

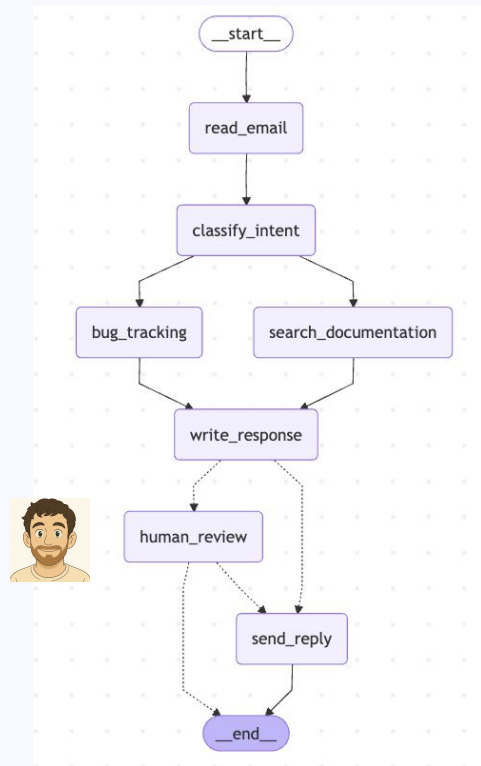


LangChain

LangGraph: StateGraph Essentials

Course Outline

- 1 LangGraph Orientation
- 2 LangGraph Foundations
- 3 Build an Application



LangGraph Orientation

LangGraph

Agents and LLM applications have these challenges

- Latency in the seconds vs ms

<https://blog.langchain.com/building-langgraph>

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 - Streaming – to save perceived latency

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 - Human-in-the-loop - to collaborate with the user
 - Tracing, Observation and Evaluation (LangSmith)

<https://blog.langchain.com/building-langgraph>



Layer Diagram

Your Application

StateGraph SDK

Functional SDK

PregelLoop - Message Passing Runtime

LangGraph

LangChain

Hosted on LangSmith Deployments

Studio



LangSmith

Debugging

Playground

Prompt Management

Annotation

Testing

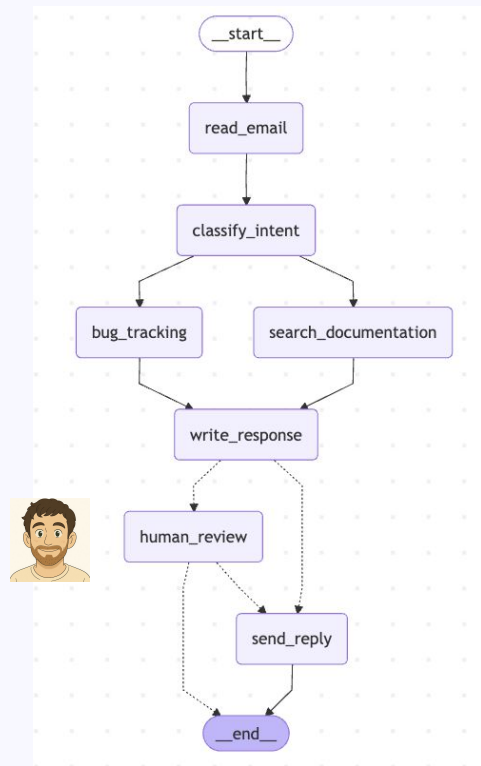
Monitoring

COMMERCIAL



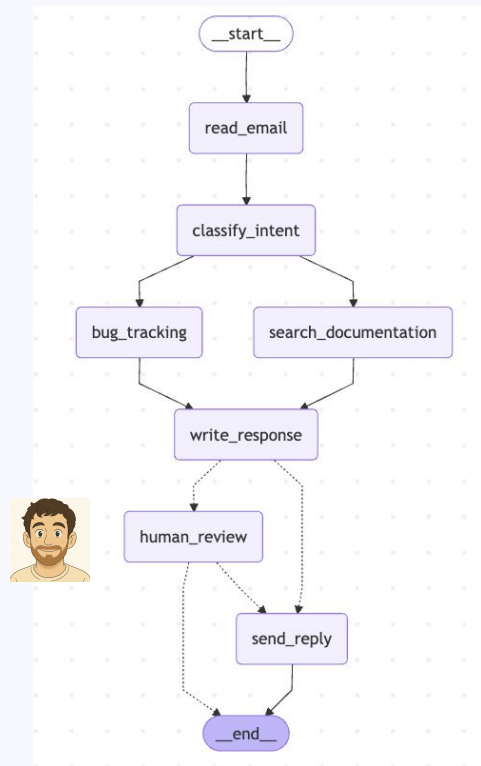
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- 1 LangGraph Orientation
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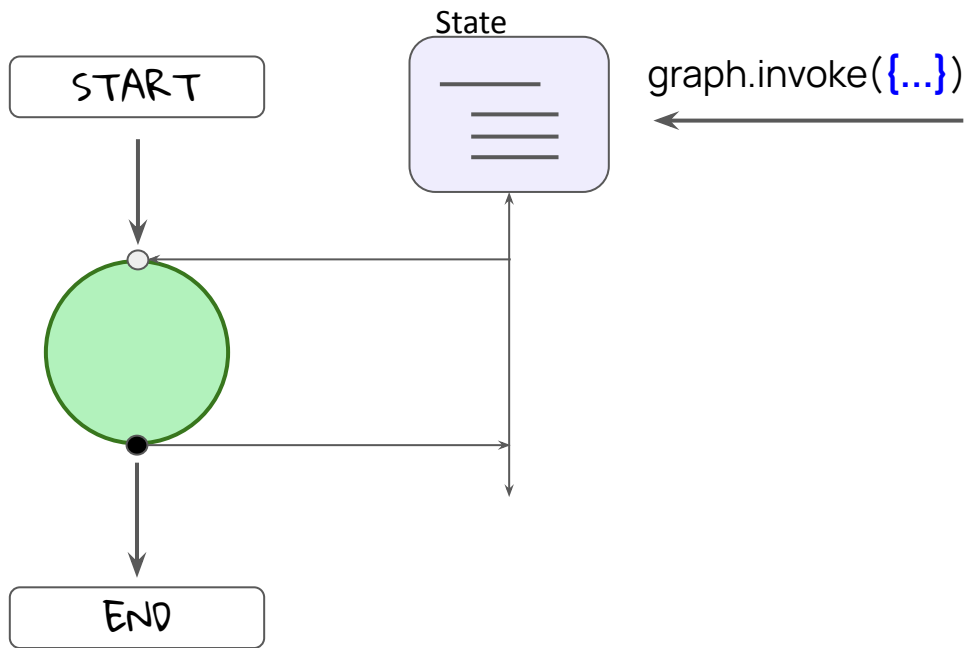
LangGraph: StateGraph

Components and Capabilities

- State: Data
- Node: Functions
- Edges: Control Flow
 - Serial, Parallel
 - Conditional
- Checkpointing/Memory
- Human In the Loop: Interrupts

Nodes and State

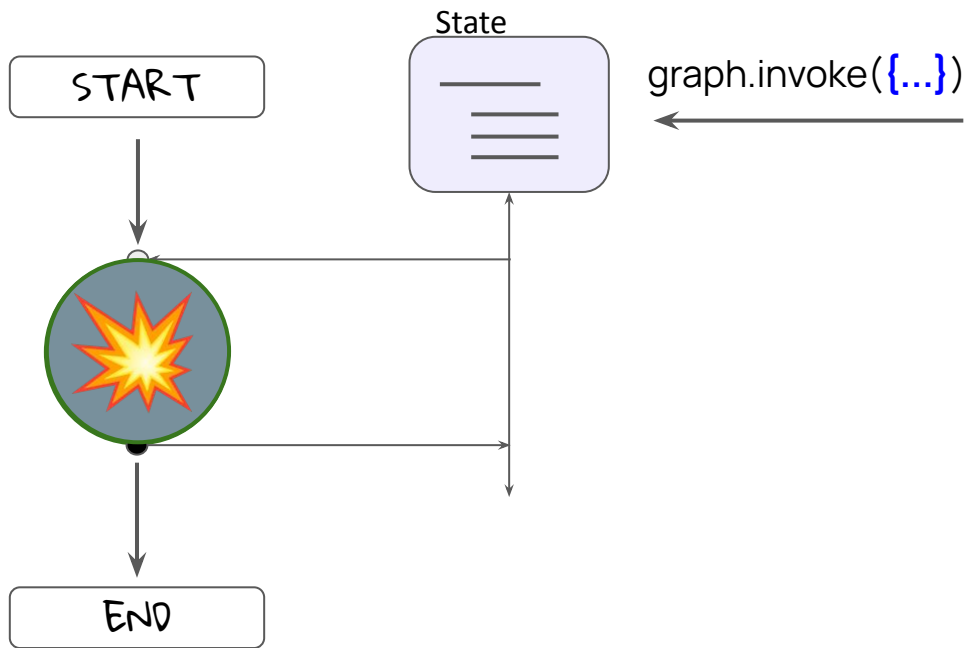
State, Nodes



```
class State(TypedDict):  
    nlist : List[str]
```

```
def node_a(state: State):  
    ...  
    return({"nlist": [note]})
```

State, Nodes



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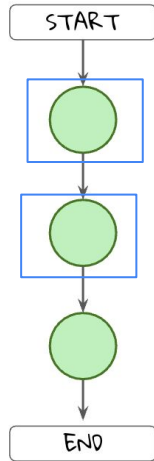
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Edges

Edges: Control Flow

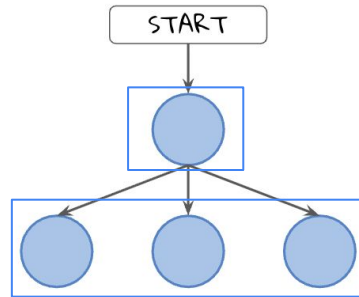
Edge

Serial



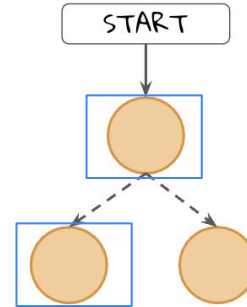
superstep

Parallel

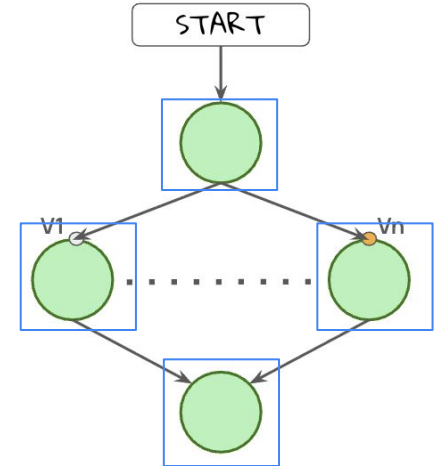


Conditional Edge

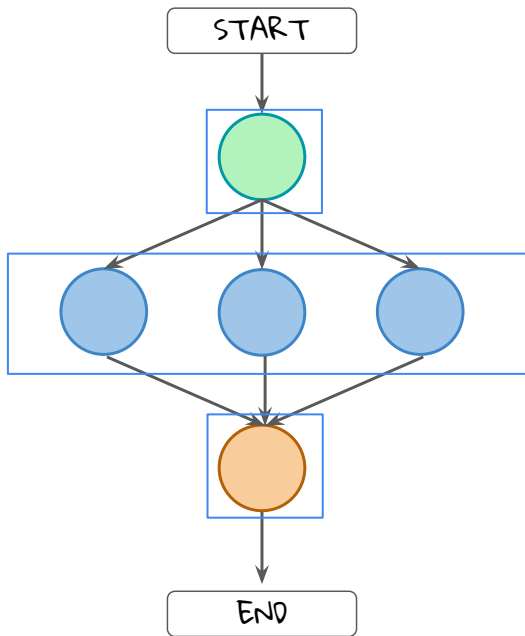
Conditional



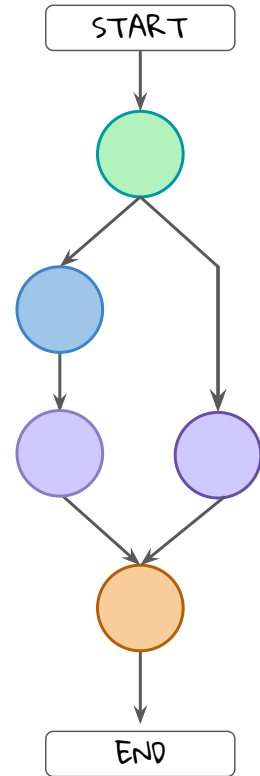
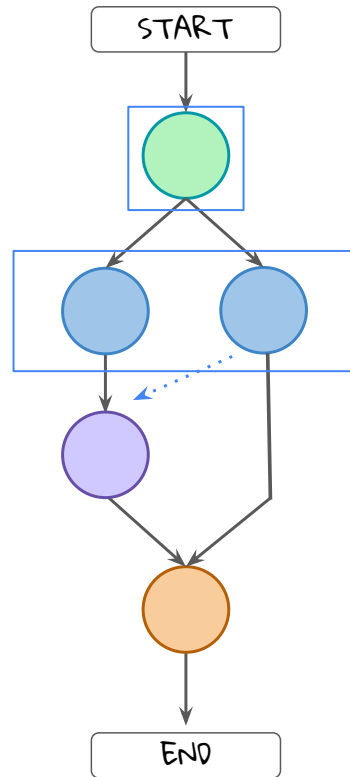
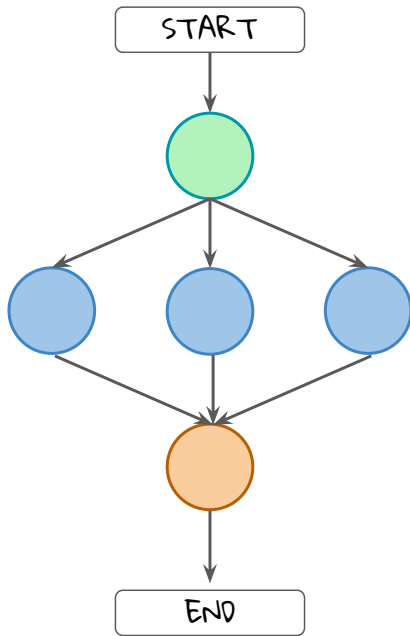
Map-Reduce



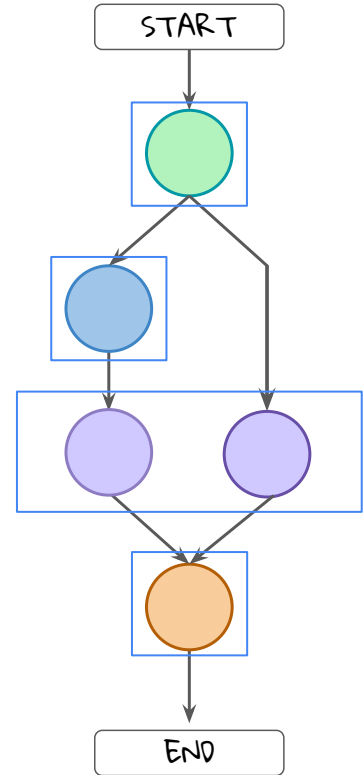
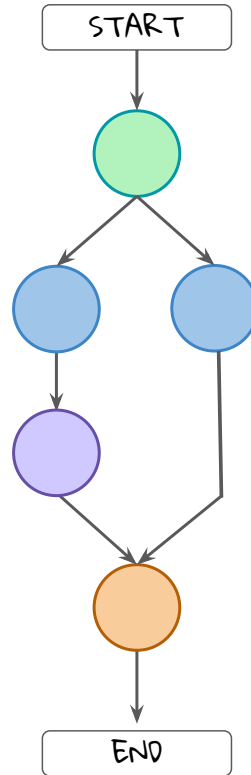
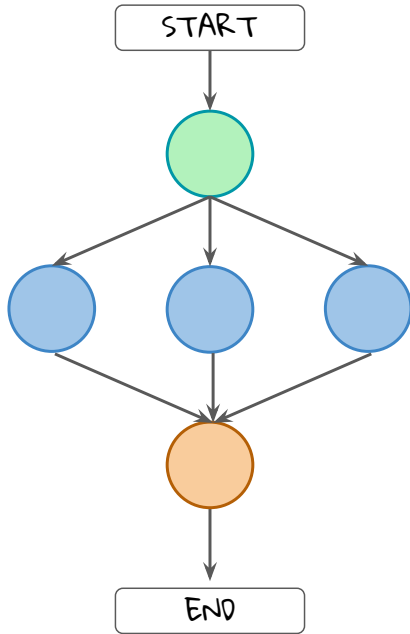
Super Steps



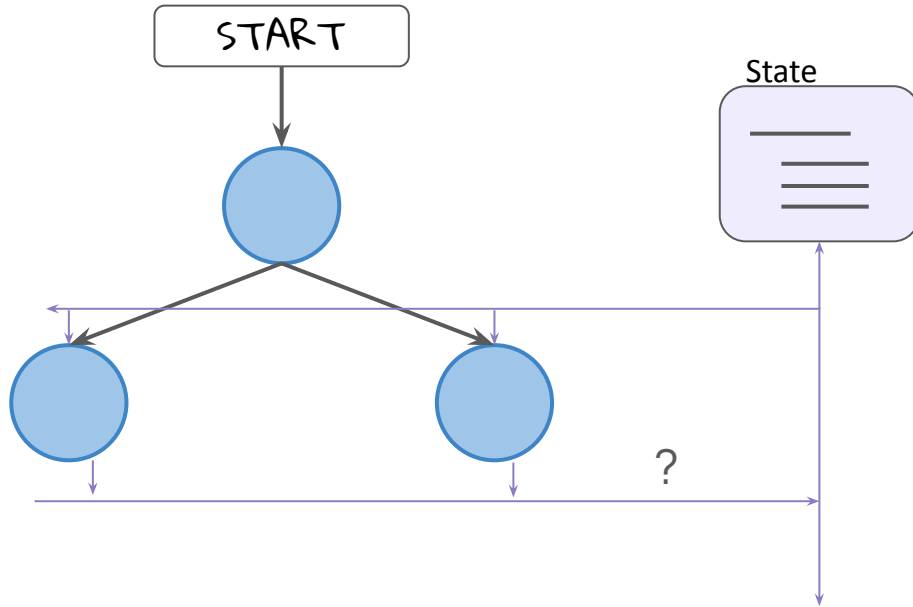
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Super Steps

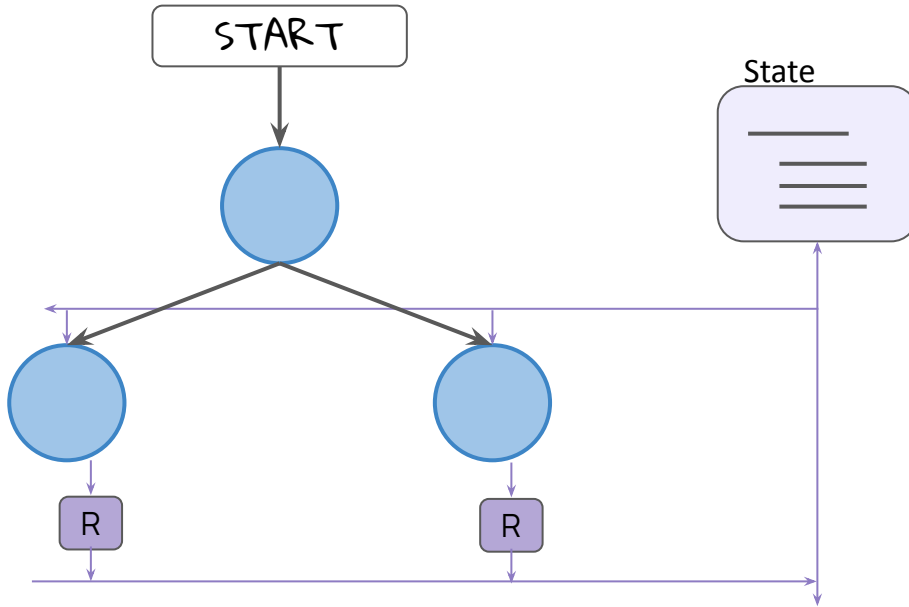


Reducers



```
class State(TypedDict):  
    nlist : List[str]
```


Reducers



```
class State(TypedDict):  
    nlist : Annotated[list[str], operator.add]
```

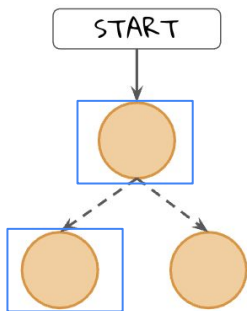
type

reducer function

Conditional Edges

Edges: Control Flow

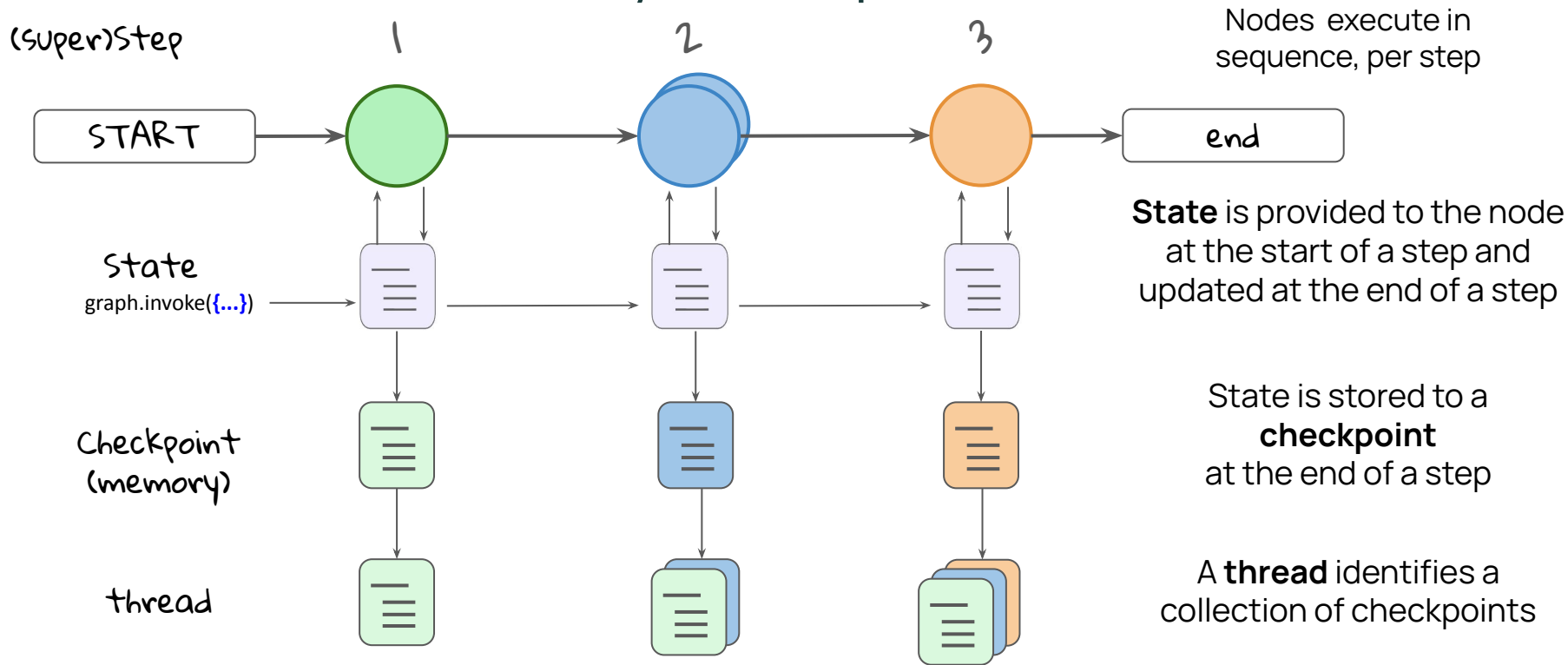
Conditional



- Conditional Edge
- Comman

Memory

Memory / Checkpointers



Memory

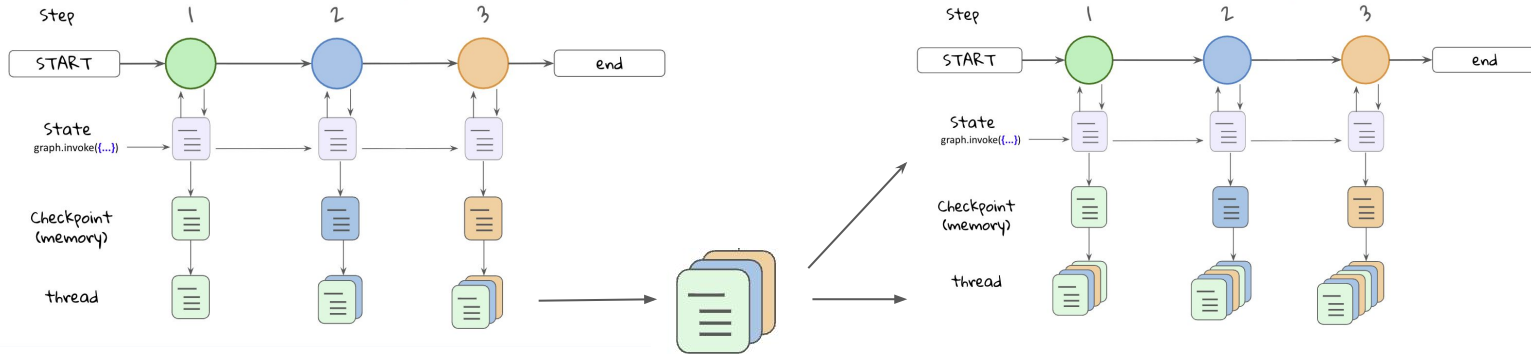
Benefits

- **Recover gracefully from failures** — resume without losing progress.
- **Time travel** — roll back to a known good point and continue forward.

Memory / Checkpointers

Benefits

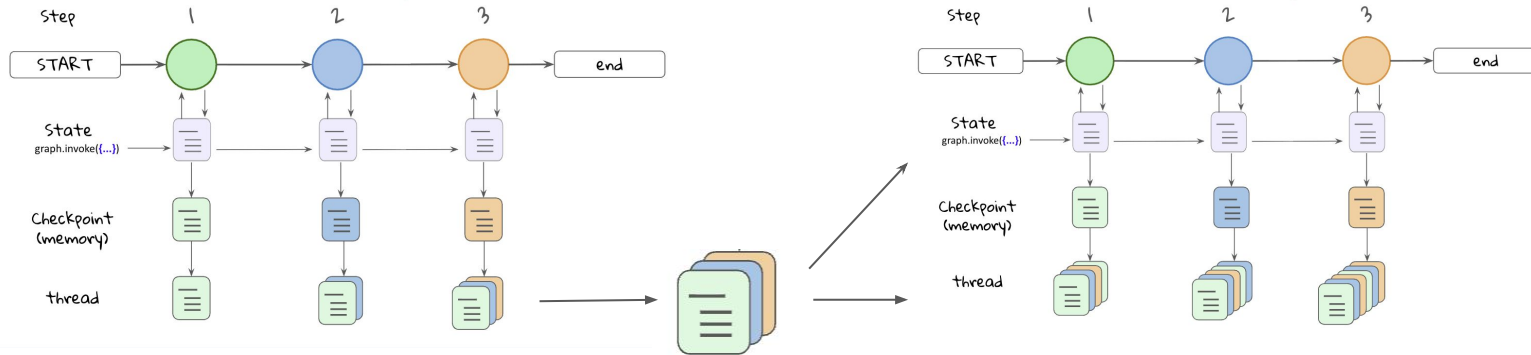
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- **Persistent state** — data is preserved even when the graph is not running.



Memory / Checkpointers

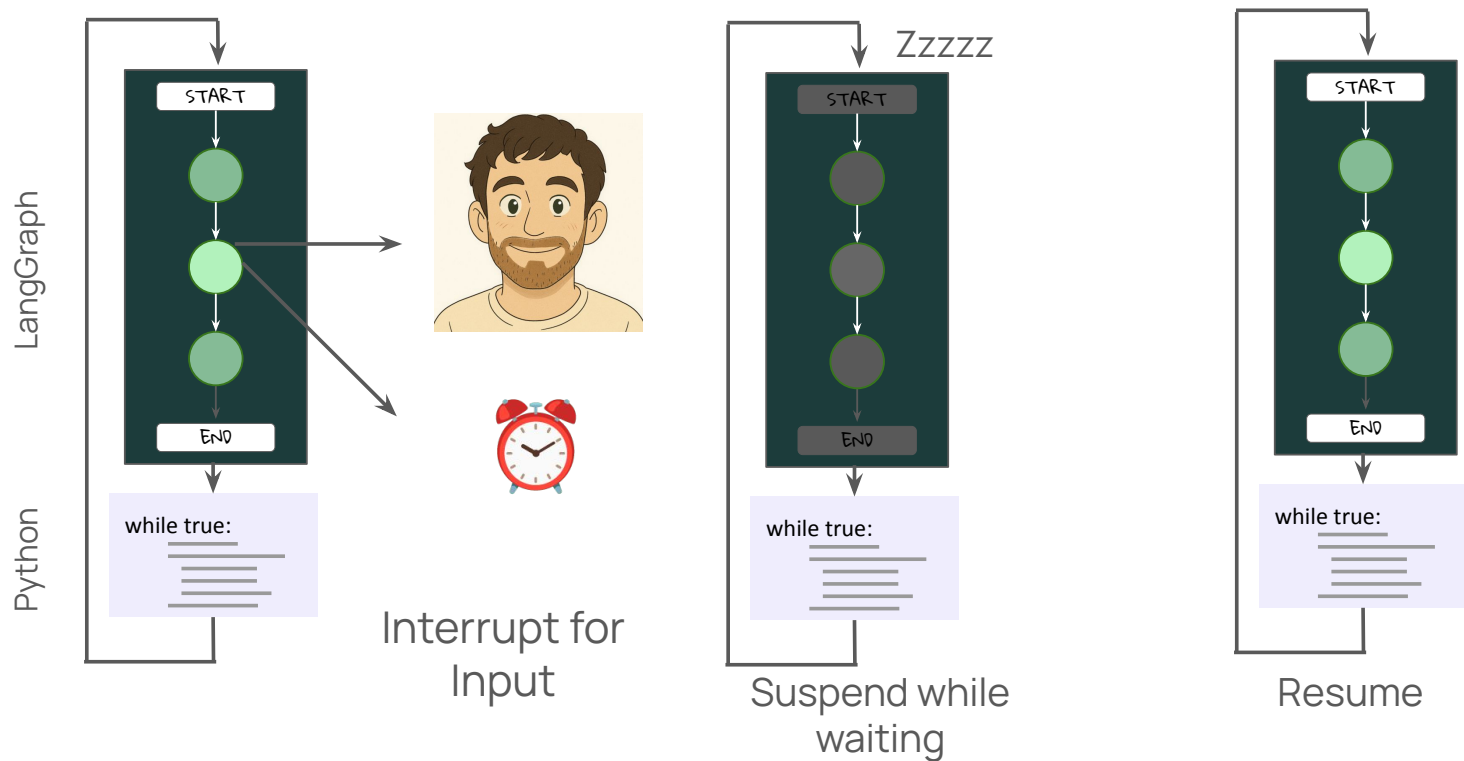
Benefits

- **Recover gracefully from failures** — resume without losing progress.
- **Time travel** — roll back to a known good point and continue forward.
- **Persistent state** — data is preserved even when the graph is not running.
- **Restore state at any step** — pick up execution from where you left off.



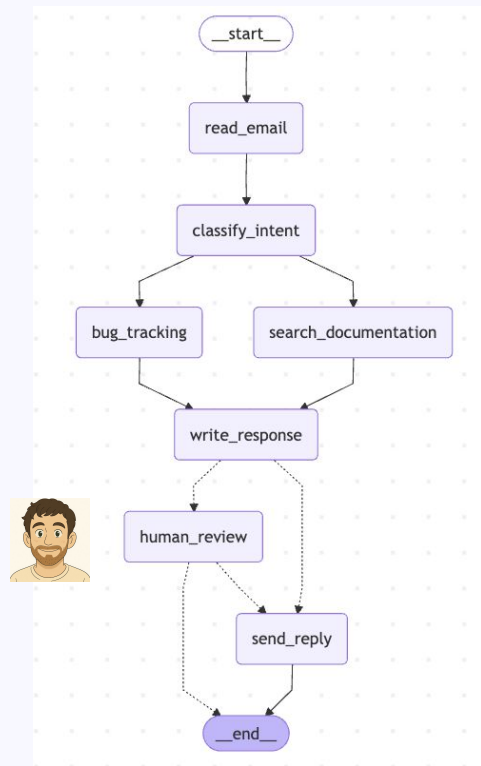
Interrupts

Human In the Loop: Interrupt

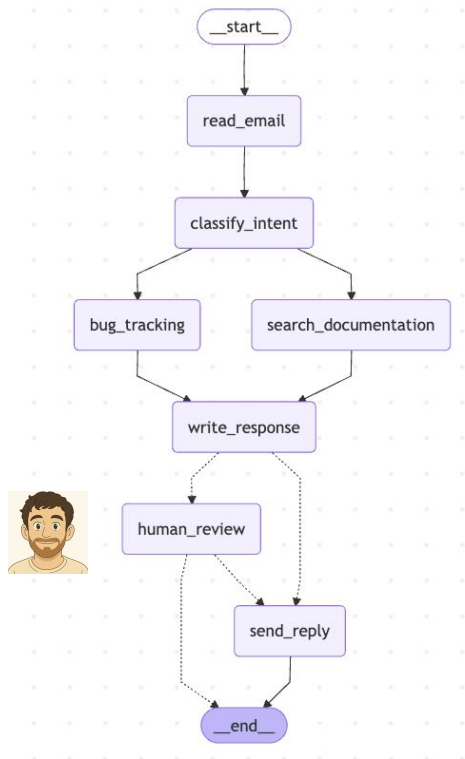


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Email Support Workflow



Simulate a email customer support workflow

Focus on LangGraph aspects:

- State, Nodes,
- Edges
 - Serial
 - Parallel
 - Conditional
- Memory
- Interrupt

Conclusion

Congratulations!

