



# Generative AI LangChain

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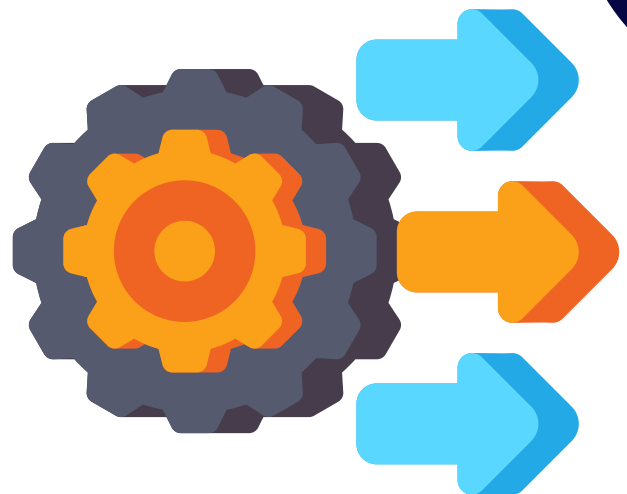


# Model I/O

In this part of the course, we'll start learning about LangChain. We'll learn how to ask questions to models and understand their answers.

LangChain mainly sends messages to language models and deals with what they say back.

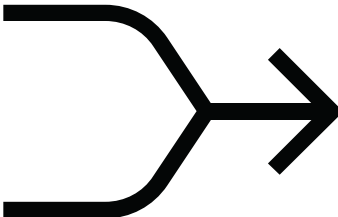
LangChain gives the building blocks to interface with any language model.



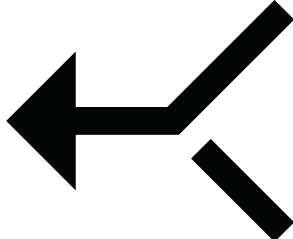
MODEL  
I/O

x = 'India'  
y = 'China'

How far {x} is from  
{y}?

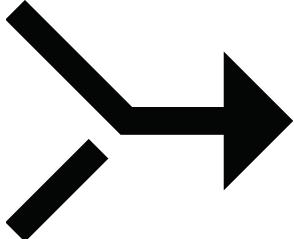


How far India is from  
China?



LLM

CHAT  
MODELS



The distance between  
India and China is ...

# INPUT (PROMPTS)

instructions or input that set the  
context and parameters for the  
model's operation

# MODELS

LLM takes a text string as input \  
and  
returns a text string

Chat model that takes a list of  
messages  
as input  
and returns a message

# OUTPUT (PARSER)

LLM response to the  
Prompt

# MODELS

# Large Language Models



LLMs are powerful AI models trained on massive amounts of text data. They can **generate text, translate languages**, write different kinds of creative content, and answer your questions in an informative way.

They can handle various tasks like **summarization, question answering**, and text generation based on the prompt or input provided.

Chat models are specialized for **conversational interactions**. They take a sequence of messages (like a chat history) as input and generate a response that fits the conversation context.

They are designed to understand the **flow of conversation**, consider previous messages, and respond accordingly. They are ideal for building chatbots or virtual assistants that can engage in natural and engaging dialogues.

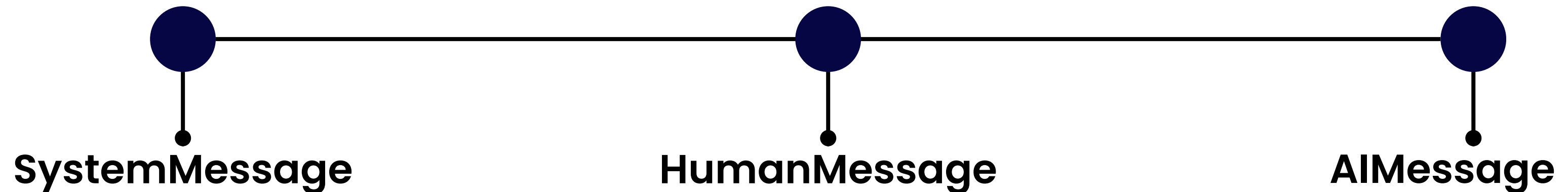
?

**When to  
Use  
Which?**

# MESSAGES

# Messages are fundamental units that represent the information exchanged within a conversation or workflow.

ChatModels take a list of messages as input and return a message.



This type is used for internal communication within the Langchain workflow. It doesn't directly appear in the conversation with the user but provides instructions or configuration details for the agent's behavior.

This type represents a message originating from the human user interacting with your conversational AI agent. It typically contains the user's query, request, or response within the conversation.

This type represents a message generated by the AI agent itself. It can be a response to a user's query, the output from a Langchain tool, or any information the agent provides during the interaction.



You are a witty and intelligent teenager.

Tell me about stars.



Ah, stars, those sparkling celestial beauties!  
They're like the glitter in the universe's cosmic confetti.

# PROMPTS



# Prompts

Prompts are crucial elements that guide Large Language Models (LLMs) and other tools towards generating the desired output. They act as instructions or inputs that set the context and parameters for the model's operation.

The objects that take user input and transform it into the final string or messages are known as "Prompt Templates".

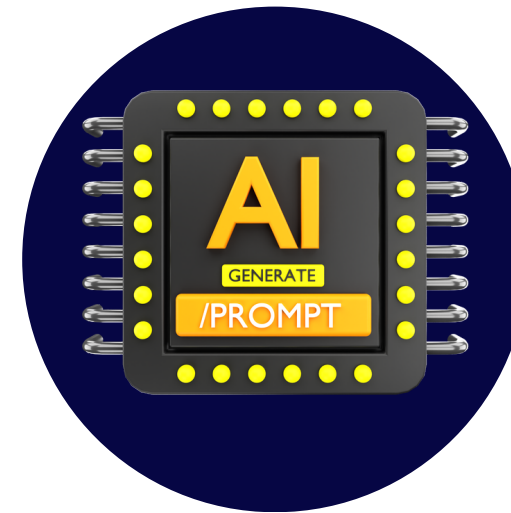


## PromptTemplate

create a template for a string prompt

```
from langchain.prompts import PromptTemplate
```

```
prompt = PromptTemplate.from_template("Suggest me a good company name for {agency}?")  
prompt.format(agency="digital marketinng service agency")
```



## ChatPromptTemplate

The prompt to chat models is a list of chat messages

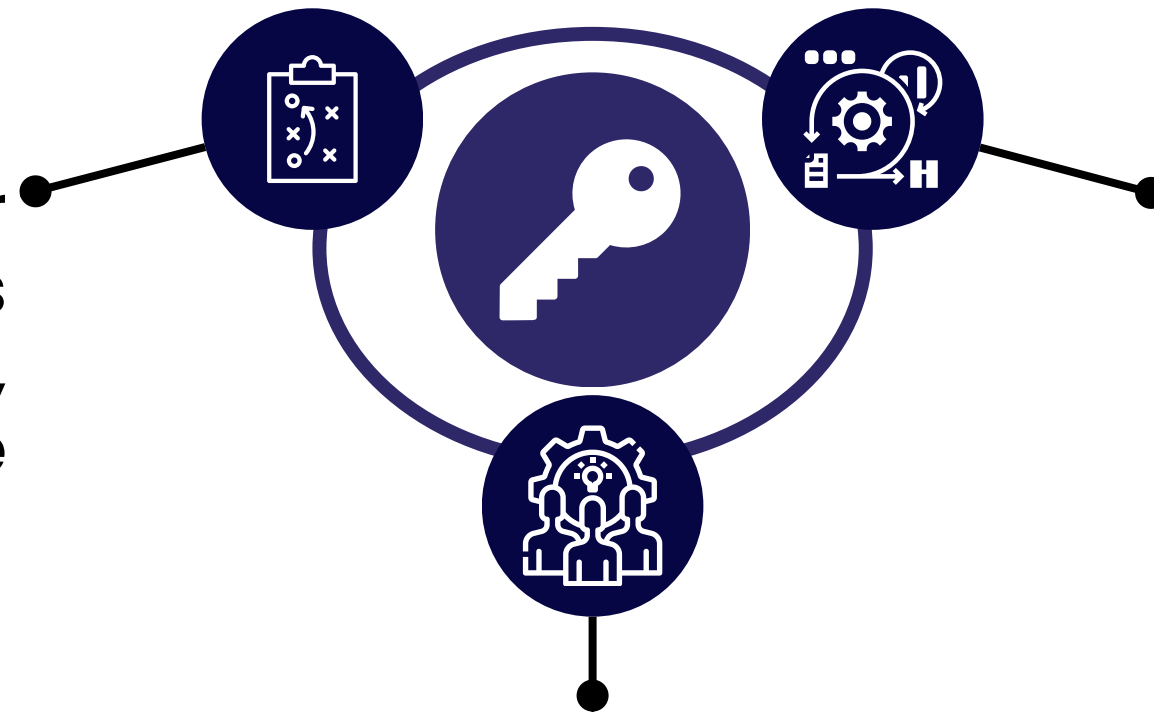
```
from langchain_core.prompts import ChatPromptTemplate
```

```
chat_template = ChatPromptTemplate.from_messages(  
    [  
        ("system", "You are a helpful AI assistant bot. Your name is {name}."),  
        ("human", "Hi, how are you?"),  
        ("ai", "I am doing good"),  
        ("human", "{user_input}"),  
    ]  
)
```

```
messages = chat_template.format_messages(name="Lily", user_input="What is your name?")
```

# BENEFITS OF GOOD PROMPTS

Improved Output Quality: Clear and well-designed prompts lead to more accurate, relevant, and informative responses from LLMs and tools.



Flexibility: Prompts allow you to adapt your workflows to handle various tasks and user interactions effectively.

Enhanced Control: By carefully crafting prompts, you can exercise greater control over the direction and style of the generated text.

# SELECTORS



- Imagine you have a large number of examples. Then you need to select which one to include in the prompt.
- Selectors act like filters, analyzing the prompts and choosing the one that best fits your needs.

Similarity

This selects examples based on cosine similarity to the inputs.

MMR (Maximal Marginal Relevance)

This approach balances similarity and diversity. It selects examples based on a combination of which examples have the most similarity to the inputs, but also maintain the diversity by penalizing them for their closeness.

Length

This selector chooses examples based on a desired length (e.g., selecting the shortest or longest response).

N-gram Overlap

This method focuses on the overlap of n-grams (sequences of n words) between the examples and input. The selector picks the response with the highest n-gram overlap.

Improved Output Quality

Reduced Manual Work

Fine-Tuning Output

# Few-shot prompt templates



- In `FewShotPromptTemplate`, we provide the examples to the LLM model. It will tell the LLM about the context and how we want the output to look like.

```
# Examples of a pretend task of creating antonyms.
```

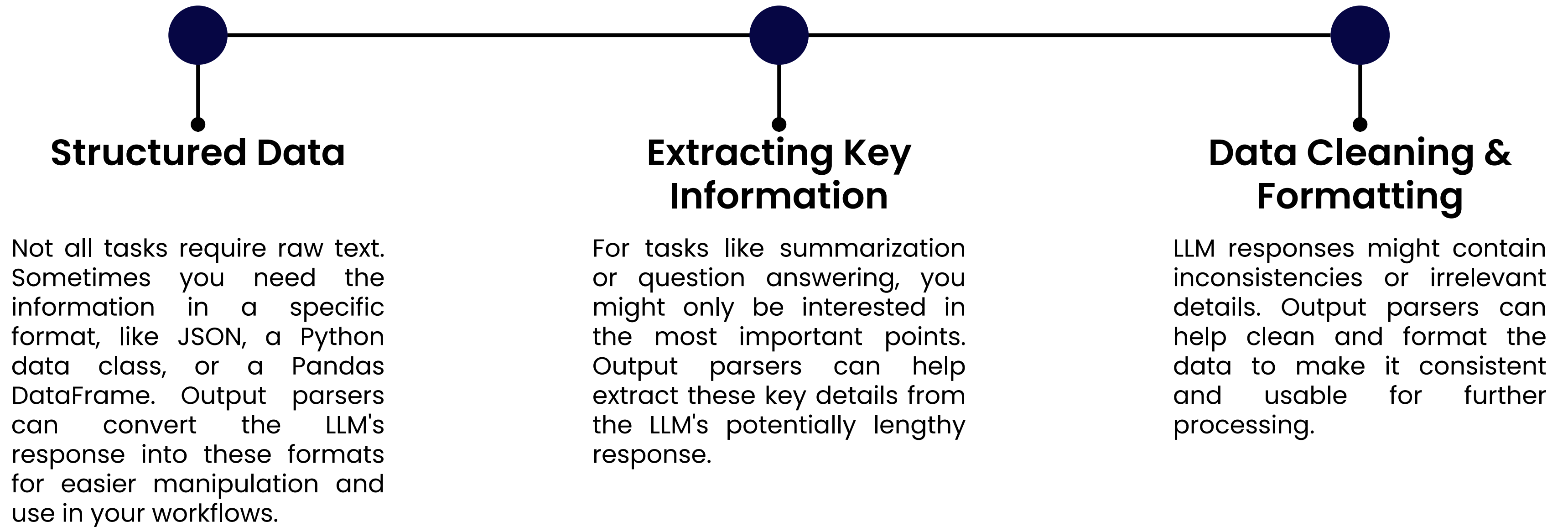
```
examples = [  
    {"input": "parrot", "output": "bird"},  
    {"input": "cow", "output": "animal"},  
    {"input": "snake", "output": "reptile"},  
    {"input": "frog", "output": "amphibian"},  
    {"input": "dog", "output": "mammal"},  
]
```

```
example_prompt = PromptTemplate(  
    input_variables=["input", "output"],  
    template="Input: {input} \nOutput: {output}",
```

# PARSERS

# OUTPUT PARSER

In LangChain, output parsers are specialized tools that take the raw text response generated by a large language model (LLM) and transform it into a more structured and usable format.



# Types of Output Parsers

## Pydantic Parser

Converts the LLM response into a user-defined Pydantic data class, allowing for structured data manipulation.

## JSON Parser

Transforms the response into a JSON object for easy integration with other tools and workflows.

## List Parser

Parses the output into a list of strings, useful for scenarios where the LLM generates multiple items.

## Datetime Parser

Extracts and formats dates or times mentioned in the response.

## Structured Parser (Basic)

This is a simpler option that parses the response into a dictionary structure, suitable for less complex data.

output parsers are a powerful tool in LangChain that bridges the gap between the raw LLM output and your desired data format. They allow you to effectively leverage the capabilities of LLMs for various tasks by structuring and extracting the information you need.