

Building Context-Aware Reasoning Applications with LangChain

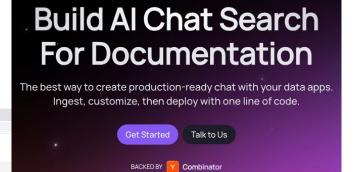
what's the weather like in San Francisco right now?

I'm sorry but I currently do not have the capability to access real-time data, \$

information for San Francisco, or any other location, I'd recommend checking a reliable weather website or using a weather app on your mobile device.

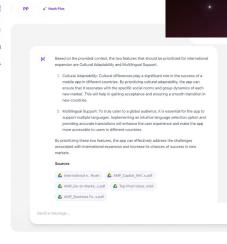
including current weather information. If you want up-to-date weather

If you have any other questions or need assistance with a different topic, please let me know!

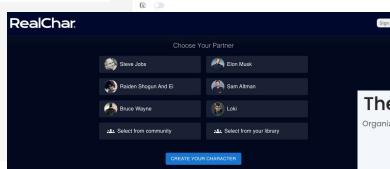


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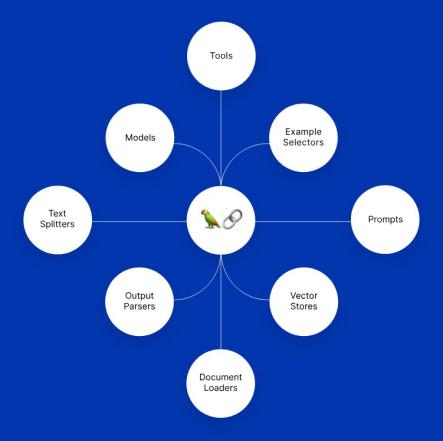
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Context-aware



4 Types of Context-Awareness

- 1. Instruction Prompting
- 2. Few Shot Examples
- 3. RAG
- 4. Fine-tuning

1. Instruction Prompting

Be a helpful chatbot Respond in a friendly tone Only talk about LangChain

2. Few Shot Examples

Hi // Hi. I'm a LangChain Chatbot.

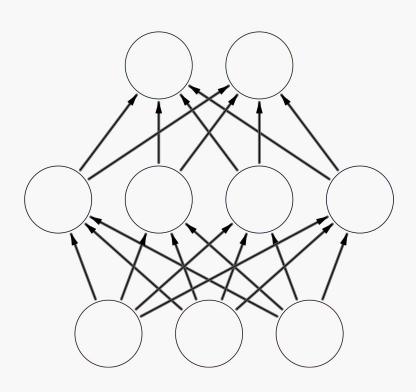
Can you tell me a joke? // I only help with LangChain

3. Retrieval Augmented Generation (RAG)

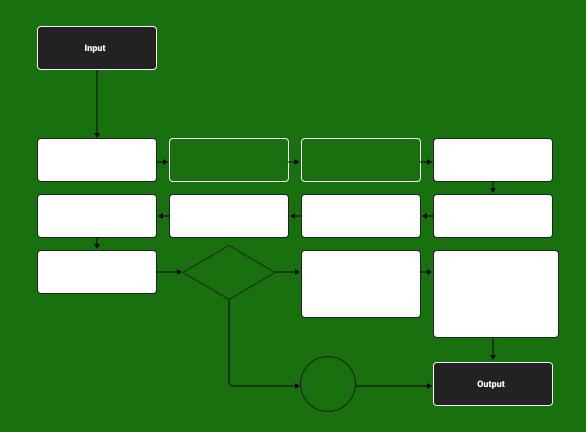
answer the user question based only on the context below

LangChain is a framework for developing applications powered by large language models. It enables context-aware, reasoning applications. The main value props of LangChain are components and off-the-shelf chains. Components are abstractions for working with large language models, along with a collection of implementations for each abstraction. Components are modular and easy-to-use, whether you are using the rest of the LangChain framework or not. Off-the-shelf chains are structured assemblies of components for accomplishing higher-level tasks.

4. Fine-tuning



Reasoning



5 Levels of Reasoning

- 1. LLM Call
- 2. Chaining
- 3. Routing
- 4. Automatons / State Machines
- 5. Autonomous Agents

1. LLM Call



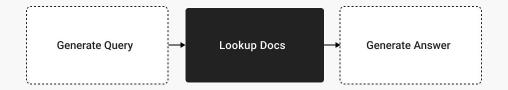


Llama 2 A Falcon 180B

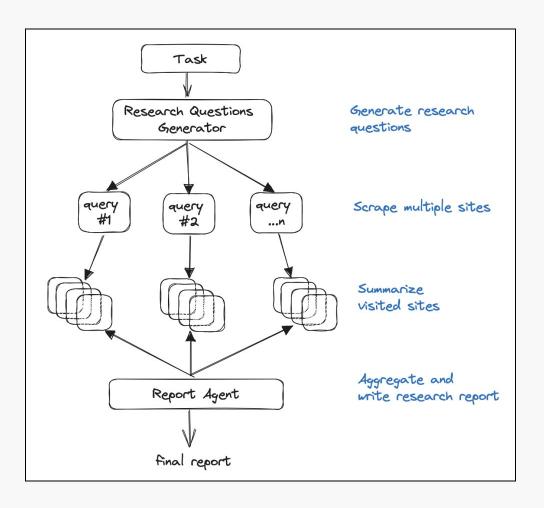




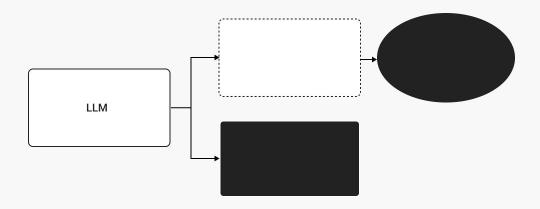
2. Chaining



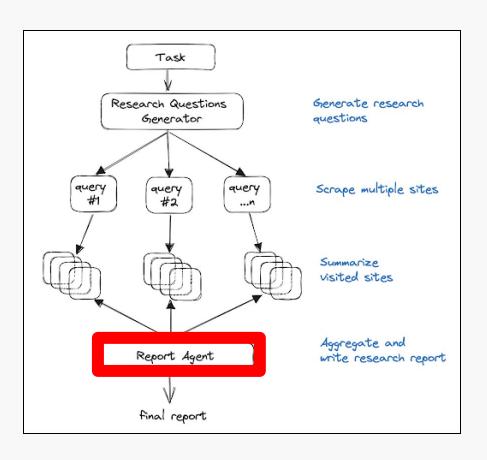
GPT Reseacher



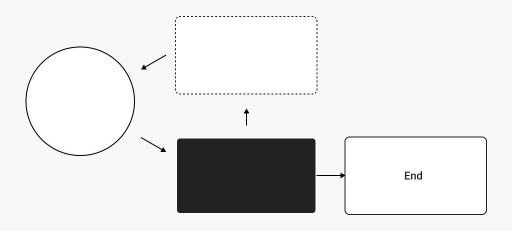
3. Routing



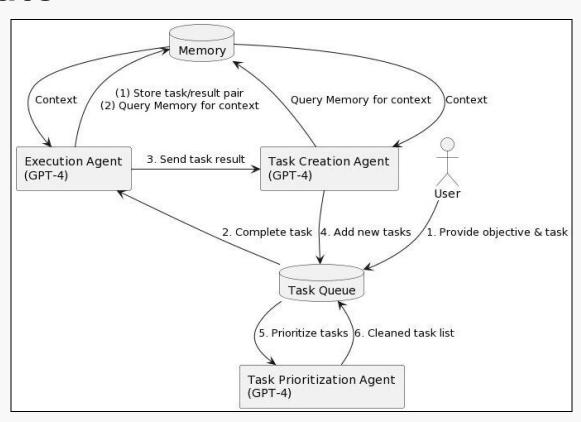
GPT Reseacher



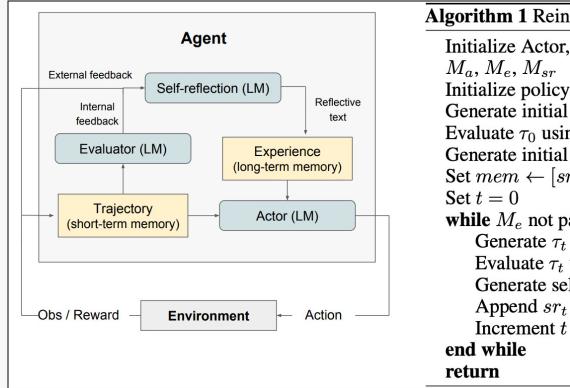
4. Automatons / State Machines



Plan-and-Execute



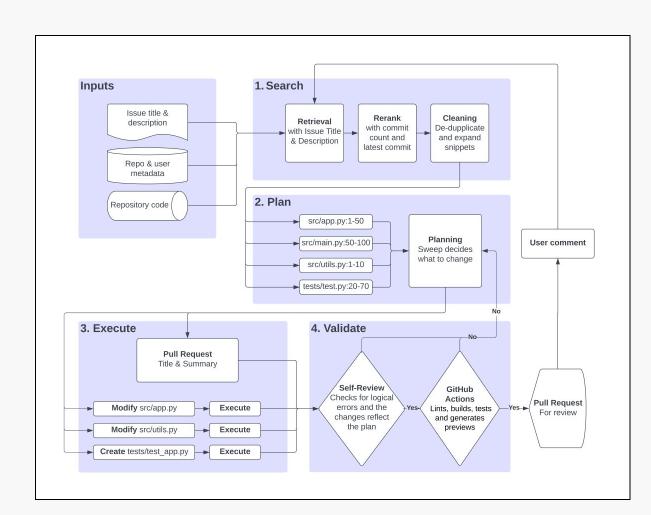
Reflexion



Algorithm 1 Reinforcement via self-reflection

```
Initialize Actor, Evaluator, Self-Reflection:
Initialize policy \pi_{\theta}(a_i|s_i), \theta = \{M_a, mem\}
Generate initial trajectory using \pi_{\theta}
Evaluate \tau_0 using M_e
Generate initial self-reflection sr_0 using M_{sr}
Set mem \leftarrow [sr_0]
while M_e not pass or t < \max trials do
    Generate \tau_t = [a_0, o_0, \dots a_i, o_i] using \pi_{\theta}
    Evaluate \tau_t using M_e
    Generate self-reflection sr_t using M_{sr}
    Append sr_t to mem
    Increment t
```

Sweep.dev



5. Autonomous Agents



Levels of autonomy in LLM applications



code



LLMs

			Output Is	to Take	Transitions
HUMAN-DRIVEN	1	Code		>	<u> </u>
	2	LLM Call		one step only	<u> </u>
	3	Chain		multiple steps	<u> </u>
	4	Router		L	no cycles
AGENT-EXECUTED	5	State Machine	L		cycles
	6	Agent			

Decide What

Decide What Steps

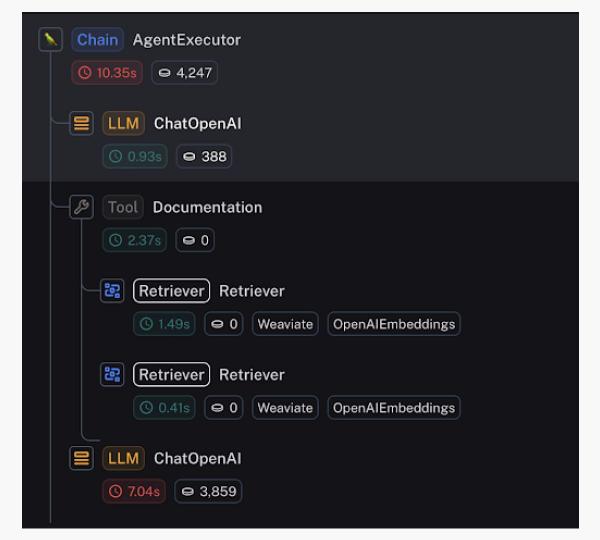
Define State



Challenges

- Orchestration
- Data Engineering
- Prompt Engineering
- Debugging
- Evaluation
- Collaboration

Orchestration



Data Engineering

```
□ Copy
INPUT
     query: build agent
                                                     Copy
OUTPUT
     documents:
       - metadata:
           source:
     https://python.langchain.com/docs/use_cases/more/agen
     ts/agents/
         page_content: ⊢
           Modify Agent • The built-in LangChain agent
     types are designed to work well in
           generic situations, but you may be able to
     improve performance by modifying
           the agent implementation. There are several
     ways you could do this: Modify the
           base prompt. This can be used to give the
     agent more context on how it should
           hohava atc Madify the autnut nameen This is
```

Prompt Engineering

Prompt

TEMPLATE

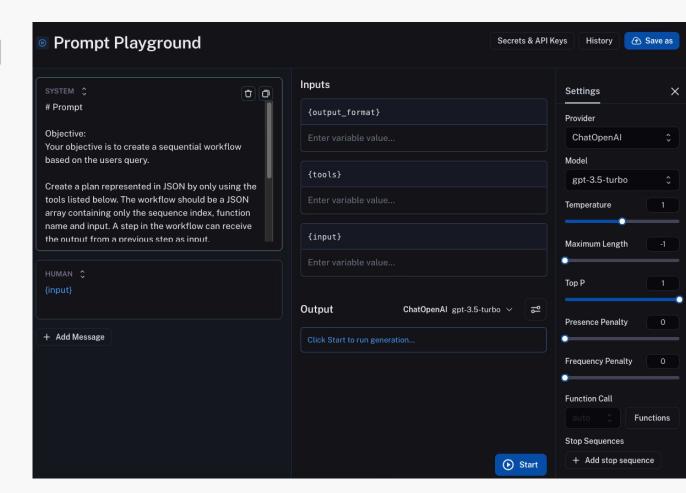


- Your goal is to create a well crafted set of answers for a test for a specific question.
- Your answers will be used on a test to access a students knowledge. You will be given the question and your goal is to follow the output format below w/ guidelines

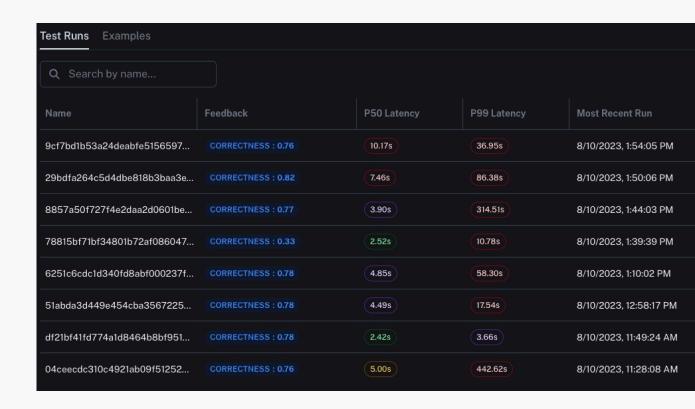
3

- 4 Answer Choice Guidelines
- Position the correct option so that it appears approximately the same number of times in each possible position for a guiz
- Answer choices should be written clearly and similarly to each other in content, length, and grammar; avoid giving clues through the use of faulty grammatical construction.
- 7 Make all distractors plausible; they should be common misconceptions that learners may have.
 - In answer choices, avoid "all of the above" and "none of the above," which can lead to artificially higher levels of

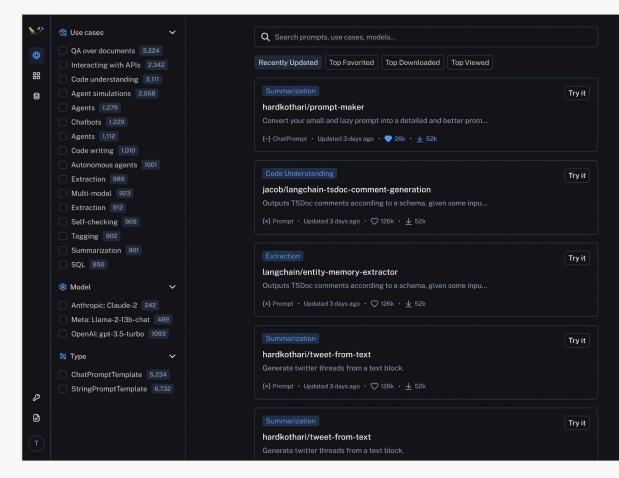
Debugging



Evaluation



Collaboration





Thank you!