Getting Started with LangChain

Lucas Soares

21-02-2024

• The presentation will be organized into the following structure:

- The presentation will be organized into the following structure:
 - Presentation Block

- The presentation will be organized into the following structure:
 - Presentation Block
 - Notebook Demo

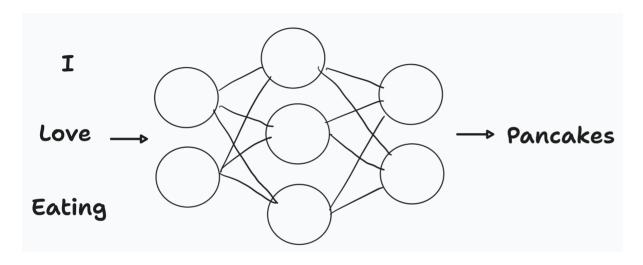
- The presentation will be organized into the following structure:
 - Presentation Block
 - Notebook Demo
 - Quick Q&A + Summary

- The presentation will be organized into the following structure:
 - Presentation Block
 - Notebook Demo
 - Quick Q&A + Summary
 - **Optional Exercise** During Q&A (for those that don't have questions and want to try something out)

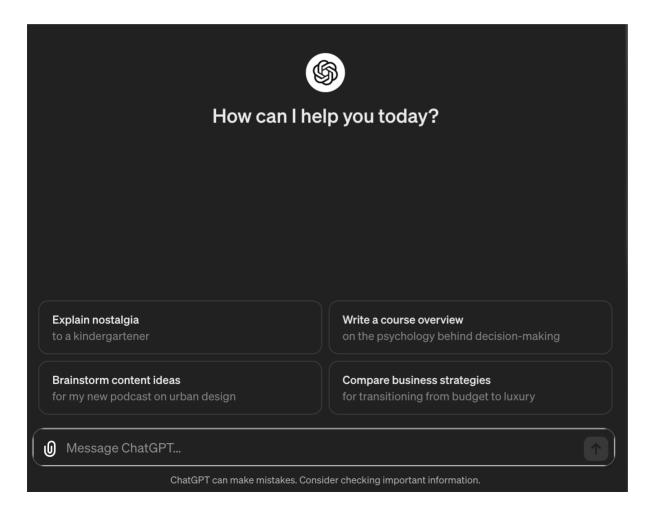
- The presentation will be organized into the following structure:
 - Presentation Block
 - Notebook Demo
 - Quick Q&A + Summary
 - Optional Exercise During Q&A (for those that don't have questions and want to try something out)
- Repeat

Large Language Models

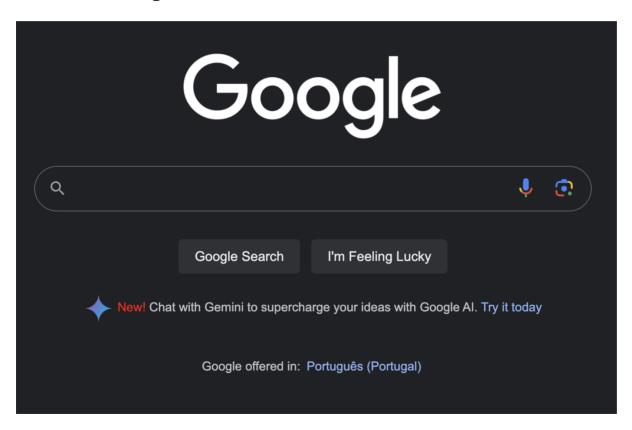
Large Language Models Predict the Next Word



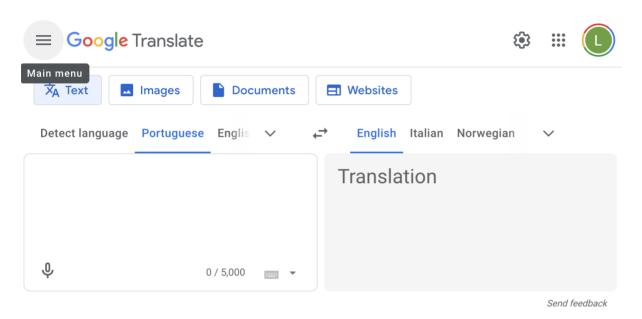
Conversational ChatBots



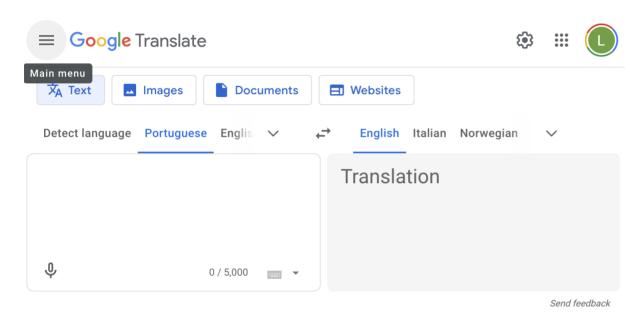
• Search Engines



Translation



Translation



• And so much more from Q&A over PDFs to personalized tutoring.





• LangChain is a framework that facilitates creation of LLM-based applications



- LangChain is a framework that facilitates creation of LLM-based applications
- Main features:



- LangChain is a framework that facilitates creation of LLM-based applications
- Main features:
 - \circ components



- LangChain is a framework that facilitates creation of LLM-based applications
- Main features:
 - \circ components
 - o off-the-shelf-chains



- LangChain is a framework that facilitates creation of LLM-based applications
- Main features:
 - components
 - o off-the-shelf-chains
- Meaning: LangChain gives you building blocks for building interesting and powerful LLM applications

Models

Models

• Abstractions over the LLM APIs like the ChatGPT API

Models

Abstractions over the LLM APIs like the ChatGPT API

```
from langchain_openai import ChatOpenAI

chat_model = ChatOpenAI(model="gpt-3.5-turbo-0125")

output = chat_model.invoke("I am teaching a live-training\
    about LLMs!")

print(output.content)
```

Prompt Templates

Prompt Templates

• Abstractions over standard prompts to LLMs

Prompt Templates

• Abstractions over standard prompts to LLMs

```
from langchain_core.prompts import ChatPromptTemplate

prompt = ChatPromptTemplate.from_template(
    """Show me 5 examples of this concept: {concept}"""
    )

prompt.format(concept="animal")

# Output
# 'Human: Show me 5 examples of this concept: animal'
```

Output Parsers

Output Parsers

• Translates raw output from LLM to a workable format

Output Parsers

• Translates raw output from LLM to a workable format

```
from langchain_core.output_parsers import StrOutputParser
output_parser = StrOutputParser()
```

Chain = Model + Prompt + Output Parser

Chain = Model + Prompt + Output Parser

• Chains are the building blocks in LangChain

Chain = Model + Prompt + Output Parser

- Chains are the building blocks in LangChain
- They are used to compose abstractions that go from simple to complex components

Chain = Model + Prompt + Output Parser

- Chains are the building blocks in LangChain
- They are used to compose abstractions that go from simple to complex components

LCEL - LangChain Expression Language

Composing Chains with LCEL

Composing Chains with LCEL

• LCEL is a powerful, declarative language designed to streamline the composition of complex chains in the LangChain framework.

Composing Chains with LCEL

- LCEL is a powerful, declarative language designed to streamline the composition of complex chains in the LangChain framework.
- Pipe syntax



Composing Chains with LCEL

- LCEL is a powerful, declarative language designed to streamline the composition of complex chains in the LangChain framework.
- Pipe syntax



```
chain = prompt | llm | output_parser
```

Composing Chains with LCEL

- LCEL is a powerful, declarative language designed to streamline the composition of complex chains in the LangChain framework.
- Pipe syntax



```
chain = prompt | llm | output_parser
```

• Allows you to build complex chain pipelines with a simple standard interface

LCEL - Runnables

• To facilitate creation of custom chains, LangChain has implemented a "Runnable" protocol.

LCEL - Runnables

- To facilitate creation of custom chains, LangChain has implemented a "Runnable" protocol.
- The standard interface includes stream, invoke, and batch methods. Async methods are also available

LCEL - Runnables

- To facilitate creation of custom chains, LangChain has implemented a "Runnable" protocol.
- The standard interface includes stream, invoke, and batch methods. Async methods are also available
- The input type and output type vary by component:

Component	Input Type	Output Type
Prompt	Dictionary	PromptValue
ChatModel	Single string, list of chat messages or a PromptValue	ChatMessage
LLM	Single string, list of chat messages or a PromptValue	String
OutputParser	The output of an LLM or ChatModel	Depends on the parser
Retriever	Single string	List of Documents
Tool	Single string or dictionary, depending on the tool	Depends on the tool

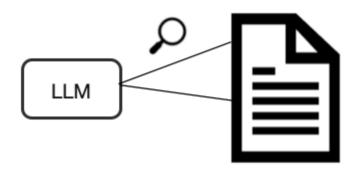
Notebook Demo - Intro to LangChain

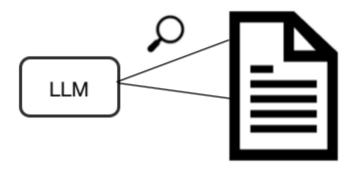
Q&A & Summary

- LLMs can predict the next word in a sequence. ("I Like eating...? ;P ")
- LangChain framework: eases the creation of LLM-based applications, featuring chains and the following basic components:
 - **Models**: Abstractions over LLM APIs (e.g ChatGPT).
 - **Prompt Templates**: Abstractions over prompts (makes them dynamic).
 - **Output Parsers**: Converts LLM outputs into usable formats (e.g string, json).
- **Chains** are the building blocks in LangChain, composed of Models, Prompt Templates, and Output Parsers.
- **LCEL** is a declarative language that users the Unix pipe symbol to build complex chain pipelines with a simple standard interface.
- Optional Exercise During Q&A

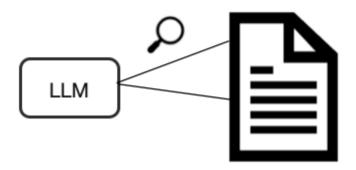
Create a simple chain for summarization of content.

Break 5 minutes

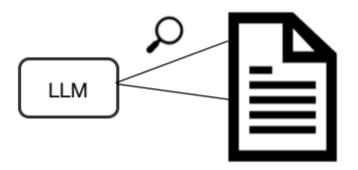




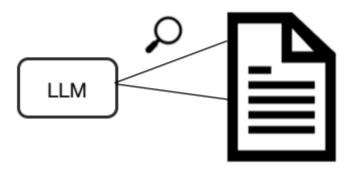
• RAG = **R**etrieval **A**ugmented **G**eneration



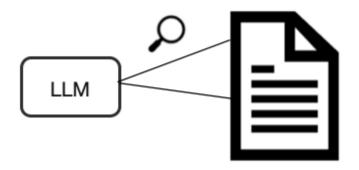
- RAG = **R**etrieval **A**ugmented **G**eneration
- It's about connecting LLMs to documents like PDFs, Text files, HTML, etc.



- RAG = **R**etrieval **A**ugmented **G**eneration
- It's about connecting LLMs to documents like PDFs, Text files, HTML, etc.
- How do we get around the context length limitations of LLMs?

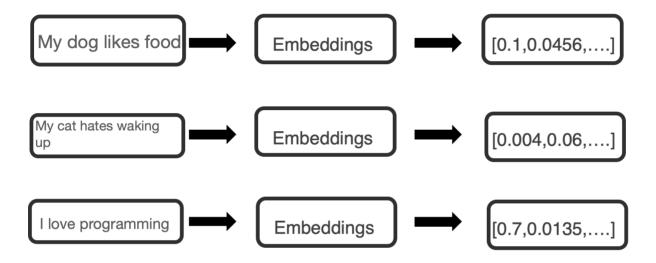


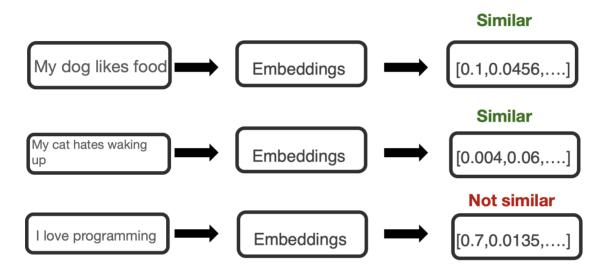
- RAG = **R**etrieval **A**ugmented **G**eneration
- It's about connecting LLMs to documents like PDFs, Text files, HTML, etc.
- How do we get around the context length limitations of LLMs?
- Quick Answer is **Embeddings**!



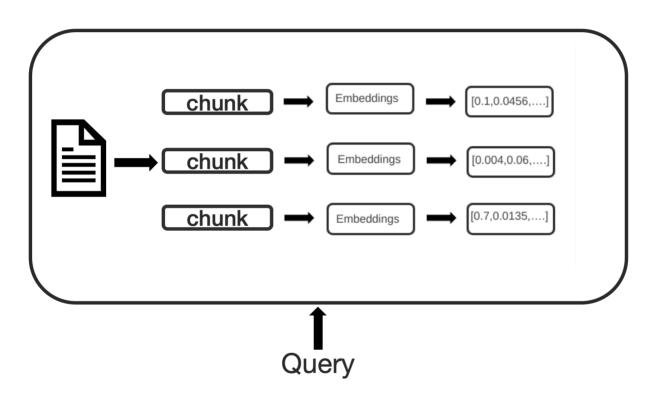
• Embeddings are vectorized representations of text



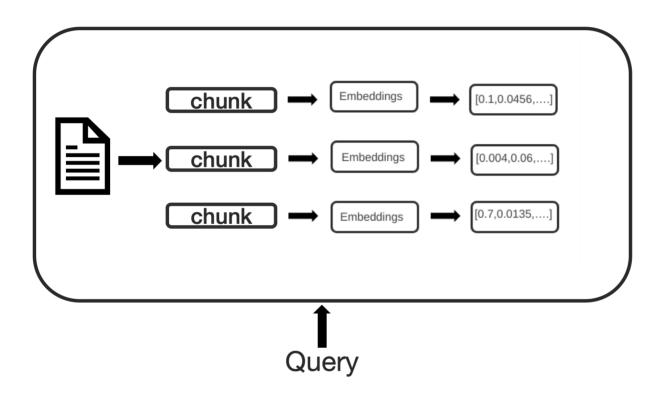




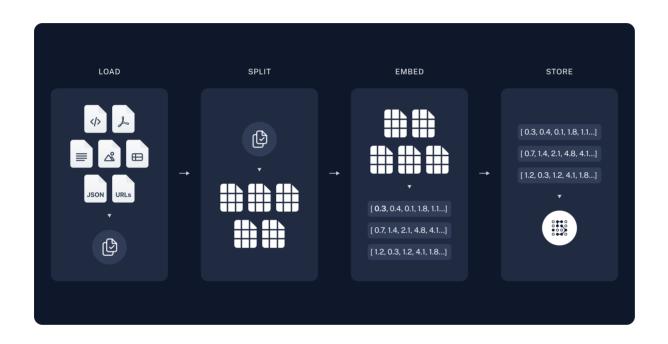
Embeddings capture content and meaning

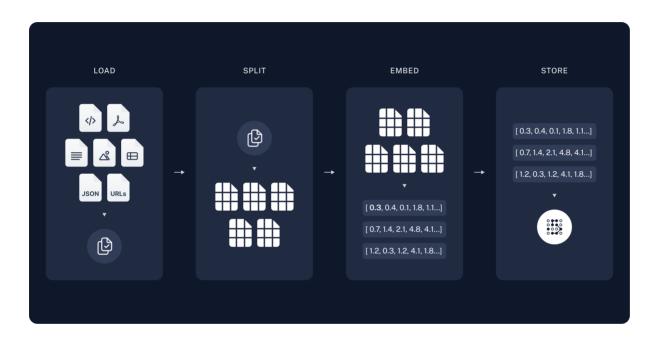


- Embeddings capture content and meaning
- Vector DBs

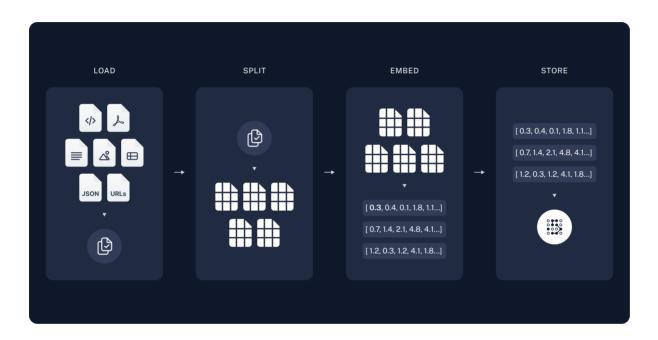


- Embeddings capture content and meaning
- Vector DBs
- How to build RAG systems with LangChain?

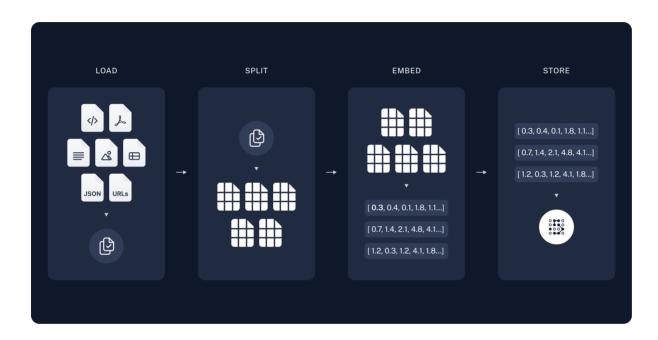




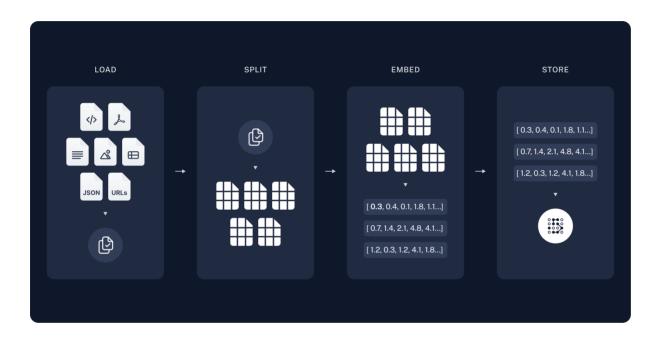
Load



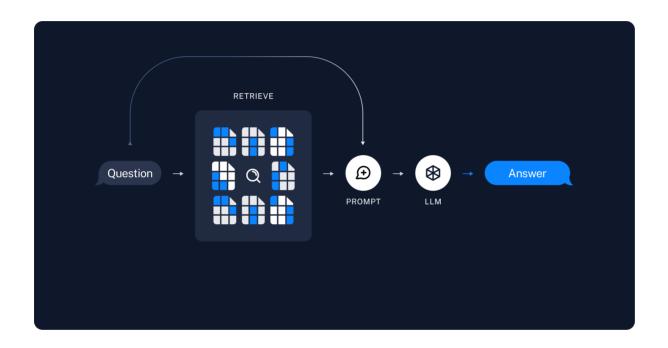
- Load
- Split

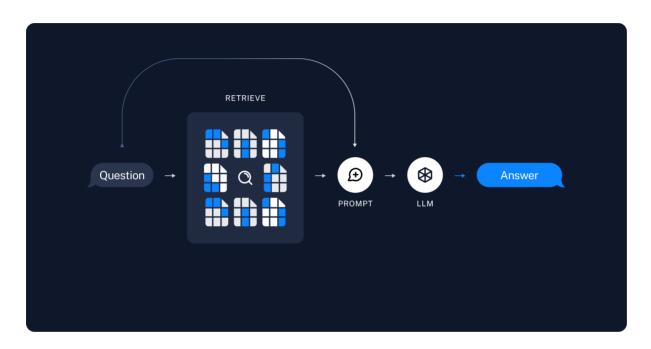


- Load
- Split
- Embed

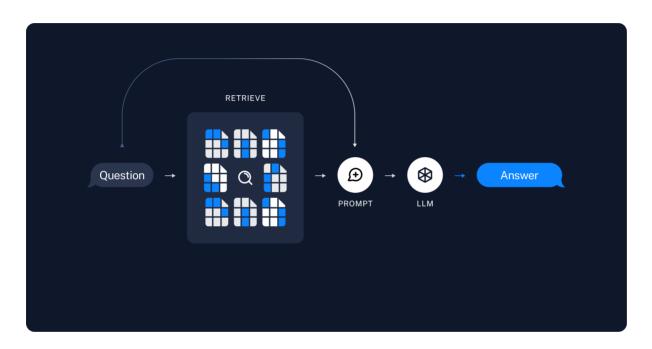


- Load
- Split
- Embed
- Store

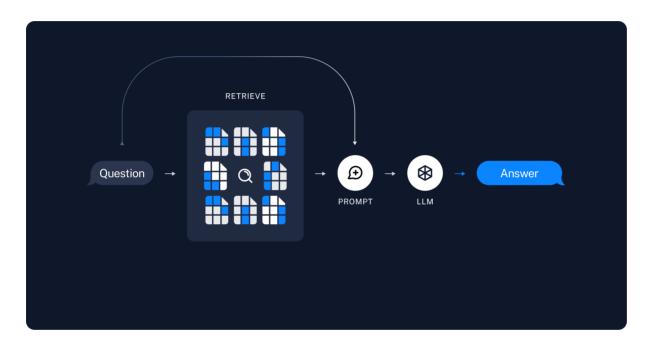




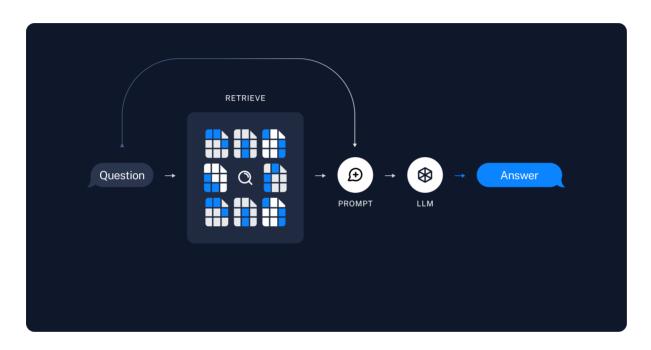
• Retrieval Piepeline



- Retrieval Piepeline
 - Input Question



- Retrieval Piepeline
 - Input Question
 - Retrieve Relevant Documents



- Retrieval Piepeline
 - Input Question
 - Retrieve Relevant Documents
 - LLM uses the prompt question + retrieved data to produce a final answer

• Sample Code

```
from langchain import hub
from langchain_community.vectorstores import Chroma
from langchain_openai import ChatOpenAI, OpenAIEmbeddings
from langchain.document_loaders import PyPDFLoader
from langchain.chains import RetrievalQA

pdf_path = "path-to-pdf.pdf"
loader = PyPDFLoader(pdf_path) # LOAD
pdf_docs = loader.load_and_split() # SPLIT
embeddings = OpenAIEmbeddings() # EMBED
vectordb = Chroma.from_documents(pdf_docs, embedding=embeddings) # STORE
retriever = vectordb.as_retriever()
llm = ChatOpenAI(model="gpt-3.5-turbo-0125")
pdf_qa = RetrievalQA.from_llm(llm=llm, retriever=retriever) # RETRIEVE
pdf_qa.invoke("What is this paper about?") # ANSWER
```

Sample Code

```
from langchain import hub
from langchain_community.vectorstores import Chroma
from langchain_openai import ChatOpenAI, OpenAIEmbeddings
from langchain.document_loaders import PyPDFLoader
from langchain.chains import RetrievalQA

pdf_path = "path-to-pdf.pdf"
loader = PyPDFLoader(pdf_path) # LOAD
pdf_docs = loader.load_and_split() # SPLIT
embeddings = OpenAIEmbeddings() # EMBED
vectordb = Chroma.from_documents(pdf_docs, embedding=embeddings) # STORE
retriever = vectordb.as_retriever()
llm = ChatOpenAI(model="gpt-3.5-turbo-0125")
pdf_qa = RetrievalQA.from_llm(llm=llm, retriever=retriever) # RETRIEVE
pdf_qa.invoke("What is this paper about?") # ANSWER
```

Notebook Demo - Q&A with LangChain

Q&A & Summary

- **RAG** = **R**etrieval **A**ugmented **G**eneration
- **RAG** is about connecting LLMs to documents like PDFs, Text files, HTML, etc.
- **Embeddings** are vectorized representations of text that capture content and meaning.
- Vector DBs are used to store and retrieve embeddings.
- **RAG** systems with LangChain are built using a pipeline that includes loading, splitting, embedding, and storing documents.
- Optional Exercise During Q&A

Create a simple RAG system with LangChain that can answer questions about pdfs or csvs.

Break 5 minutes

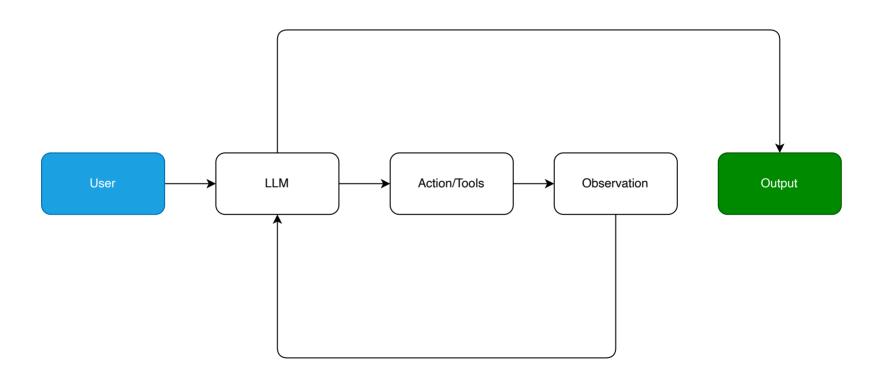
Building Agents with LangChain

Building Agents with LangChain

The Agent Loop

Building Agents with LangChain

The Agent Loop



Schema

• LangChain provides many abstractions for ease of use

- LangChain provides many abstractions for ease of use
- AgentAction: Represents the action an agent should take.

- LangChain provides many abstractions for ease of use
- AgentAction: Represents the action an agent should take.
- **AgentFinish**: Represents the final result to return to the user.

- LangChain provides many abstractions for ease of use
- **AgentAction**: Represents the action an agent should take.
- AgentFinish: Represents the final result to return to the user.
- **Intermediate Steps**: Previous actions and outputs for the current agent run.

- LangChain provides many abstractions for ease of use
- **AgentAction**: Represents the action an agent should take.
- AgentFinish: Represents the final result to return to the user.
- **Intermediate Steps**: Previous actions and outputs for the current agent run.
- **Agent**: Chain responsible for deciding the next step, powered by a language model.

Agent Inputs

Agent Inputs

• Key-value mapping.

Agent Inputs

- Key-value mapping.
- Required key: intermediate_steps.

Agent Inputs

- Key-value mapping.
- Required key: intermediate_steps.

Agent Outputs

Agent Inputs

- Key-value mapping.
- Required key: intermediate_steps.

Agent Outputs

• Next actions or final response (AgentActions or AgentFinish).

Agent Inputs

- Key-value mapping.
- Required key: intermediate_steps.

Agent Outputs

- Next actions or final response (AgentActions or AgentFinish).
- Handled by the output parser.

• The agent executor is the runtime for an agent.

- The agent executor is the runtime for an agent.
- It calls the agent, executes the actions it chooses, passes the action outputs back to the agent, and repeats.

- The agent executor is the runtime for an agent.
- It calls the agent, executes the actions it chooses, passes the action outputs back to the agent, and repeats.
- In pseudocode, this looks roughly like:

```
next_action = agent.get_action(...)
while next_action != AgentFinish:
    observation = run(next_action)
    next_action = agent.get_action(..., next_action, observation)
return next_action
```

- The agent executor is the runtime for an agent.
- It calls the agent, executes the actions it chooses, passes the action outputs back to the agent, and repeats.
- In pseudocode, this looks roughly like:

```
next_action = agent.get_action(...)
while next_action != AgentFinish:
    observation = run(next_action)
    next_action = agent.get_action(..., next_action, observation)
return next_action
```

- Runtime handles things like:
 - Handling cases where the agent selects a non-existent tool

- The agent executor is the runtime for an agent.
- It calls the agent, executes the actions it chooses, passes the action outputs back to the agent, and repeats.
- In pseudocode, this looks roughly like:

```
next_action = agent.get_action(...)
while next_action != AgentFinish:
    observation = run(next_action)
    next_action = agent.get_action(..., next_action, observation)
return next_action
```

- Runtime handles things like:
 - Handling cases where the agent selects a non-existent tool
 - Handling cases where the tool errors

- The agent executor is the runtime for an agent.
- It calls the agent, executes the actions it chooses, passes the action outputs back to the agent, and repeats.
- In pseudocode, this looks roughly like:

```
next_action = agent.get_action(...)
while next_action != AgentFinish:
    observation = run(next_action)
    next_action = agent.get_action(..., next_action, observation)
return next action
```

- Runtime handles things like:
 - Handling cases where the agent selects a non-existent tool
 - Handling cases where the tool errors
 - Handling cases where the agent produces output that cannot be parsed into a tool invocation

- The agent executor is the runtime for an agent.
- It calls the agent, executes the actions it chooses, passes the action outputs back to the agent, and repeats.
- In pseudocode, this looks roughly like:

```
next_action = agent.get_action(...)
while next_action != AgentFinish:
    observation = run(next_action)
    next_action = agent.get_action(..., next_action, observation)
return next action
```

- Runtime handles things like:
 - Handling cases where the agent selects a non-existent tool
 - Handling cases where the tool errors
 - Handling cases where the agent produces output that cannot be parsed into a tool invocation
 - Logging and observability at all levels (agent decisions, tool calls) to stdout and/or to LangSmith.

• Functions that an agent can call.

- Functions that an agent can call.
- Consists of:

- Functions that an agent can call.
- Consists of:
 - Input schema for the tool.

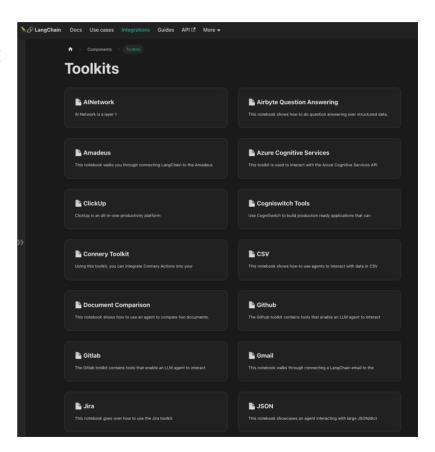
- Functions that an agent can call.
- Consists of:
 - Input schema for the tool.
 - Function to run.

- Functions that an agent can call.
- Consists of:
 - Input schema for the tool.
 - Function to run.
- Important for building a working agent.

Langchain Toolkits

Langchain Toolkits

- LangChain provides a wide set of toolkits.
- Groups of 3-5 tools for specific objectives.
- Example: GitHub toolkit for interacting with GitHub.



Let's Build Agents!

Notebook Demo - Building LLM Agents with LangChain; Github Agent; Research Assistant

Q&A & Summary

- LangChain provides abstractions for building agents, including AgentAction, AgentFinish, and Agent.
- **AgentExecutor** is the runtime for an agent, handling agent decisions, tool calls, and observability.
- **Tools** in LangChain are functions that an agent can call, consisting of an input schema and a function to run.
- LangChain provides a wide set of toolkits, groups of 3-5 tools for specific objectives, an example is GitHub toolkit for interacting with GitHub.
- Optional Exercise During Q&A

Create a simple agent that can create a schedule for you given a table of tasks and dead table format = task | date

Break 5 minutes

• LangServe helps developers deploy LangChain runnables and chains as a REST API.

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.
- It provides a client that can be used to call into runnables deployed on a server.

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.
- It provides a client that can be used to call into runnables deployed on a server.
- Input and Output schemas automatically inferred from your LangChain object, and enforced on every API call, with rich error messages

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.
- It provides a client that can be used to call into runnables deployed on a server.
- Input and Output schemas automatically inferred from your LangChain object, and enforced on every API call, with rich error messages
- Efficient /invoke/, /batch/ and /stream/ endpoints with support for many concurrent requests on a single server

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.
- It provides a client that can be used to call into runnables deployed on a server.
- Input and Output schemas automatically inferred from your LangChain object, and enforced on every API call, with rich error messages
- Efficient /invoke/, /batch/ and /stream/ endpoints with support for many concurrent requests on a single server
- Playground page at /playground/ with streaming output and intermediate steps

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.
- It provides a client that can be used to call into runnables deployed on a server.
- Input and Output schemas automatically inferred from your LangChain object, and enforced on every API call, with rich error messages
- Efficient /invoke/, /batch/ and /stream/ endpoints with support for many concurrent requests on a single server
- Playground page at /playground/ with streaming output and intermediate steps
- Built-in (optional) tracing to LangSmith, just add your API key (see Instructions)

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.
- It provides a client that can be used to call into runnables deployed on a server.
- Input and Output schemas automatically inferred from your LangChain object, and enforced on every API call, with rich error messages
- Efficient /invoke/, /batch/ and /stream/ endpoints with support for many concurrent requests on a single server
- Playground page at /playground/ with streaming output and intermediate steps
- Built-in (optional) tracing to LangSmith, just add your API key (see Instructions)
- Use the client SDK to call a LangServe server as if it was a Runnable running locally (or call the HTTP API directly)

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.
- It provides a client that can be used to call into runnables deployed on a server.
- Input and Output schemas automatically inferred from your LangChain object, and enforced on every API call, with rich error messages
- Efficient /invoke/, /batch/ and /stream/ endpoints with support for many concurrent requests on a single server
- Playground page at /playground/ with streaming output and intermediate steps
- Built-in (optional) tracing to LangSmith, just add your API key (see Instructions)
- Use the client SDK to call a LangServe server as if it was a Runnable running locally (or call the HTTP API directly)

Final Q&A & Summary

- LangServe helps developers deploy LangChain runnables and chains as a REST API.
- It is integrated with FastAPI and uses pydantic for data validation.

References

- <u>LangChain Intro Docs</u>
- LangChain Documentation
- Gen Agents
- WebGPT
- OpenAI
- OpenAI Function Calling
- AutoGPT
- <u>GPT-Engineer</u>
- <u>BabyAGI</u>
- Karpathy on Agents
- ReACT Paper