# Getting Started with LangGraph By Lucas Soares

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## **Lucas Soares**

• AI Engineer



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- AI Engineer
- Curious about Building Tools with AI



# **Understanding Agentic Systems**

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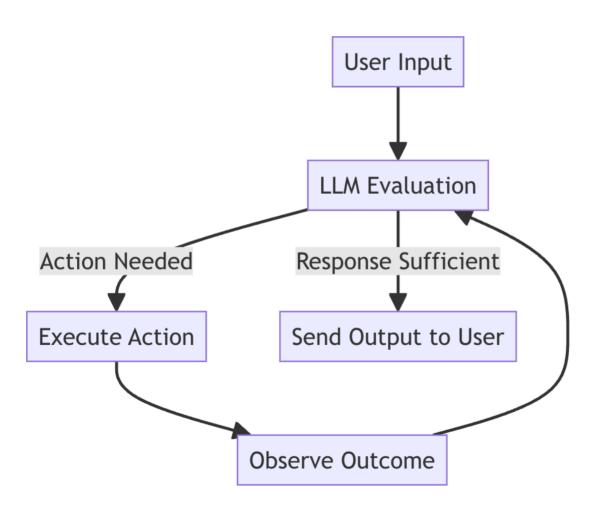
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- **Agentic Systems**: Systems that utilize large language models (LLMs) to manage the control flow of applications.
- **Key Functions**: Routing decisions, tool selection, and evaluating output sufficiency.
- **Agent Loop**: Continuous decision-making process that enables agents to solve complex tasks.

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- **User Input**: Customer asks about order status.
- **LLM Decision**: Determines if it can provide the status directly or if it needs to fetch data from the database.
- Action Taken: If data fetch is needed, the agent queries the database and updates the user with the order status.

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- **Multi-Agent Collaboration**: Specialized LLM agents can collaborate to perform complex tasks.

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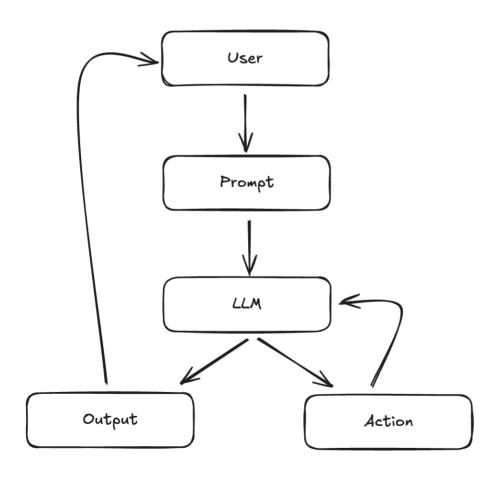
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- 2. **Action Taking**: Executing actions based on LLM outputs.
- 3. **Memory**: Keeping track of interactions for context-aware responses.
- 4. **Planning**: Structuring steps to ensure optimal decision-making.

# Agents as Graphs

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• Workflows built with agents are usually structured as graphs!!



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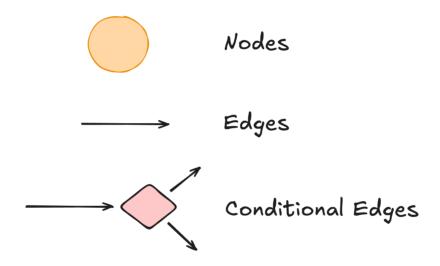
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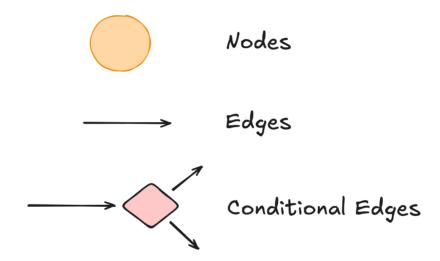
- **Controllability**: Offers low-level control which increases reliability in agentic systems.
- **Human-in-the-Loop**: Built-in persistence layer enhances human-agent interaction patterns.
- **Streaming First**: Supports streaming of events and tokens, providing real-time feedback to users.

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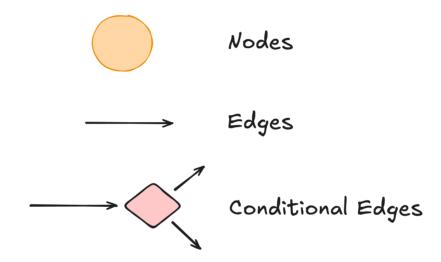


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- Edges/Conditional Edges: Functions that implement fixed/conditional transitions to determine which Node to execute next based on the current State.

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- **States**: Shared data structures that evolve over time as Nodes execute and pass messages along Edges.
- **Message Passing**: Nodes send messages to activate other Nodes, facilitating the execution of workflows in discrete iterations or "supersteps".

# Notebook Demo: Introduction to LangGraph

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- **Functionality**: Each Node is a Python function that processes the current State and outputs an updated State.
- **Execution**: Nodes can run synchronously or asynchronously, and are added to the graph using the add\_node method.
- **Special Nodes**: Includes START and END Nodes to manage the flow of execution in the graph.

# Notebook Demo: Nodes in LangGraph

Edges define the routing logic in LangGraph:

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- **Parallel Execution**: Multiple outgoing edges from a Node can trigger parallel execution of destination Nodes.

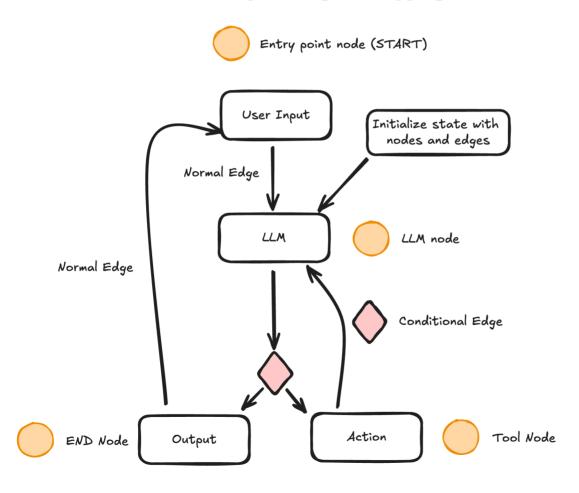
# Notebook Demo: Edges in LangGraph

# Agent Loop in LangGraph

• Outline of an basic agent loop in langgraph:

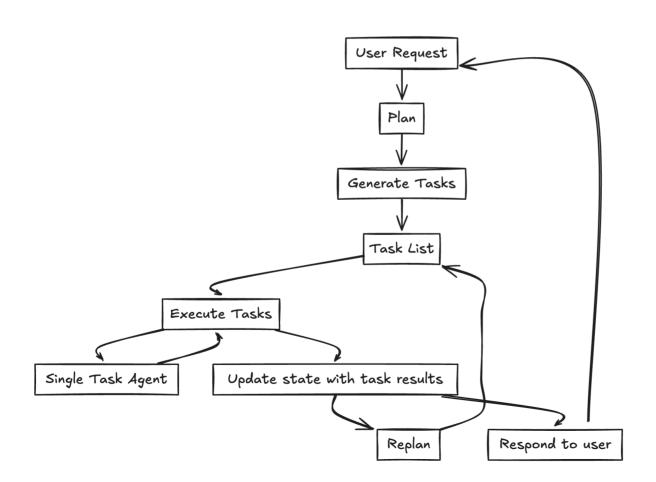
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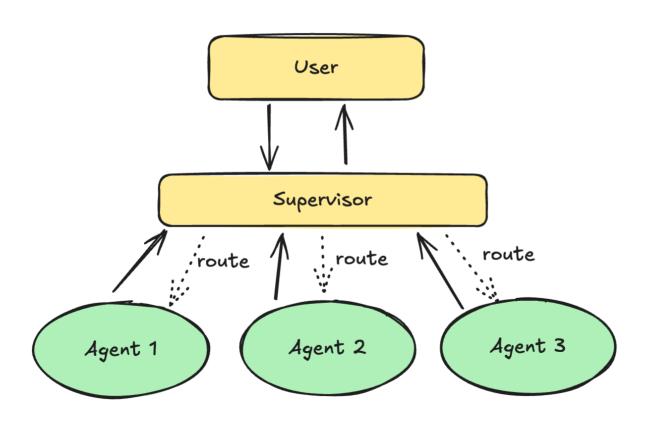


# Notebook Demo: A Basic Research Agent in LangGraph

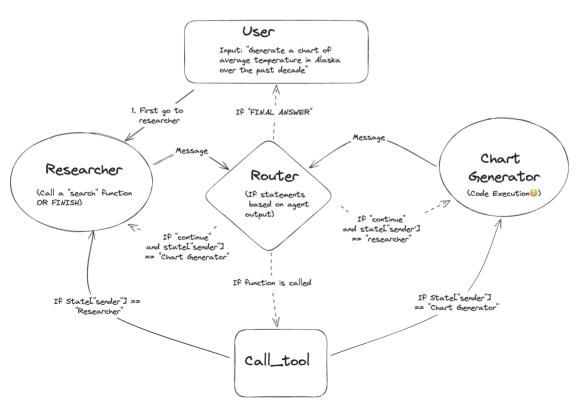
#### Plan & Execute



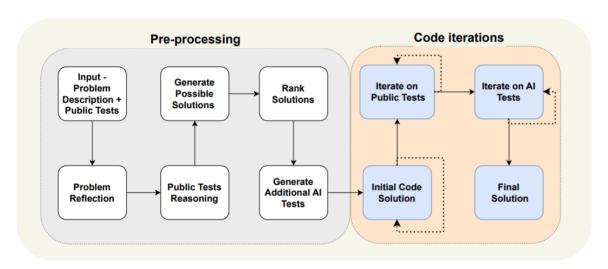
## Supervisor



## Multi-Agent



## Flow Engineering



(a) The proposed AlphaCodium flow.

# Notebook Demo: Self-Reflective RAG Agent with LangGraph