

# LangChain Cheat Sheet

Getting Started with LangChain (slides + demos)

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## Core idea

- LLM: predicts the next token in a sequence.
- LangChain: framework for building LLM-powered apps with reusable abstractions (models, prompts, parsers, chains).
- Typical session flow: slides → notebook demo → Q&A; (+ optional exercise).

## LangChain components

- **Models** - wrappers over LLM APIs (OpenAI, etc.).
- **Prompt Templates** - reusable prompts with variables ("{concept}").
- **Output Parsers** - convert raw model output into usable types (string/JSON/etc.).

### Pattern

```
chain = prompt | llm |  
StrOutputParser()  
chain.invoke({'text': '...'})
```

## Chains + LCEL

- A **chain** composes steps: prompt → model → parser.
- LCEL uses pipe syntax (|) to compose and reuse chains.

### Pattern

```
chain = prompt | llm | parser  
result = chain.invoke({'topic':  
'...'})
```

**Why it matters:** you can swap pieces (prompt/model/parser) without rewriting everything.

## LCEL: runnables

- Many LangChain building blocks implement the **Runnable** interface.
- Standard methods: **invoke**, **batch**, **stream** (async variants available).
- Inputs/outputs vary by component (prompt, model, parser, chain).

### OPTIONAL EXERCISE

Build a summarization chain with **prompt** | **model** | **parser** and test it on a paragraph from your notes.

Stretch: add **stream()** so you can see tokens arrive live.

# LangChain Cheat Sheet

RAG + Agents + Deployment

## RAG: chat over documents

- **RAG** = Retrieval Augmented Generation (LLM + relevant retrieved docs).
- Motivation: work around context-length limits and improve factual grounding.
- **Embeddings**: convert text into vectors so you can do similarity search.

**Pipeline**: Load → Split → Embed → Store → Retrieve → Generate

### TIP

If you remember one thing: embeddings let you find the most relevant chunks fast, then you feed only those to the LLM.

## Self-check

- **LangChain advantage**: unified abstractions that streamline development. (B)
- **Prompt template**: reusable prompt structure with variables. (B)
- **Embeddings in RAG**: vectors for similarity comparison. (B)
- **Agent loop**: iterative decide-act-observe cycle. (B)

## References

- [LangChain: Introduction](#)
- [LangChain: Agents Concepts](#)
- [LangGraph: Concepts](#)
- [Gen Agents \(arXiv:2304.03442\)](#)

## Agents

- **Agent loop**: decide action → call tool(s) → observe results → repeat until done.
- **Tools**: functions with an input schema + execution logic (often bundled in toolkits).
- Use agents when you need routing decisions, tool selection, and iterative problem solving.

## LangGraph

- Models agent workflows as **graphs**.
- **Nodes**: functions that update a shared **State**.
- **Edges**: fixed/conditional transitions (routing logic).
- Designed for controllability, persistence (human-in-the-loop), and streaming-first UX.

## LangServe

- Deploy chains/runnables as a REST API (FastAPI + Pydantic validation).
- Common endpoints: **/invoke**, **/batch**, **/stream**.
- Includes a **/playground** UI; optional tracing to LangSmith.

### OPTIONAL EXERCISE

Build a simple RAG pipeline (load → split → embed → store), then answer questions about a PDF.