

# Integrating document loaders

DEVELOPING LLM APPLICATIONS WITH LANGCHAIN

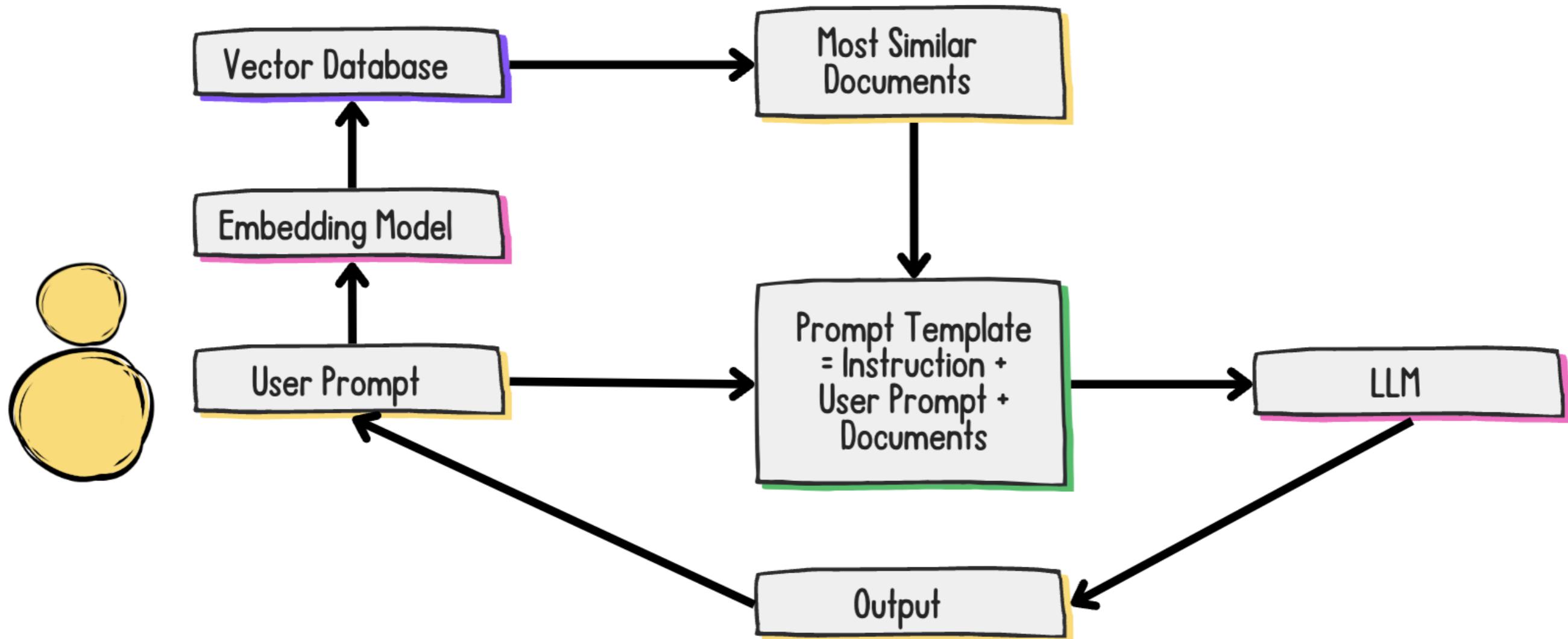


**Jonathan Bennion**

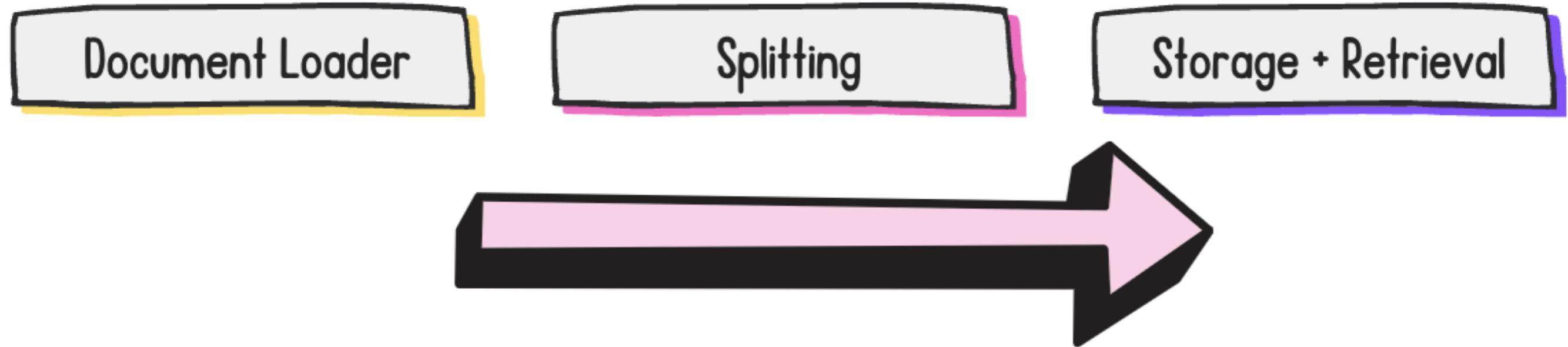
AI Engineer & LangChain Contributor

# Retrieval Augmented Generation (RAG)

- Use embeddings to *retrieve* relevant information to integrate into the *prompt*



# RAG development steps



# LangChain document loaders

- Classes designed to *load* and *configure* documents for system integration
- Document loaders for common file types:  
.pdf , .csv
- 3rd party loaders: S3, .ipynb , .wav



<sup>1</sup> [https://python.langchain.com/docs/integrations/document\\_loaders](https://python.langchain.com/docs/integrations/document_loaders)

# PDF document loader

- Requires installation of the `pypdf` package: `pip install pypdf`

```
from langchain_community.document_loaders import PyPDFLoader
loader = PyPDFLoader("path/to/file/attention_is_all_you_need.pdf")

data = loader.load()
print(data[0])
```

```
Document(page_content='Provided proper attribution is provided, Google hereby grants
permission to\nreproduce the tables and figures in this paper solely for use in [...]')
```

# CSV document loader

```
from langchain_community.document_loaders.csv_loader import CSVLoader\n\nloader = CSVLoader('fifa_countries_audience.csv')\n\ndata = loader.load()\nprint(data[0])
```

```
Document(page_content='country: United States\\nconfederation: CONCACAF\\npopulation_share: [...]
```

# HTML document loader

- Requires installation of the `unstructured` package: `pip install unstructured`

```
from langchain_community.document_loaders import UnstructuredHTMLLoader

loader = UnstructuredHTMLLoader("white_house_executive_order_nov_2023.html")
data = loader.load()

print(data[0])
print(data[0].metadata)
```

```
page_content="To search this site, enter a search term\n\nSearch\n\nExecutive Order on the Safe, Secure,  
and Trustworthy Development and Use of Artificial Intelligence\n\nHome\n\nBriefing Room\n\nPresidential  
Actions\n\nBy the authority vested in me as President by the Constitution and the laws of the United  
States of America, it is hereby ordered as follows: ..."

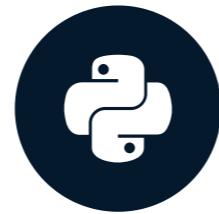
{'source': 'white_house_executive_order_nov_2023.html'}
```

# **Let's practice!**

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# Splitting external data for retrieval

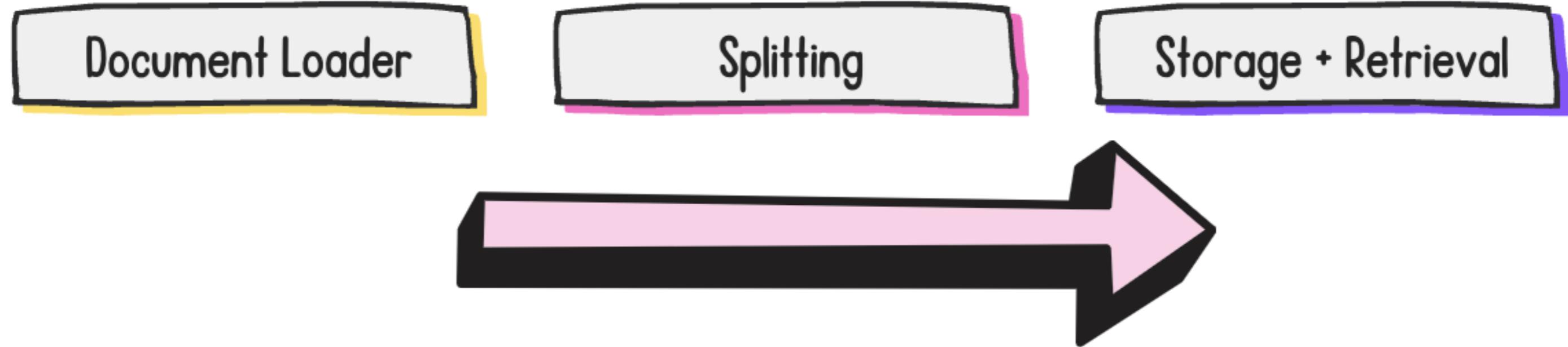
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# RAG development steps



- **Document splitting:** split document into *chunks*
- Break documents up to fit within an LLM's *context window*

# Thinking about splitting...

## 1 Introduction

Recurrent neural networks, long short-term memory [13] and gated recurrent [7] neural networks in particular, have been firmly established as state of the art approaches in sequence modeling and transduction problems such as language modeling and machine translation [35, 2, 5]. Numerous efforts have since continued to push the boundaries of recurrent language models and encoder-decoder architectures [38, 24, 15].

Line 1:

Recurrent neural networks, long short-term memory [13] and gated recurrent [7] neural networks

Line 2:

in particular, have been firmly established as state of the art approaches in sequence modeling and

<sup>1</sup> <https://arxiv.org/abs/1706.03762>

# Chunk overlap

Recurrent neural networks, long short-term memory [13] and gated recurrent [7] neural networks in particular, have been firmly established as state of the art approaches in sequence modeling and transduction problems such as language modeling and machine translation [35, 2, 5]. Numerous efforts have since continued to push the boundaries of recurrent language models and encoder-decoder architectures [38, 24, 15].

# What is the best document splitting strategy?



1. CharacterTextSplitter
2. RecursiveCharacterTextSplitter
3. Many others

<sup>1</sup> Wikipedia Commons

```
quote = '''One machine can do the work of fifty ordinary humans.\nNo machine can do  
the work of one extraordinary human.'''
```

```
len(quote)
```

```
103
```

```
chunk_size = 24  
chunk_overlap = 3
```

<sup>1</sup> Elbert Hubbard

```
from langchain_text_splitters import CharacterTextSplitter

ct_splitter = CharacterTextSplitter(
    separator='.',
    chunk_size=chunk_size,
    chunk_overlap=chunk_overlap)

docs = ct_splitter.split_text(quote)
print(docs)
print([len(doc) for doc in docs])
```

```
['One machine can do the work of fifty ordinary humans',
 'No machine can do the work of one extraordinary human']
[52, 53]
```

- Split on separator so < `chunk_size` , but **may not always succeed!**

```
from langchain_text_splitters import RecursiveCharacterTextSplitter

rc_splitter = RecursiveCharacterTextSplitter(
    separators=["\n\n", "\n", " ", ""],
    chunk_size=chunk_size,
    chunk_overlap=chunk_overlap)

docs = rc_splitter.split_text(quote)
print(docs)
```

# RecursiveCharacterTextSplitter

- `separators=["\n\n", "\n", " ", ""]`

```
[ 'One machine can do the',
  'work of fifty ordinary',
  'humans.',
  'No machine can do the',
  'work of one',
  'extraordinary human. ']
```

1. Try splitting by paragraph: `"\n\n"`
2. Try splitting by sentence: `"\n"`
3. Try splitting by words: `" "`

# RecursiveCharacterTextSplitter with HTML

```
from langchain_community.document_loaders import UnstructuredHTMLLoader
from langchain_text_splitters import RecursiveCharacterTextSplitter

loader = UnstructuredHTMLLoader("white_house_executive_order_nov_2023.html")
data = loader.load()

rc_splitter = RecursiveCharacterTextSplitter(
    chunk_size=chunk_size,
    chunk_overlap=chunk_overlap,
    separators=['.'])

docs = rc_splitter.split_documents(data)
print(docs[0])
```

Document(page\_content="To search this site, enter a search term [...]")

# **Let's practice!**

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# RAG storage and retrieval using vector databases

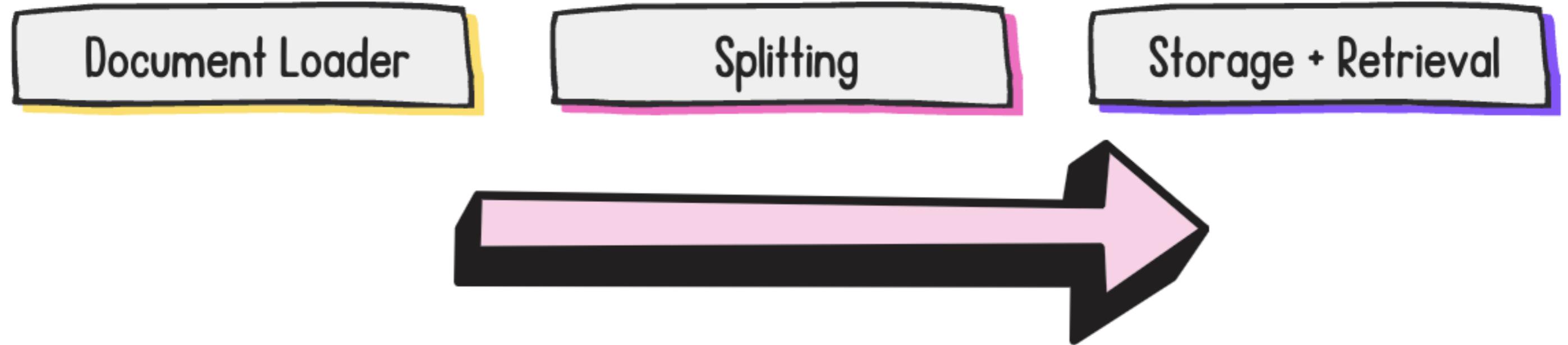
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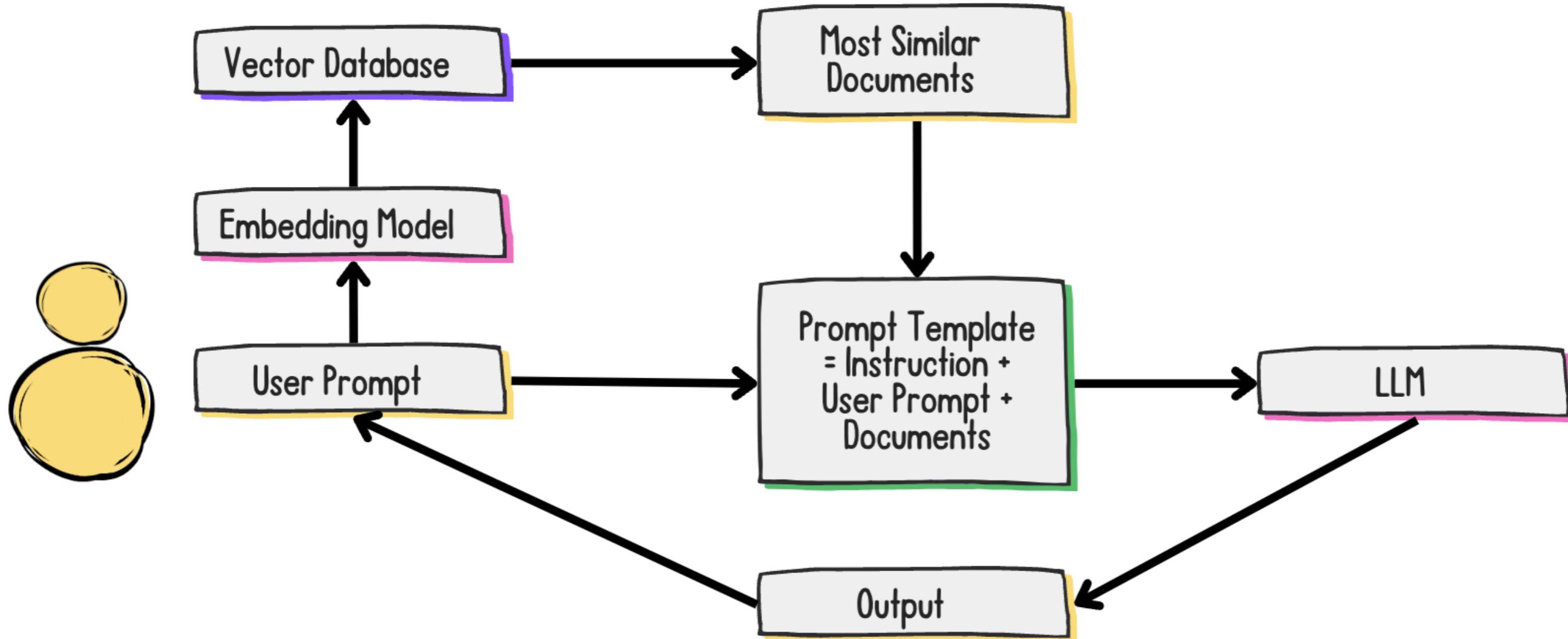


# RAG development steps

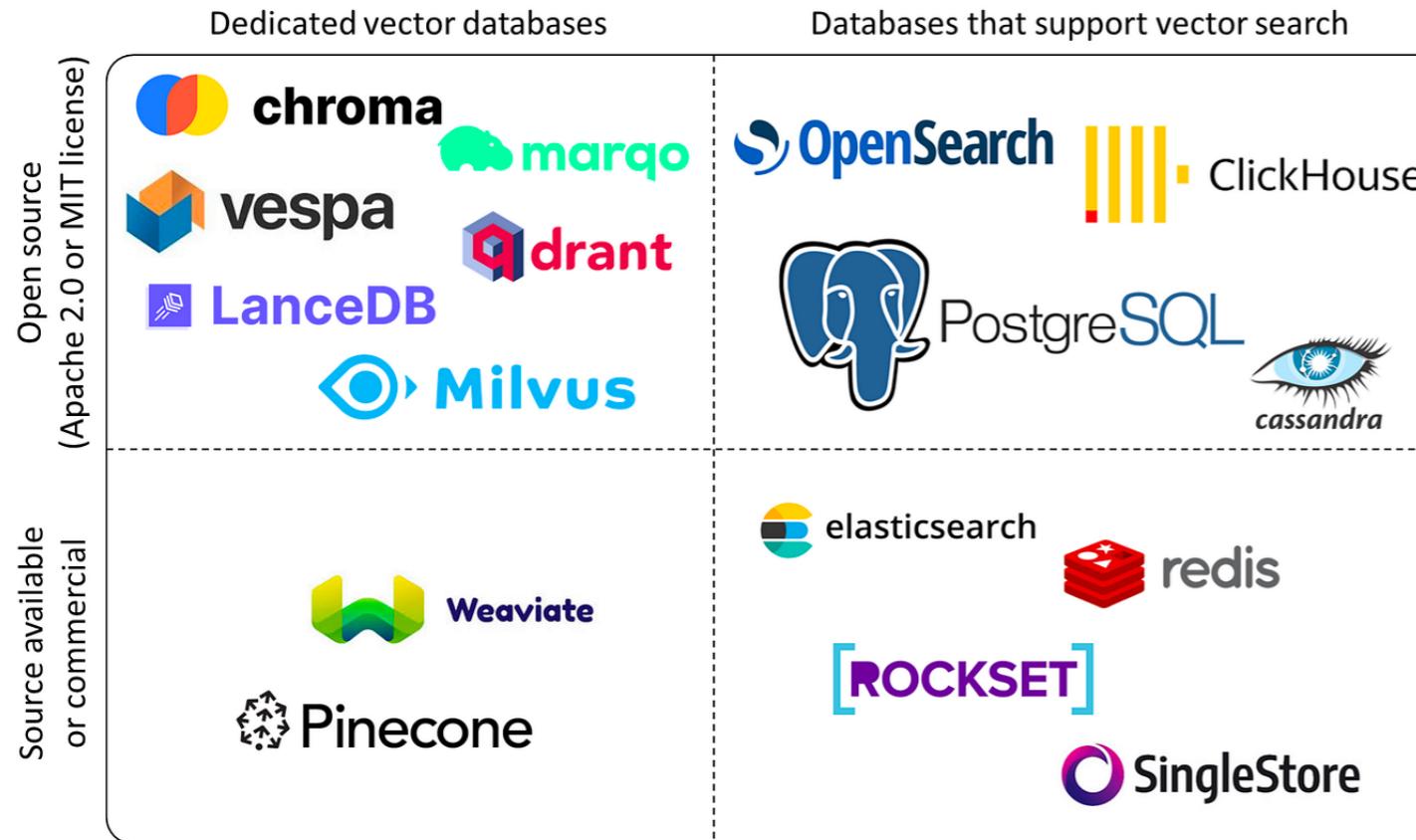


- Focus of this video: *storage and retrieval*

# What is a vector database and why do I need it?



# Which vector database should I use?



## Need to consider:

- Open source vs. closed source (license)
- Cloud vs. on-premises
- Lightweight vs. powerful

<sup>1</sup> Image Credit: Yingjun Wu

# Meet the documents...

docs

```
[  
  Document(  
    page_content="In all marketing copy, TechStack should always be written with the T and S  
    capitalized. Incorrect: techstack, Techstack, etc.",  
    metadata={"guideline": "brand-capitalization"}  
,  
  Document(  
    page_content="Our users should be referred to as techies in both internal and external  
    communications.",  
    metadata={"guideline": "referring-to-users"}  
)  
]
```

# Setting up a Chroma vector database

```
from langchain_openai import OpenAIEmbeddings
from langchain_chroma import Chroma

embedding_function = OpenAIEmbeddings(api_key=openai_api_key, model='text-embedding-3-small')

vectorstore = Chroma.from_documents(
    docs,
    embedding=embedding_function,
    persist_directory="path/to/directory"
)

retriever = vectorstore.as_retriever(
    search_type="similarity",
    search_kwargs={"k": 2}
)
```

# Building a prompt template

```
from langchain_core.prompts import ChatPromptTemplate
```

```
message = """
```

Review and fix the following TechStack marketing copy with the following guidelines in consideration:

Guidelines:

```
{guidelines}
```

Copy:

```
{copy}
```

Fixed Copy:

```
"""
```

```
prompt_template = ChatPromptTemplate.from_messages([('human', message)])
```

# Chaining it all together!

```
from langchain_core.runnables import RunnablePassthrough

rag_chain = ({"guidelines": retriever, "copy": RunnablePassthrough()
              | prompt_template
              | llm)

response = rag_chain.invoke("Here at techstack, our users are the best in the world!")
print(response.content)
```

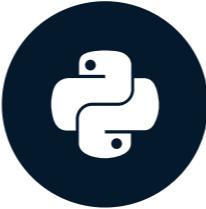
Here at TechStack, our techies are the best in the world!

# **Let's practice!**

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# Wrap-up!

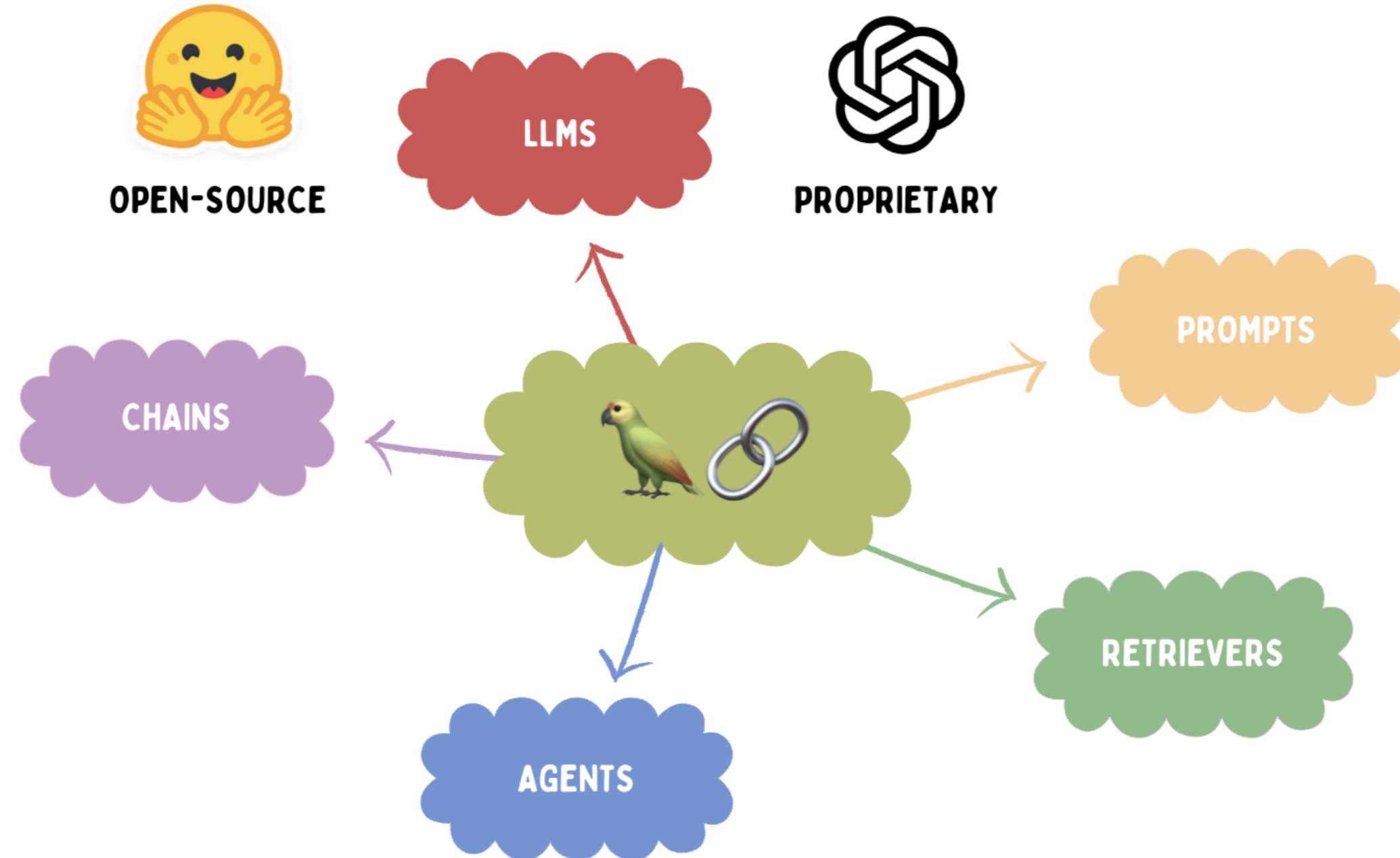
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# LangChain's core components



# Chains and agents

User Input: Why isn't my code working? Here it is...

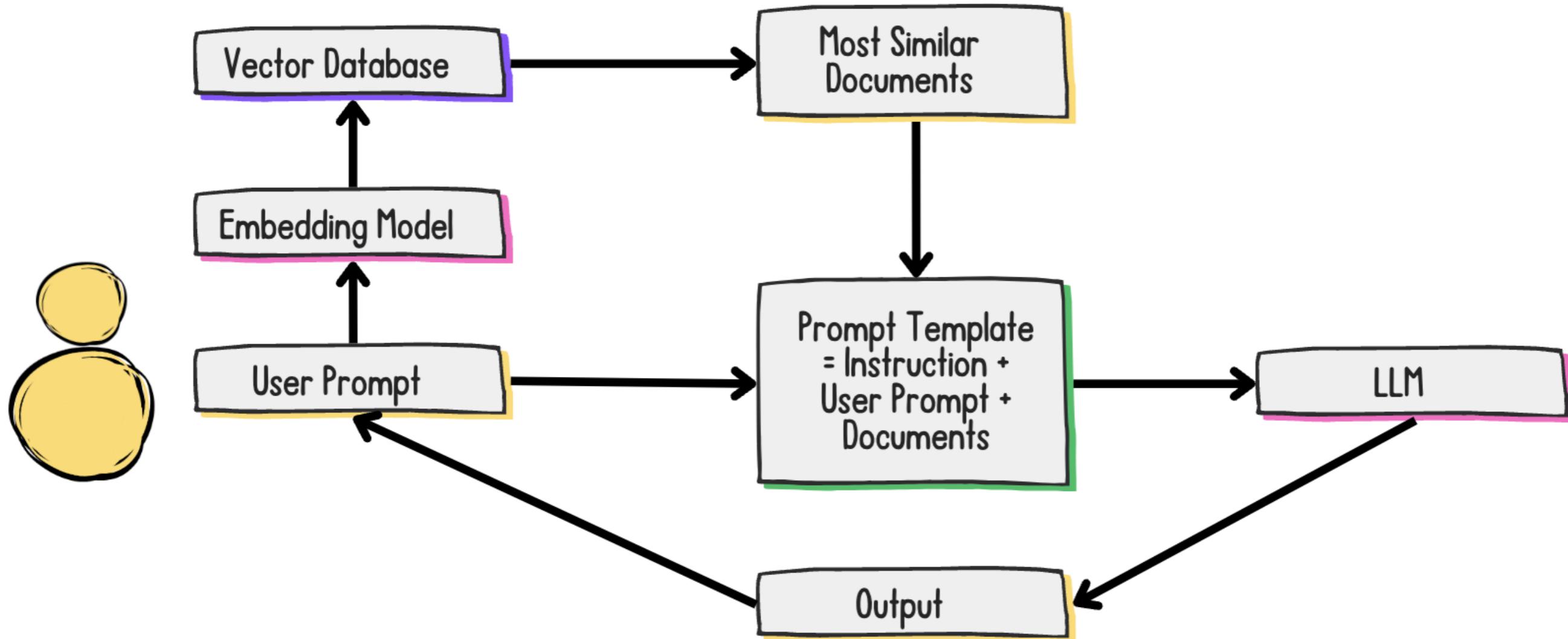


Agent

- Run Code
- Search Internet
- Load Document

Tools

# Retrieval Augmented Generation (RAG)



# LangChain Hub

The screenshot shows the LangChain Hub interface. On the left is a sidebar with various filters:

- Use Cases:**
  - Agent simulations (4)
  - Agents (53)
  - Autonomous agents (11)
  - Chatbots (73)
  - Classification (5)
  - Code understanding (17)
  - Code writing (19)
  - Evaluation (21)
  - Extraction (38)
  - Interacting with APIs (17)
  - Multi-modal (3)
  - QA over documents (59)
  - Self-checking (8)
  - SQL (5)
  - Summarization (59)
  - Tagging (9)
- Type:**
  - ChatPromptTemp... (240)
  - StringPromptTem... (183)
- Language:**

The main area features a search bar at the top: "Search for prompts, use cases, models..." followed by four filter buttons: "Top Favorited", "Top Viewed", "Top Downloaded", and "Recently Updated". Below these are three examples of prompts:

- homanp/superagent**

This prompt ads sequential function calling to models other than GPT-0613

{x} Prompt • Updated 3 months ago • ❤️ 62 • 🎁 29.2k • ⏪ 1.91k • ➔ 11

Try it
- hardkothari/prompt-maker**

Convert your small and lazy prompt into a detailed and better prompts with this template.

{x} Prompt • Updated 3 months ago • ❤️ 52 • 🎁 14k • ⏪ 1.47k • ➔ 1

Try it
- smithing-gold/assumption-checker**

Assert whether assumptions are made in a user's query and provide follow up questions to debunk their claims.

Try it

Access the LangChain Hub at: <https://smith.langchain.com/hub>

[+ Request a template](#)

## Featured

rag      [OpenAI](#) [Pinecone](#)

**rag-conversation**  
by Elastic  
Conversational RAG using Pinecone

[Github](#) [11](#)

extraction      [OpenAI](#) [Function Calling](#)

**extraction-openai-functions**  
by LangChain  
Use OpenAI function calling for tasks like...

[Github](#) [12](#)

agent      [Anthropic](#)

**xml-agent**  
by LangChain  
Agent that uses XML syntax to communicat...

[Github](#) [6](#)

rag      [OpenAI](#) [Chroma](#) [Gpt4all](#)

**rag-chroma-private**  
by LangChain  
Private RAG using local LLM, embeddings,...

[Github](#) [14](#)

research      [OpenAI](#) [Tavily](#)

**openai-functions-agent**  
by LangChain  
Agent using OpenAI function calling to...

[Github](#) [11](#)

# The LangChain ecosystem



**LangSmith:** troubleshooting and evaluating applications

**LangServe:** deploying applications

**LangGraph:** multi-agent knowledge graphs

# **Let's practice!**

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