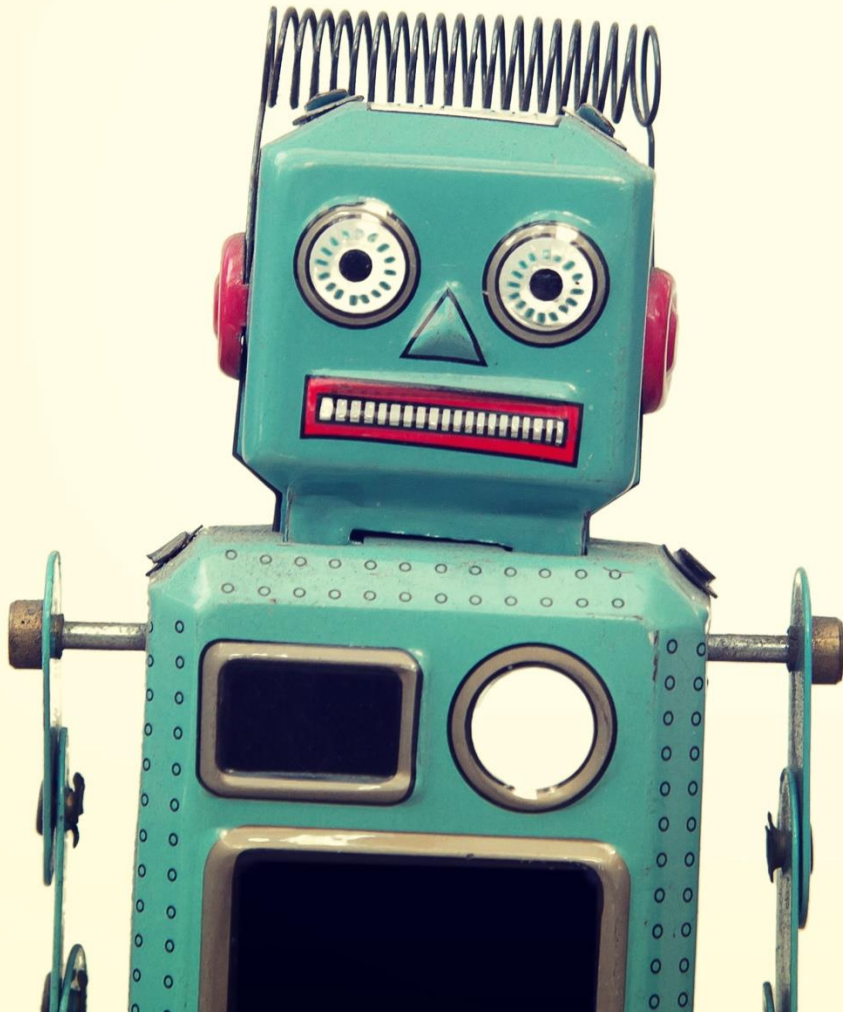
## 8) Human in the Loop

- Integrates AI with human input to enhance agent capabilities.
- Optimal AI performance frequently requires human insight
  - In scenarios with **high complexity** or **ethical** considerations.
- Lack of scalability, operators cannot manage millions of tasks
- Requires a hybrid approach combining automation for scale and HITL for accuracy.
- The effectiveness is heavily dependent on the **expertise** of the human operators.

## 8) Human in the Loop



# Demo: Resume Chatbot



# References

1. [A Hands-On Guide to Building Intelligent Systems](#), Antonio Gulli
2. [LangChain Academy](#) – curated courses & tutorials.  
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3. [LangGraph Cookbook](#)
4. [LangChain Tutorials](#)
5. [Interrupt](#) – the AI Agent Conference by LangChain
6. [Prof. Ghassemi Lectures and Tutorials](#), AI Agents lectures

# Thank you for your attention!


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[https://github.com/maryambrij/  
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