### LangGraph vs LangChain vs LangFlow vs LangSmith: Which One To Use & Why?



#### 70.256 views Aug 21, 2024

Are you curious about the differences between LangChain, Langflow, Langgraph, and Langsmith? In this video, I'll explain what is LangChain, what is Langflow, what is Langgraph, and what is Langsmith, breaking down each tool's features, use cases, and why they were created.

We'll explore what Langsmith is used for and how it integrates with LangChain for monitoring and debugging workflows. Learn how Langgraph simplifies agent management and discover the differences in Langchain vs Langgraph and Langgraph vs Langchain for building Al applications. I also cover the Langchain vs Langflow comparison, showing how Langflow's visual interface can speed up prototyping for non-coders.

If you're wondering about Langchain vs Langsmith, this video also breaks down how Langsmith helps in tracking and optimizing your applications. Whether you're a developer or just starting with AI workflows, this video will clear up the confusion around these tools, helping you understand which one to choose for your next project.

#### ★ Timestamps:

00:00 Intro

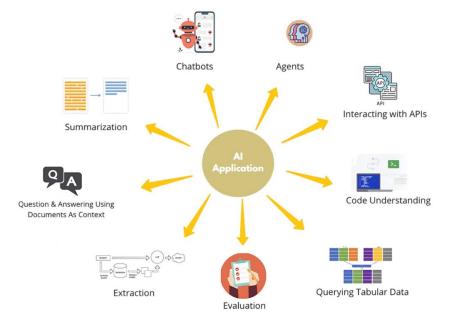
00:40 LangChain (What, Why, How)

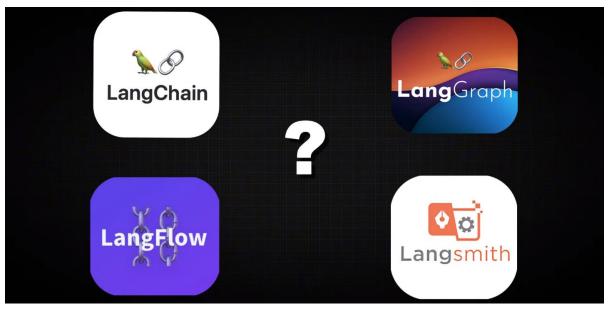
03:40 LangGraph (What, Why, How)

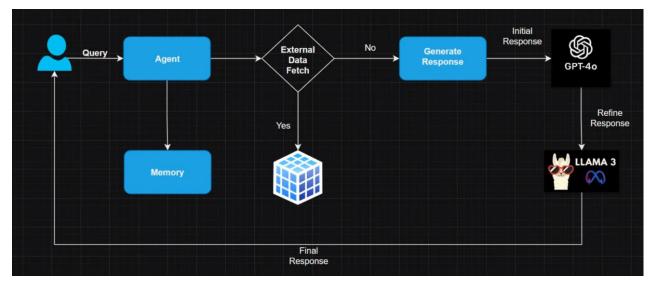
05:37 LangFlow (What, Why, How)

07:11 LangSmith (What, Why, How)

14:10 What's Next?



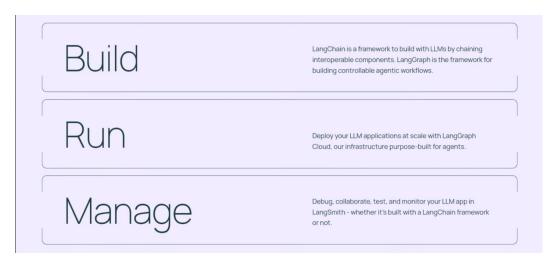


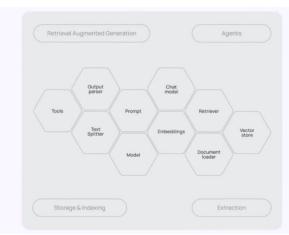


```
import openai
import requests
                                         \downarrow
def query_gpt4(prompt):
   openai.api_key = "your-gpt4-api-key"
    response = openai.Completion.create(
        engine="gpt-4",
        prompt=prompt,
        max_tokens=100
   return response.choices[0].text
def query_llama3(prompt):
    url = "https://llama3-api-endpoint"
   headers = {
    data = {
        "prompt": prompt,
    response = requests.post(url, headers=headers, json=data)
    return response.json()["response"]
```

```
_llama3(prompt):
    headers = {
    data = {
        "prompt": prompt,
    response = requests.post(url, headers=headers, json=data)
    return response.json()["response"]
# Simulating Memory (you might use a file or database in reality)
                                       (\Lambda)
memory = {}
def update_memory(user_input, response):
   memory["last_interaction"] = {"input": user_input, "response": response}
    return memory.get("last_interaction")
# Agent function: Decides which model to use
def agent(query):
        "prompt": prompt,
    response = requests.post(url, headers=headers, json=data)
    return response.json()["response"]
# Simulating Memory (you might use a file or database in reality)
                                       ( \downarrow )
memory = {}
# Manually managing memory
def update_memory(user_input, response):
   memory["last_interaction"] = {"input": user_input, "response": response}
    return memory.get("last_interaction")
def agent(query):
    # Retrieve memory
    last_interaction = get_memory()
    # Logic for using GPT-4 or Llama 3 based on the query
    if "complex" in query:
        response = query_gpt4(query)
        response = query_llama3(query)
```

You can write the code to manually do all the above things as above or you can use a framework like LangGraph







## Build your app with LangChain

Build context-aware, reasoning applications with LangChain's flexible framework that leverages your company's data and APIs. Future-proof your application by making vendor optionality part of your LLM infrastructure design.

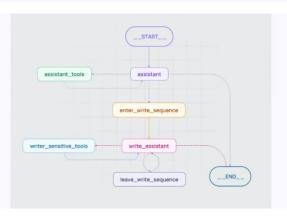
Learn more about LangChain 7

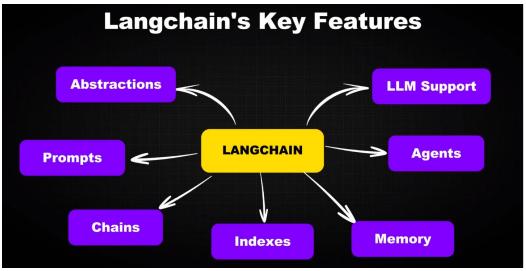


## Run at scale with LangGraph Cloud

Deploy your LangGraph app with LangGraph Cloud for fault-tolerant scalability - including support for async background jobs, built-in persistence, and distributed task queues.

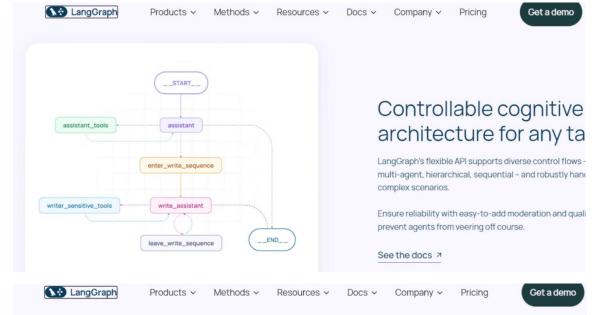
Learn more about LangGraph ¬





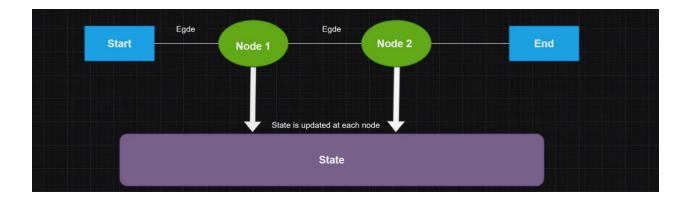
```
from langchain import OpenAI, LlamaCpp, Memory, Agent
from langchain.chains import SequentialChain
# Step 1: Define the models
gpt4_llm = OpenAI(api_key="your-gpt4-api-key", model="gpt-4")
1lama3_llm = LlamaCpp(model_path="path-to-your-llama3-model")
memory = Memory()
# Step 3: Create an agent that decides whether to use GPT-4 or Llama 3
def agent_fn(query, memory):
    last_interaction = memory.load()
    # Simple agent decision: Use GPT-4 for complex queries
    if "complex" in query:
        return gpt4_llm(query)
        return llama3_llm(query)
agent = Agent(agent_fn, memory=memory)
chain = SequentialChain([agent], memory=
                                         memory)
```





# Deploy agents at scale, monitor carefully, iterate boldly

With LangGraph Cloud, quickly deploy and scale your application, with infrastructure purpose-built for agents. Now in beta for all users on LangSmith Plus and Enterprise plans.



### Installation

```
pip install -U langgraph
```

### Example

One of the central concepts of LangGraph is state. Each graph execution creates a state that is passed between nodes in the graph as they execute, and each node updates this internal state with its return value after it executes. The way that the graph updates its internal state is defined by either the type of graph chosen or a custom function.

Let's take a look at a simple example of an agent that can use a search tool.

```
export ANTHROPIC_API_KEY=sk-...

Optionally, we can set up LangSmith for best-in-class observability.

export LANGSMITH_TRACING=true
export LANGSMITH_API_KEY=lsv2_sk_...

from typing import Annotated, Literal, TypedDict

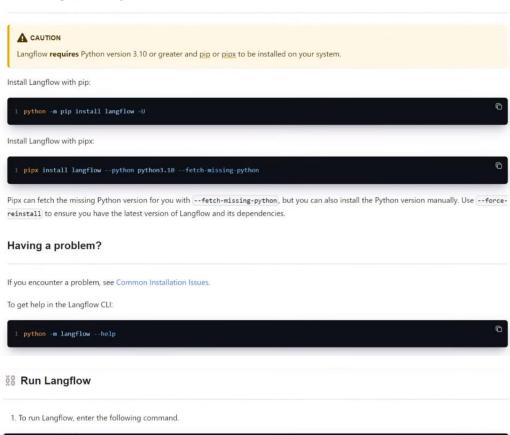
from langchain_core.messages import HumanMessage
from langchain_anthropic import ChatAnthropic
from langchain_core.tools import tool
from langgraph.checkpoint.memory import MemorySaver
from langgraph.graph import END, StateGraph, MessagesState
from langgraph.prebuilt import ToolNode
```

```
\ensuremath{\text{\#}} Define the tools for the agent to use
@tool
def search(query: str):
     """Call to surf the web."""
    # This is a placeholder, but don't tell the LLM that..
    if "sf" in query.lower() or "san francisco" in query.lower():
       return "It's 60 degrees and foggy."
    return "It's 90 degrees and sunny.
tools = [search]
tool_node = ToolNode(tools)
\verb|model| = \verb|ChatAnthropic(model="claude-3-5-sonnet-20240620", temperature=0|).bind\_tools(tools)|
# Define the function that determines whether to continue or not
def should_continue(state: MessagesState) -> Literal["tools", END]:
   messages = state['messages']
    last_message = messages[-1]
    # If the LLM makes a tool call, then we route to the "tools" node
  if last_message.tool_calls:
```



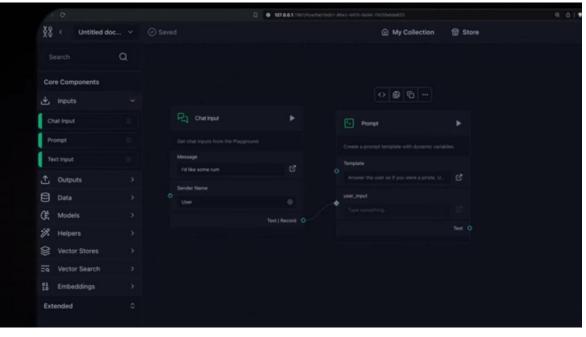
### **Install Langflow Locally**

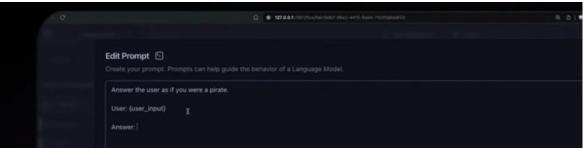
1 python -m langflow run

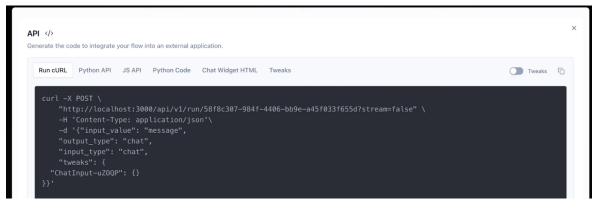


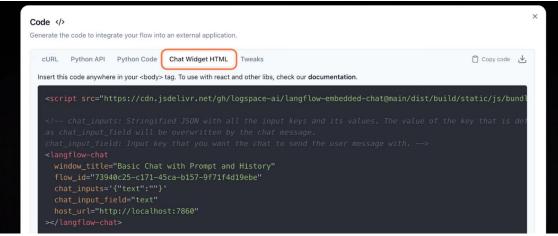
2. Confirm that a local Langflow instance starts by visiting http://127.0.0.1:7860 in a Chromium-based browser.











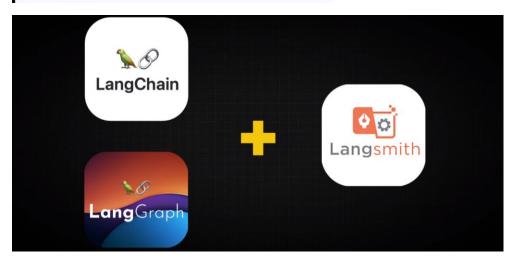


# The platform for your LLM development lifecycle

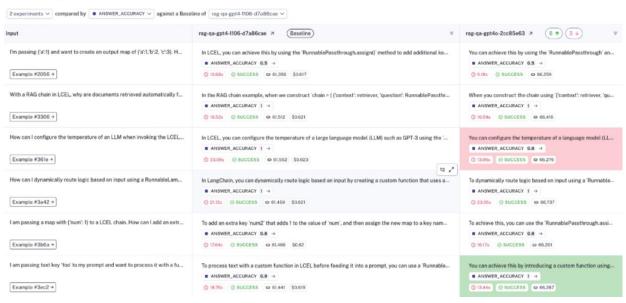
LLM-apps are powerful, but have peculiar characteristics. The non-determinism, coupled with unpredictable, natural language inputs, make for countless ways the system can fall short. Traditional engineering best practices need to be re-imagined for working with LLMs, and LangSmith supports all phases of the development lifecycle.

Sign Up ↗





### **Comparing Experiments**



## Get started with LangSmith

LangSmith is a platform for building production-grade LLM applications. It allows you to closely monitor and evaluate your application, so you can ship quickly and with confidence. Use of LangChain is not necessary - LangSmith works on its own!

### 1. Install LangSmith

```
Python TypeScript

pip install -U langsmith
```

### 2. Create an API key

To create an API key head to the Settings page. Then click Create API Key.

### 3. Set up your environment

```
export LANGCHAIN_TRACING_V2=true
export LANGCHAIN_API_KEY=<your-api-key>

# The below examples use the OpenAI API, though it's not necessary in general
export OPENAI_API_KEY=<your-openai-api-key>
```

### 4. Log your first trace

### TRACING TO LANGSMITH FOR LANGCHAIN USERS

There is no need to use the LangSmith SDK directly if your application is built entirely on LangChain (either Python and JS).

We've outlined a tracing guide specifically for LangChain users here.

We provide multiple ways to log traces to LangSmith. Below, we'll highlight how to use traceable. See more on the Annotate code for tracing page.

