

Summary of topics

Saturday, 3 May 2025 11:37 PM

Agent:

- A control flow mechanism where the LLM itself decides the execution path
- Features bi-directional processing with potential cycles
- Allows for more dynamic decision-making capabilities

Chaining:

- Takes the output of one language model and feeds it as input to another
- Creates a composed sequence of LLM calls
- Follows a unidirectional flow (left to right)
- Example: Retrieval Augmented Generation (RAG)
 - First LM processes the original question
 - Question gets embedded and similar documents are retrieved
 - Original prompt is augmented with retrieved information
 - Final LM generates the answer based on enhanced context

Router:

- A specialized chain that uses an LLM to determine execution paths
- Leverages the LLM's reasoning ability to decide which branch to follow
- Can direct flow to different operations (e.g., database search vs. web search)
- Provides increased flexibility in building complex AI systems

Key Distinction:

- Chains: Linear, predetermined flow of operations
- Agents: Dynamic flow with decision-making capabilities and potential cycles

Function Calls:

- A feature allowing LLMs to invoke external tools/functions
- Process:
 - Developer provides function descriptions (title, purpose, arguments, return values)
 - LLM determines when function use is appropriate based on the query
 - LLM specifies which functions to call with what arguments
 - Backend executes the functions and returns results
- Can be implemented using decorators for simplified orchestration

Agent Fundamental Design:

- Simple algorithmic flow:
 1. Start with user query
 2. LLM decides if a tool is needed (API call, database query, etc.)
 3. If tool is needed, execute it with LLM-specified arguments
 4. Return tool results to the LLM
 5. LLM either uses another tool or provides final answer to user
- Creates a loop of tool use and reasoning

RAG (Retrieval Augmented Generation):

- Enhances LLM responses by incorporating external knowledge
- Process:
 1. Receive user query
 2. Retrieve relevant information from external knowledge base
 3. Augment prompt with retrieved context
 4. Generate response using both query and retrieved information
- Helps overcome knowledge limitations and hallucinations

LLM.txt:

- A pattern for interfacing with language models through structured text
- Enables complex interactions via simple text-based input/output
- Standardizes how applications communicate with language models

ReAct in LangChain:

- Combines reasoning and action in agent workflows
- Process:
 1. Reasoning: LLM thinks through the problem
 2. Action: Executes tool-based reasoning

2. Action: Executes tools based on reasoning
 3. Observation: Processes results from actions
 4. Repeats cycle until completion
- Creates more effective agents through structured thought-action patterns

Pipfile:

- Pipenv manages 2 files pipfile - declares **what** packages your project needs and **Pipfile.lock** – **locks down exact versions** of all packages (and sub-dependencies) that should be installed.
- Pipfile.lock is important:
 - Ensures Reproducible Environments
 - Protects Against Breaking Changes
 - Improves Install Performance
 - Safer Deployment

MCP (Model Completion Protocol)

MCP is a structured framework for prompting language models with clear conventions for input/output formatting and function calling.

MCP follows specific conventions:

- Uses delimiters to separate different parts of the prompt
- Defines structured formats for system messages, user inputs, and model responses
- Creates standardized ways to specify tool use and function calls