# LangChain Chat with your data

### Overview





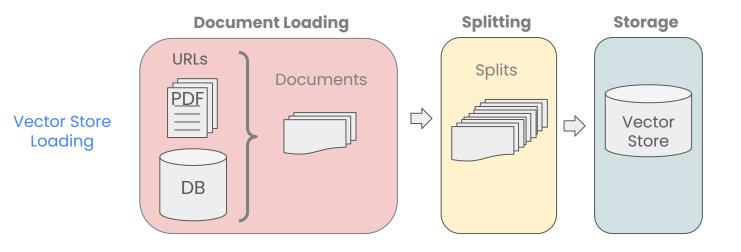


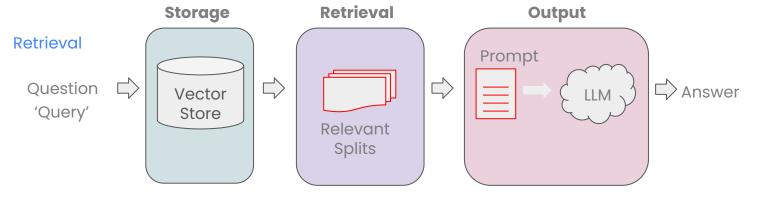


# Retrieval Augmented Generation

Retrieval Augmented Generation (RAG) is a very popular paradigm.

 Retrieve relevant documents and load into "working memory" / context window.



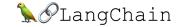


# LangChain Chat with your data

# Document Loading



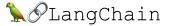




#### Loaders

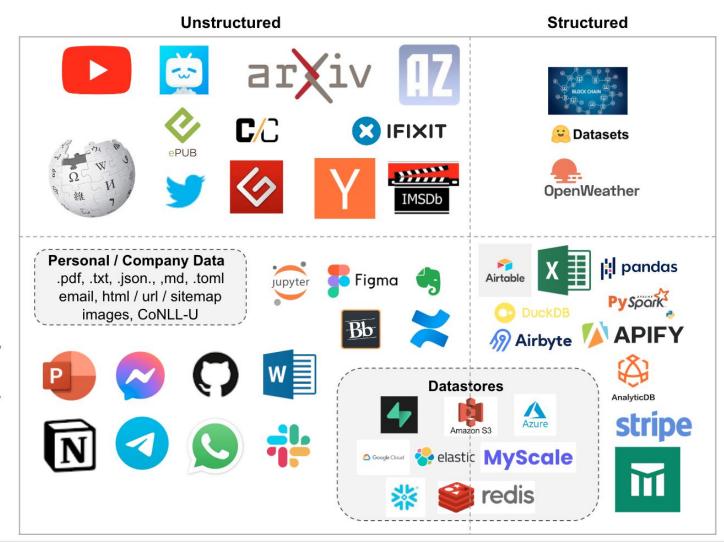
- Loaders deal with the specifics of accessing and converting data
  - Accessing
    - Web Sites
    - Data Bases
    - YouTube
    - arXiv
    - **...**
  - Data Types
    - PDF
    - HTML
    - JSON
    - Word, PowerPoint...
- Returns a list of `Document` objects:

```
[
Document(page_content='MachineLearning-Lecture01 \nInstructor (Andrew Ng): Okay.
Good morning. Welcome to CS229....',
metadata={'source': 'docs/cs229_lectures/MachineLearning-Lecture01.pdf', 'page': 0})
...
Document(page_content='[End of Audio] \nDuration: 69 minutes \,
metadata={'source': 'docs/cs229_lectures/MachineLearning-Lecture01.pdf', 'page': 21})
1
```





#### **Document Loaders**



**Public** 

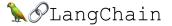
**Proprietary** 

# LangChain Chat with your data

# Document Splitting



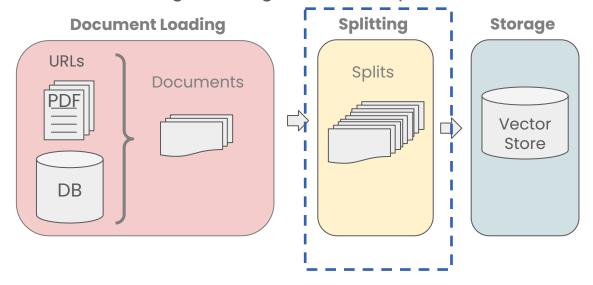






# Document Splitting

- Splitting Documents into smaller chunks
  - Retaining meaningful relationships!



•••

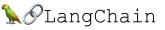
on this model. The Toyota Camry has a head-snapping 80 HP and an eight-speed automatic transmission that will

•••

Chunk 1: on this model. The Toyota Camry has a head-snapping

Chunk 2: 80 HP and an eight-speed automatic transmission that will

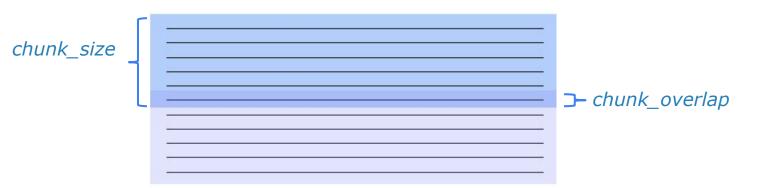
Question: What are the specifications on the Camry?



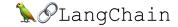


# **Example Splitter**

```
langchain.text_splitter.CharacterTextSplitter(
    separator: str = "\n\n"
    chunk_size=4000,
    chunk_overlap=200,
    length_function=<builtin function len>,
)
Methods:
create_documents() - Create documents from a list of texts.
split_documents() - Split documents.
```







# Types of splitters

#### langchain.text\_splitter.

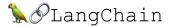
- CharacterTextSplitter()- Implementation of splitting text that looks at characters.
- MarkdownHeaderTextSplitter() Implementation of splitting markdown files based on specified headers.
- TokenTextSplitter() Implementation of splitting text that looks at tokens.
- SentenceTransformersTokenTextSplitter() Implementation of splitting text that looks at tokens.
- RecursiveCharacterTextSplitter() Implementation of splitting text that looks at characters. Recursively tries to split by different characters to find one that works.
- Language() for CPP, Python, Ruby, Markdown etc
- NLTKTextSplitter() Implementation of splitting text that looks at sentences using NLTK (Natural Language Tool Kit)
- SpacyTextSplitter() Implementation of splitting text that looks at sentences using Spacy

# LangChain Chat with your data

**Vector Stores and Embeddings** 

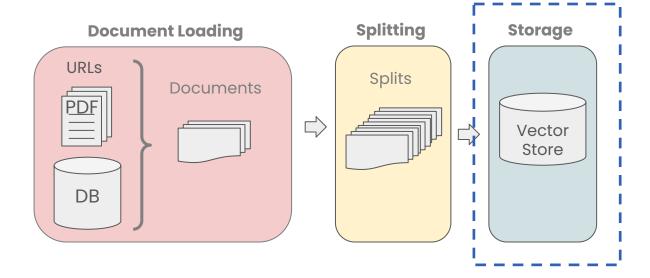


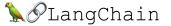






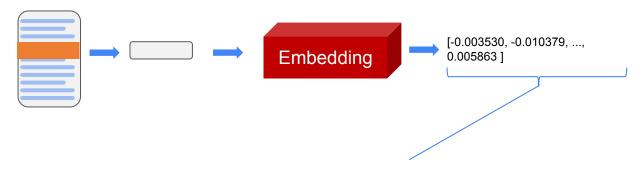
#### **Vector Stores**





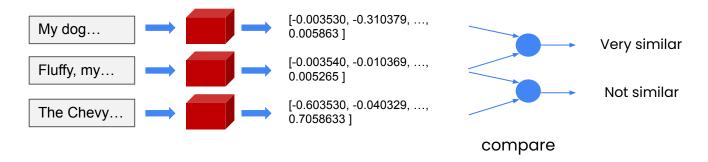


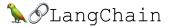
## Embeddings



- Embedding vector captures content/meaning
- Text with similar content will have similar vectors

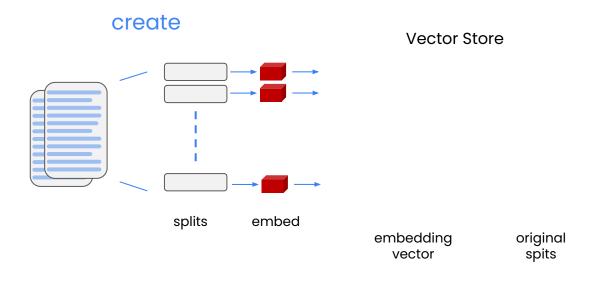
- 1) My dog Rover likes to chase squirrels.
- 2) Fluffy, my cat, refuses to eat from a can.
- 3) The Chevy Bolt accelerates to 60 mph in 6.7 seconds.



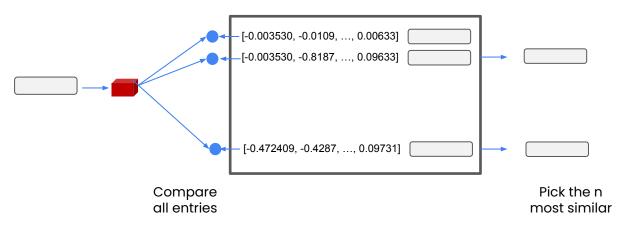


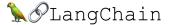


#### **Vector Store**



#### index

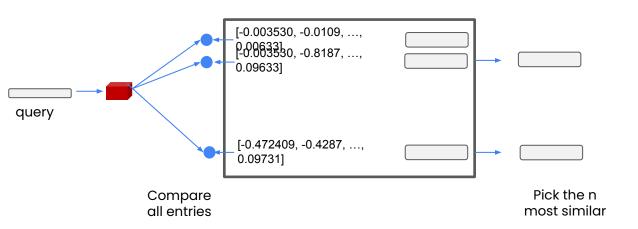






# Vector Store/Database

#### index



#### Process with Ilm



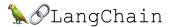
The returned values can now fit in the LLM context

# LangChain Chat with your data

#### Retrieval

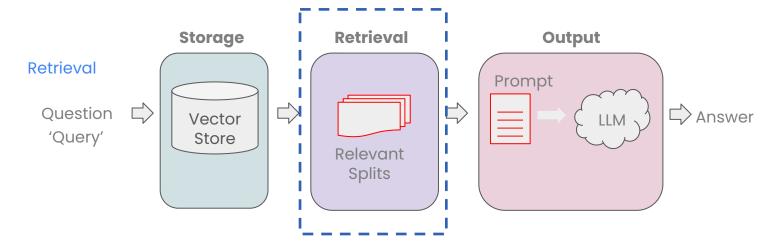




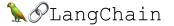




#### Retrieval



- Accessing/indexing the data in the vector store
  - Basic semantic similarity
  - Maximum marginal relevance
  - Including Metadata
- LLM Aided Retrieval





# Maximum marginal relevance(MMR)

 You may not always want to choose the most similar responses

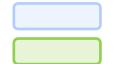


Tell me about all-white mushrooms with large fruiting bodies

The Amanita phalloides has a large and imposing epigeous (aboveground) fruiting body (basidiocarp).

A mushroom with a large fruiting body is the Amanita phalloides. Some varieties are all-white.

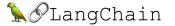
AA. phalloides, a.k.a Death Cap, is one of the most poisonous of all known mushrooms.



**Most Similar** 



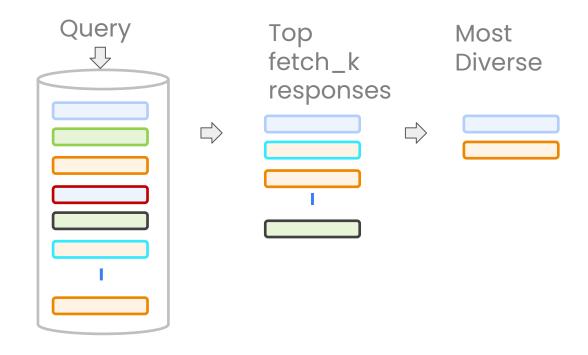
**MMR** 

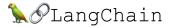




# MMR algorithm

- Query the Vector Store
- Choose the `fetch\_k` most similar responses
- Within those responses choose the `k` most diverse







#### **LLM Aided Retrieval**

- There are several situations where the Query applied to the DB is more than just the Question asked.
- One is SelfQuery

# Information Query Query Post processing Relevant splits

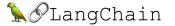
#### Self-query

Information: Query format

**Query: Question** 

Filter: eq["section", "testing"]

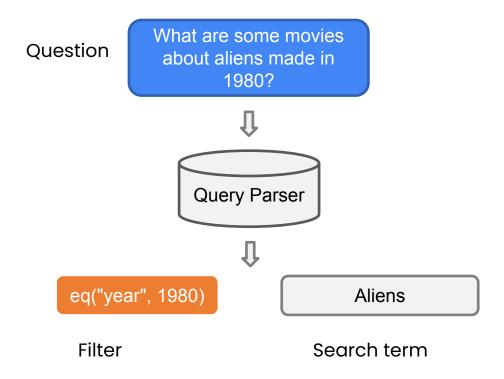
Query parser



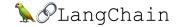


#### LLM Aided Retrieval

- There are several situations where the Query applied to the DB is more than just the Question asked.
- One is SelfQuery, where we use an LLM to convert the user question into a query

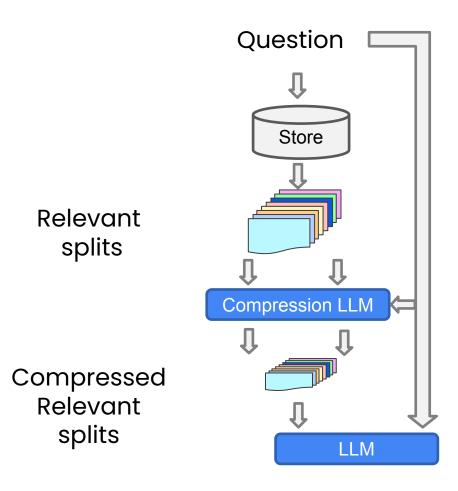






# Compression

 Increase the number of results you can put in the context by shrinking the responses to only the relevant information.

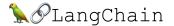


# LangChain Chat with your data

# **Question Answering**

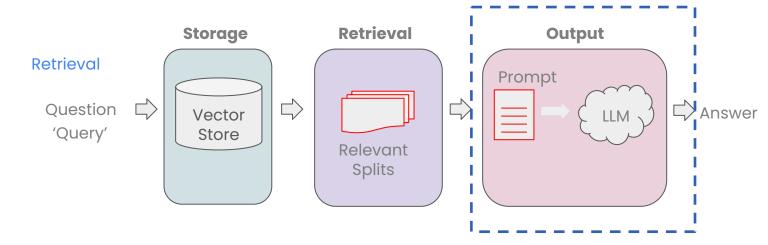




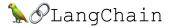




## **Question Answering**



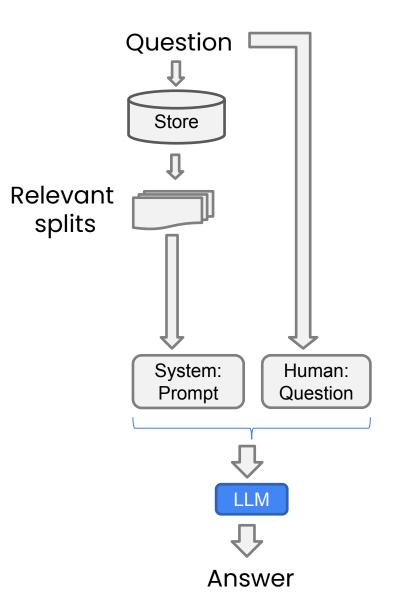
- Multiple relevant documents have been retrieved from the vector store
- Potentially compress the relevant splits to fit into the LLM context
- Send the information along with our question to an LLM to select and format an answer





## RetrievalQA chain

RetrievalQA.from\_chain\_type(, chain\_type="stuff",...)

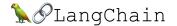


Question is applied to the Vector Store as a query

Vector store provides k relevant documents

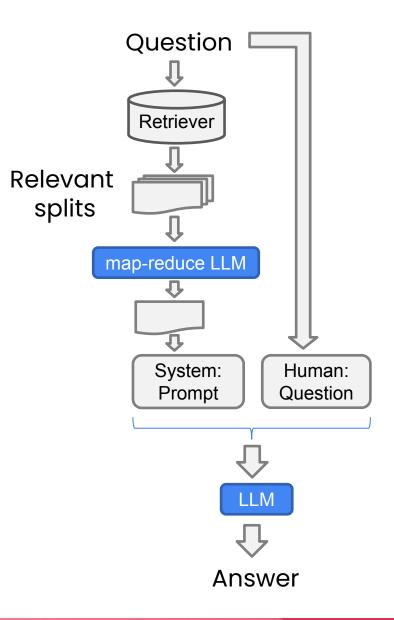
Docs and original question are sent to an LLM





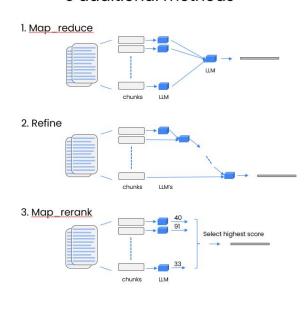
# Retrieval Chain with LLM selection

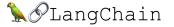
RetrievalQA.from\_chain\_type(, chain\_type="map\_reduce",...)



You many have too many docs to fit into an LLM context. The solution is to use an LLM to select the 'most relevant' information

#### 3 additional methods

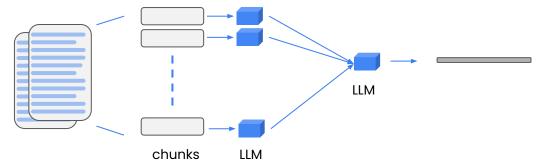




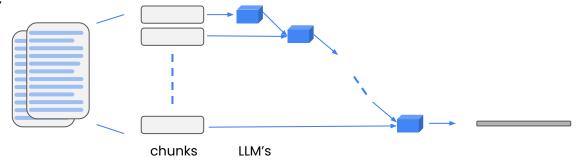


## 3 additional methods

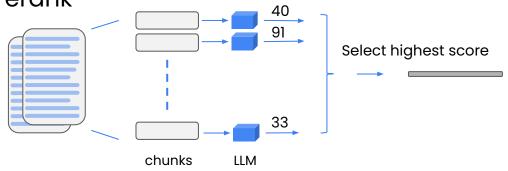
#### 1. Map\_reduce



#### 2. Refine



#### 3. Map\_rerank

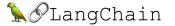


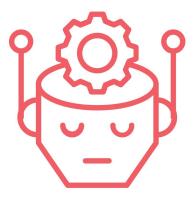
# LangChain for LLM Application Development

## Agents

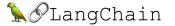




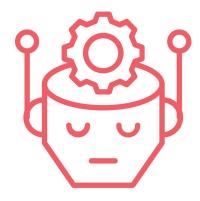




Agent refers to the idea of using large language models as reasoning engine to determine which actions to take and in what order.



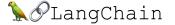


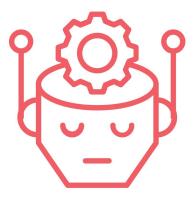


Agents use an LLM to determine which actions to take and in what order.

An action can be using a tool and observing its output and deciding what to return to the user.

from langchain.agents import initialize\_agent, AgentType





To construct an agent, you need:

- PromptTemplate: this is responsible for taking the user input and previous steps and constructing a prompt to send to the language model
- Language Model: this takes the prompt constructed by the PromptTemplate and returns some output
- Output Parser: this takes the output of the Language
   Model and parses it into an AgentAction or AgentFinish object.

