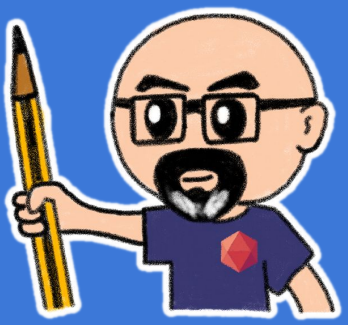
Codeurs en Seine

MCP Servers: Good Practices, Design Choices and Consequences



Horacio González 2025-09-23

clever cloud

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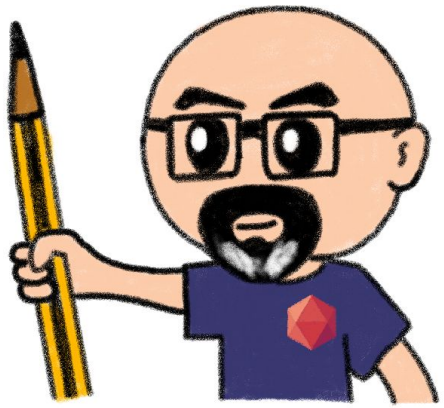
Who are we?

Introducing myself and introducing Clever Cloud



Horacio Gonzalez

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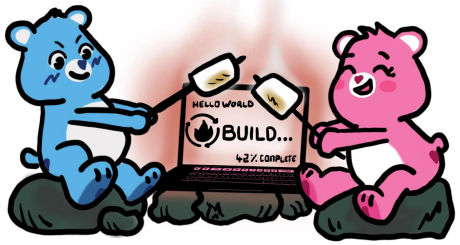
Spaniard Lost in Brittany

Finist Devs



Head of DevRel

clever cloud

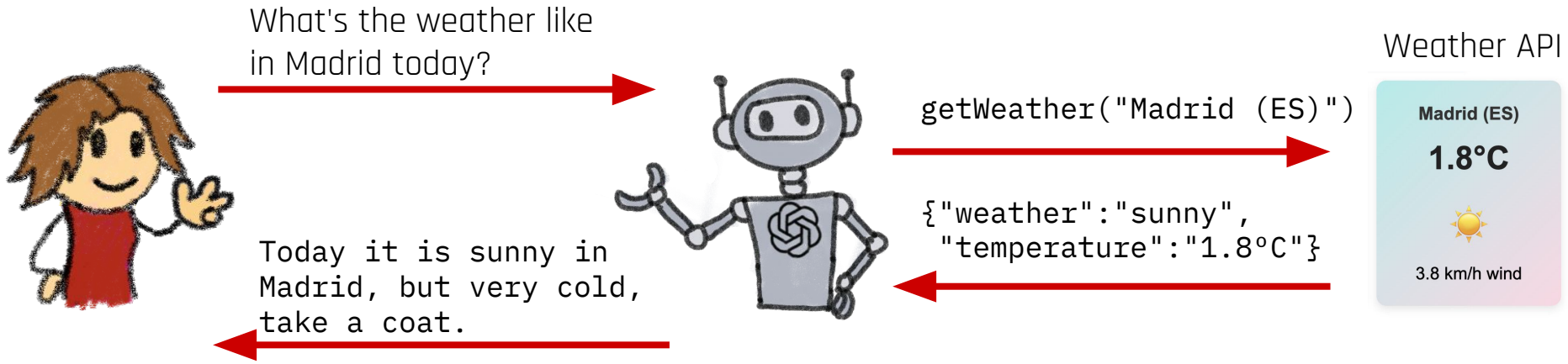
Clever Cloud

From Code to Product

clever cloud

LLM evolution

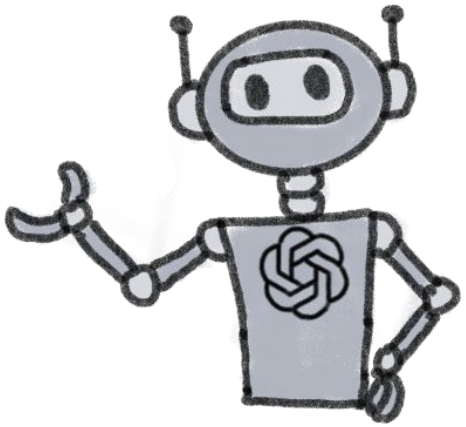
From simple chat to tool-enhanced agent!



LLM are only language models

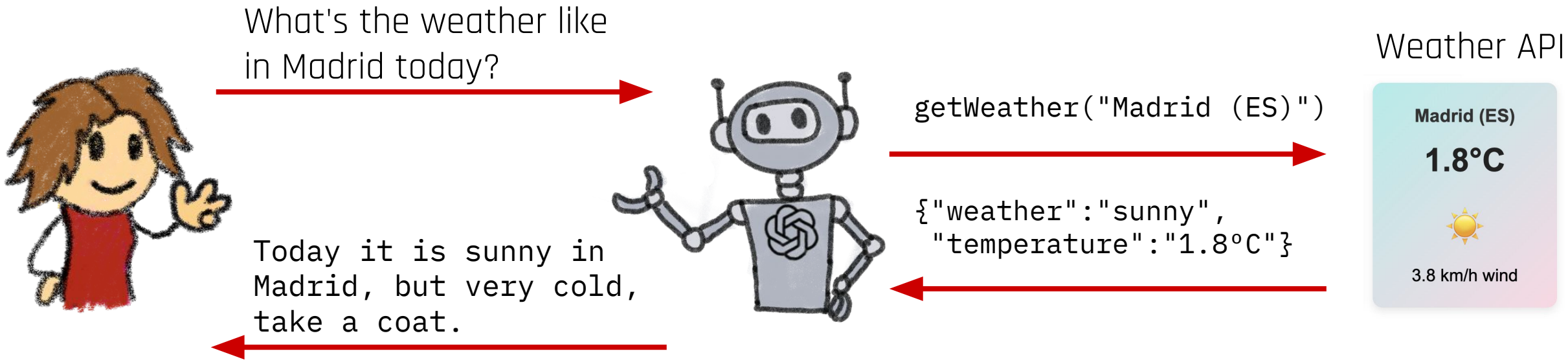
What's the weather like in Madrid today?

```
I'm unable to provide real-time
information or current weather updates.
```



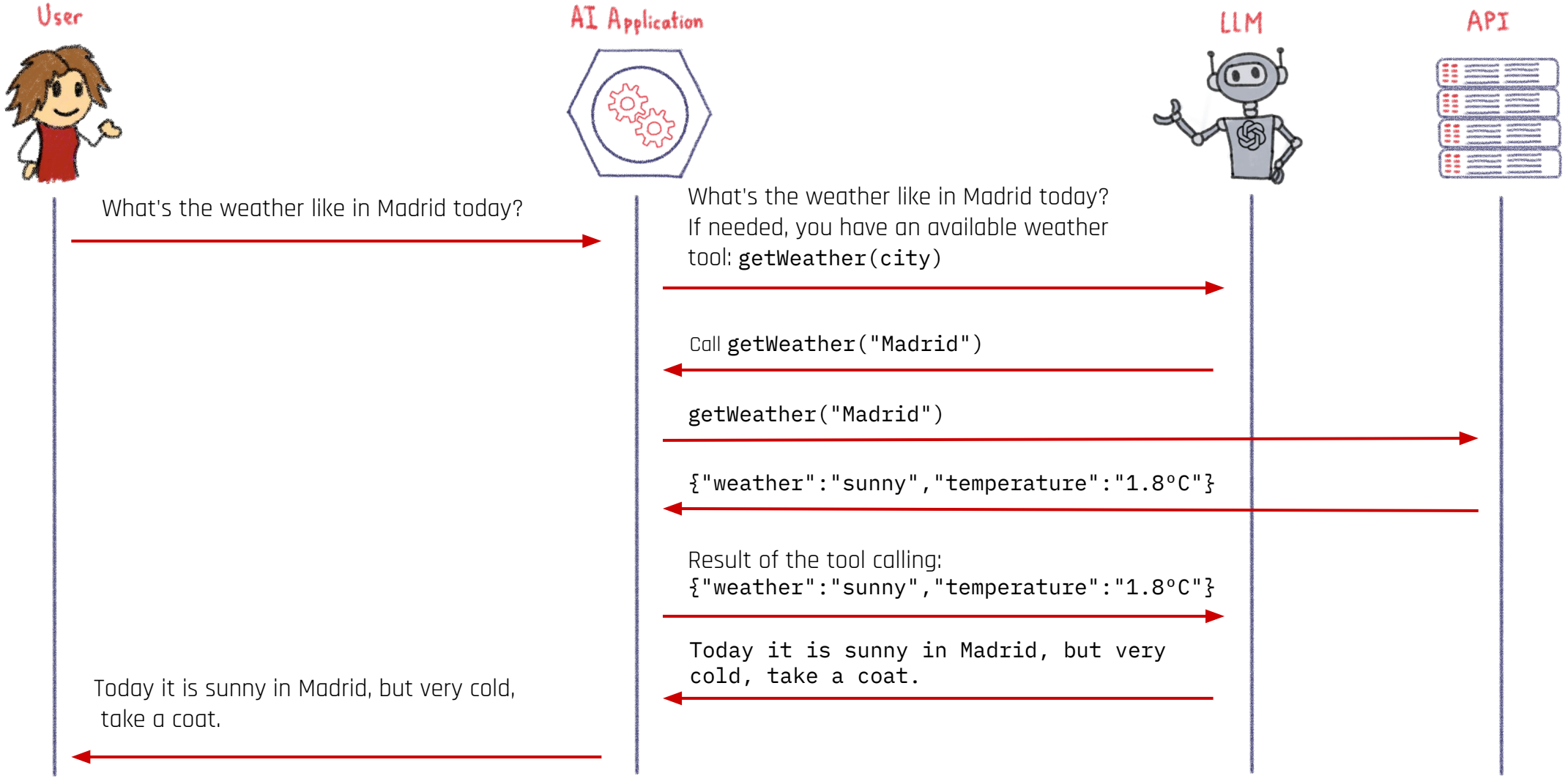
They have no built-in way to use external tools or real-time data

Tools and plugins were added



LLM recognizes it needs an external function and calls it, integrating the result into a natural-language response.

LLM don't call directly those tools



How are those LLM Tools defined?

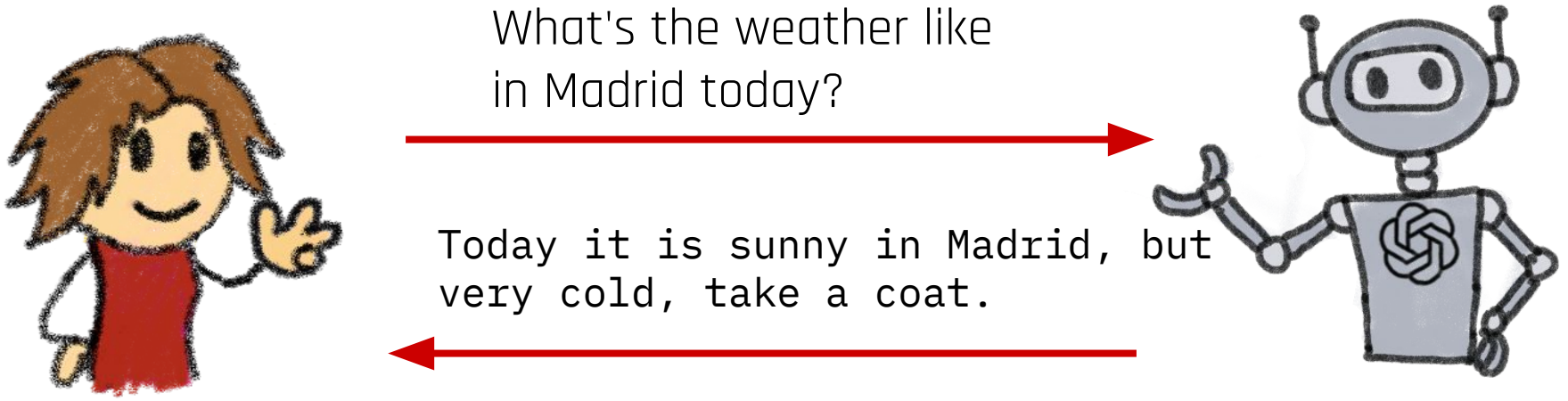
LyingWeatherTool.java

```java
//DEPS dev.langchain4j:langchain4j:1.0.0-beta1

import dev.langchain4j.agent.tool.Tool;

public class LyingWeatherTool{
 @Tool("A tool to get the current weather in a city")
 public static String getWeather(String city) {
   return "The weather in " + city + " is sunny and hot.";
 }
}
```

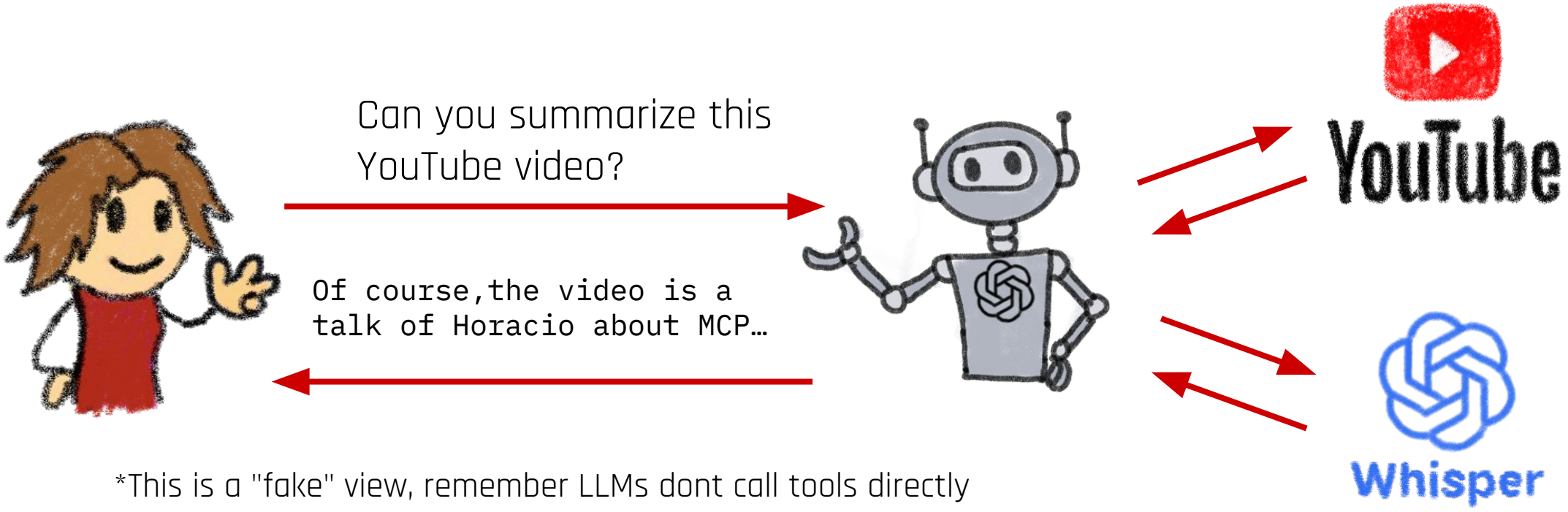
Here in Java using LangChain4j

Why this matters?

- Moves LLMs from static text generation
  - dynamic system components
- Increases accuracy & real-world usability
- Allows developers to control what the LLM can access



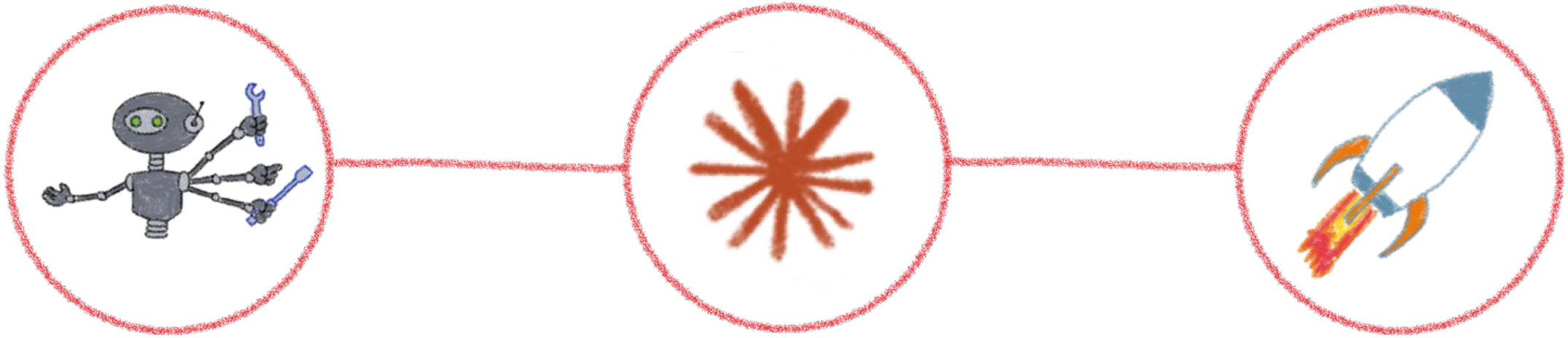
From LLM chats to LLM-powered agents



*This is a "fake" view, remember LLMs dont call tools directly
But it's the view from the Point of View of the user

LLMs act like an agent that can plan actions:
search the web, run some code, then answer

The rapid evolution of agents



Before MCP (2023–November 2024)

- Agents == niche LangChain, bespoke APIs, Copilot experiments…
- No standard way to connect LLMs to tools.

MCP Introduced (Nov 2024)

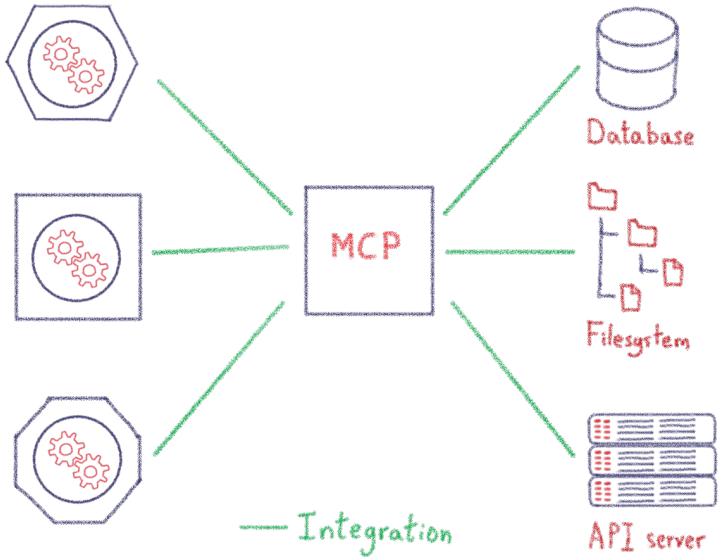
- Anthropic launches Model Context Protocol.
- Vendor-neutral, open standard for connecting LLMs.

The explosion (2025)

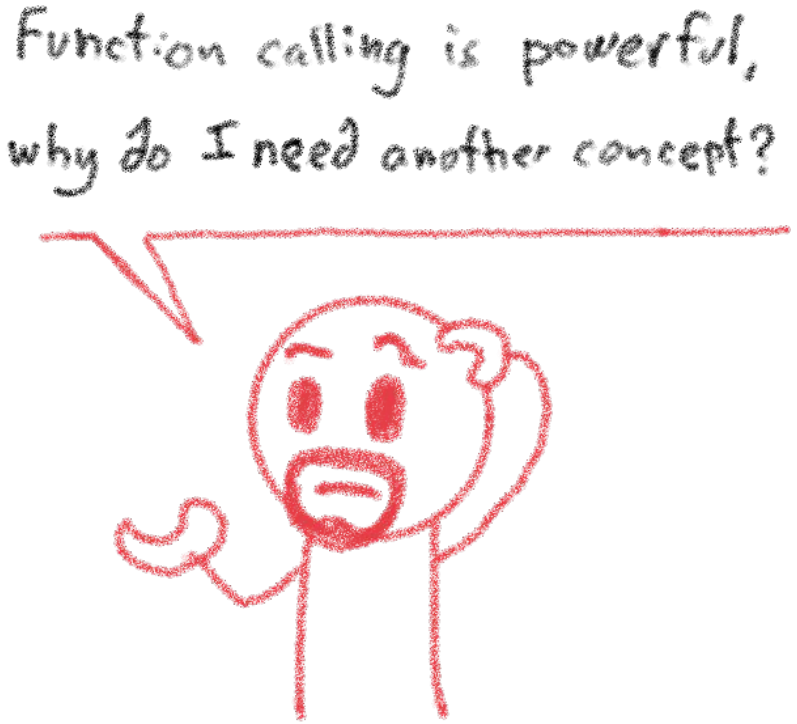
- Agents go mainstream: runtimes, orchestration, enterprise adoption.
- MCP reframed as the interoperability layer for agents.

Model Context Protocol (MCP): The missing link

MCP bridges LLMs with your applications, enabling controlled, real-world interactions



Why Do We Need MCP?

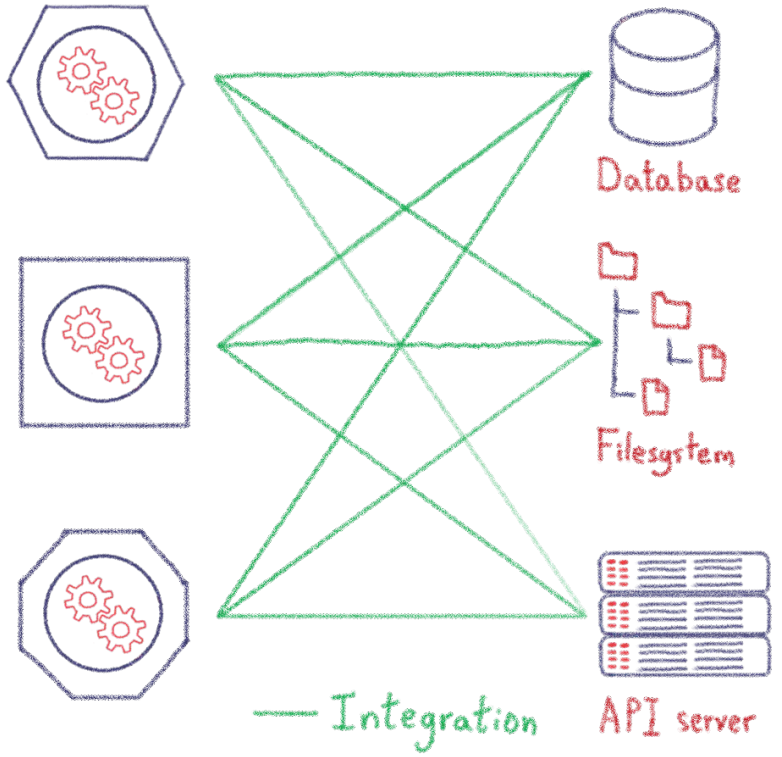


LLM function calling is useful,
but it lacks structure

Why Do We Need MCP?

Problem

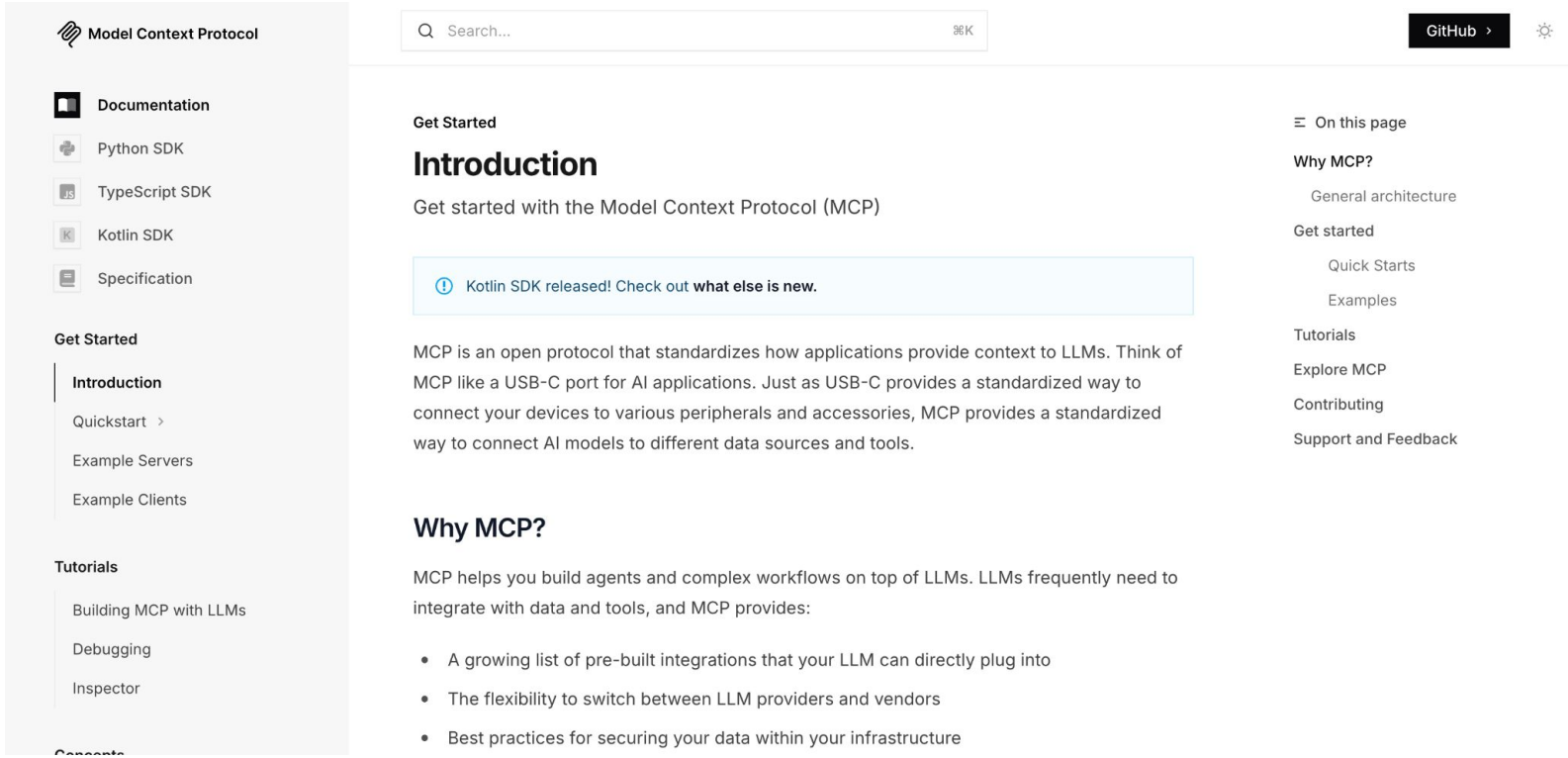
- LLMs **don’t automatically know** what functions exist.
- **No standard way** to expose an application's capabilities.
- **Hard to control** security and execution flow.
- Expensive and fragile **integration spaghetti**



Model Context Protocol

Anthropic, November 2024:
LLMs intelligence isn't the bottleneck, connectivity is

Model Context Protocol



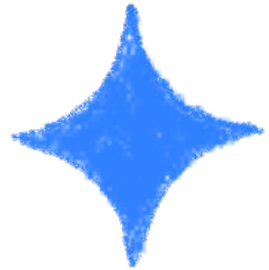
De facto standard for exposing system capabilities to LLMs

https://modelcontextprotocol.io/

The MCP Landscape Today

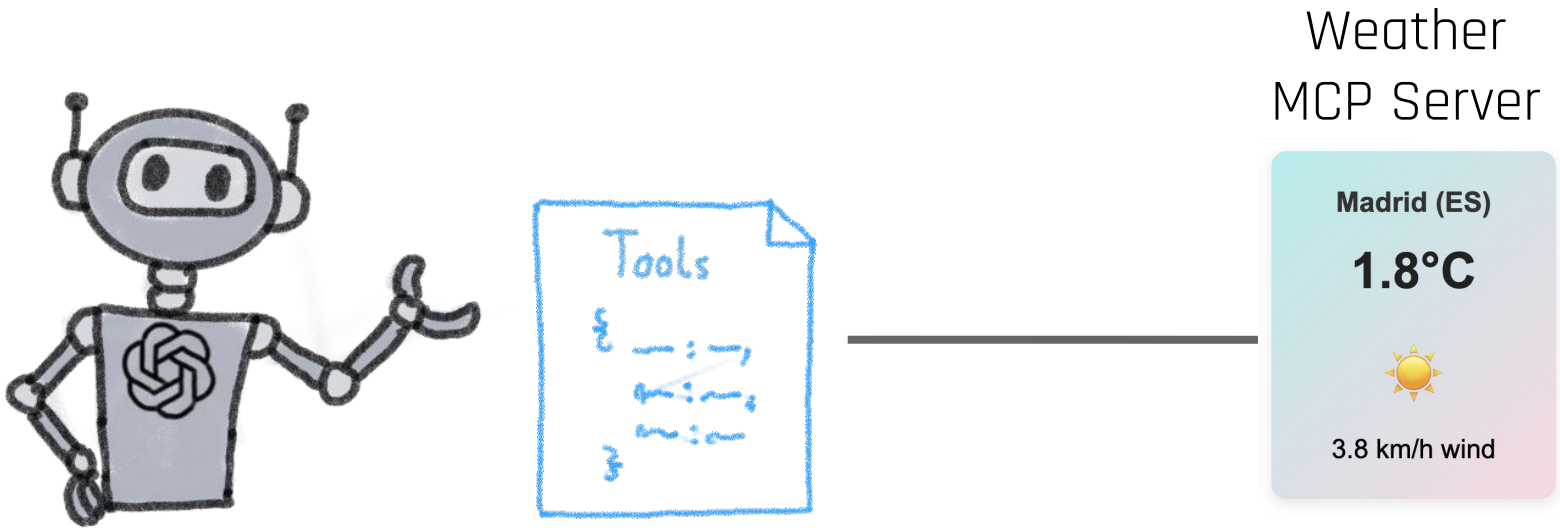
Major players adopted MCP:

- **Anthropic** – Originator and tool provider (Claude Desktop, SDKs).
- **OpenAI** – Agent SDK, ChatGPT Desktop, Responses API.
- **Google DeepMind** – Gemini support and tooling.
- **Microsoft / GitHub** – Copilot Studio, Azure, Office integration, C# SDK.
- **Developer Platforms** – Replit, JetBrains, Sourcegraph, TheiaIDE.
- **Enterprise / Services** – Block, Stripe, Cloudflare, Baidu Maps.
- Thousands of MCP servers live.



How MCP works

- Applications define an MCP manifest (structured JSON).
- The manifest describes available functions, input/output formats, and security policies.
- LLMs can discover and request function execution safely.

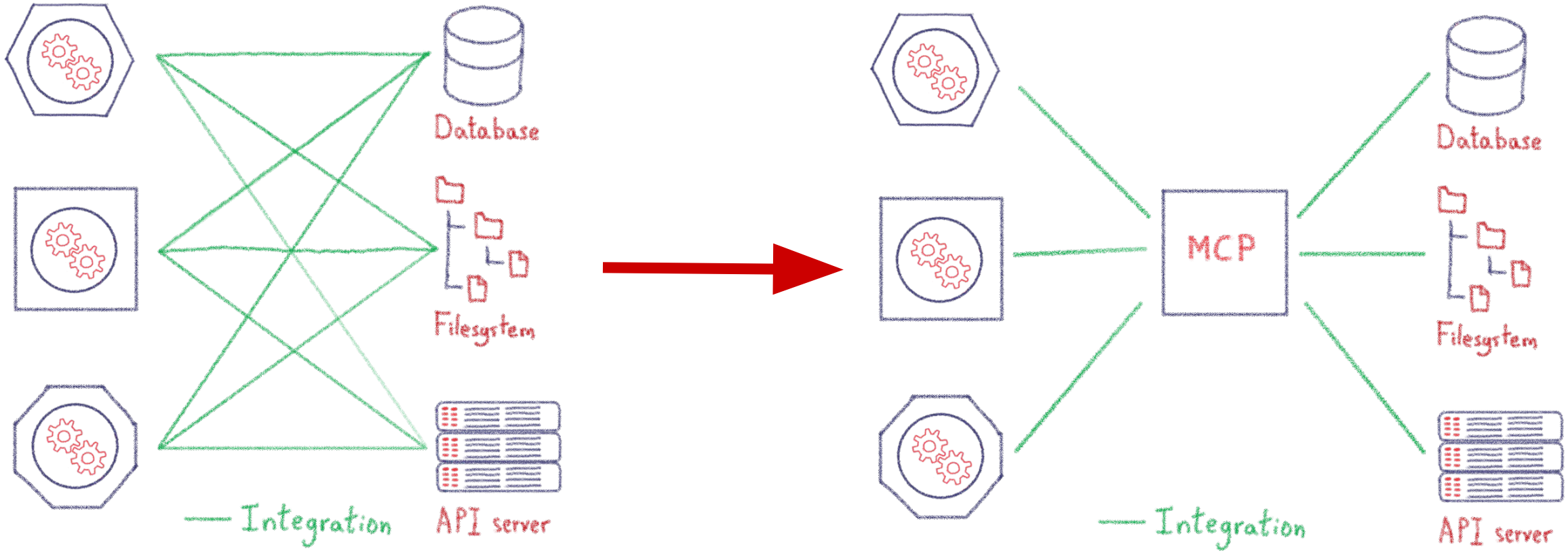


MCP is provider-agnostic

Works with any LLM provider

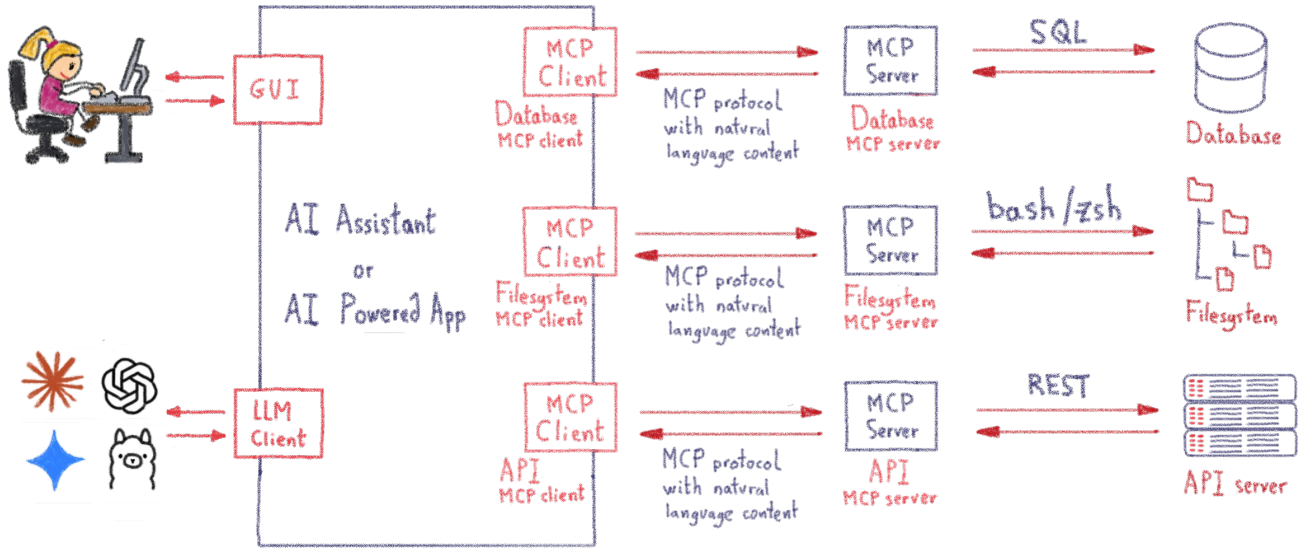
Ensures standardized function exposure
across platforms

MCP solves integration spaghetti

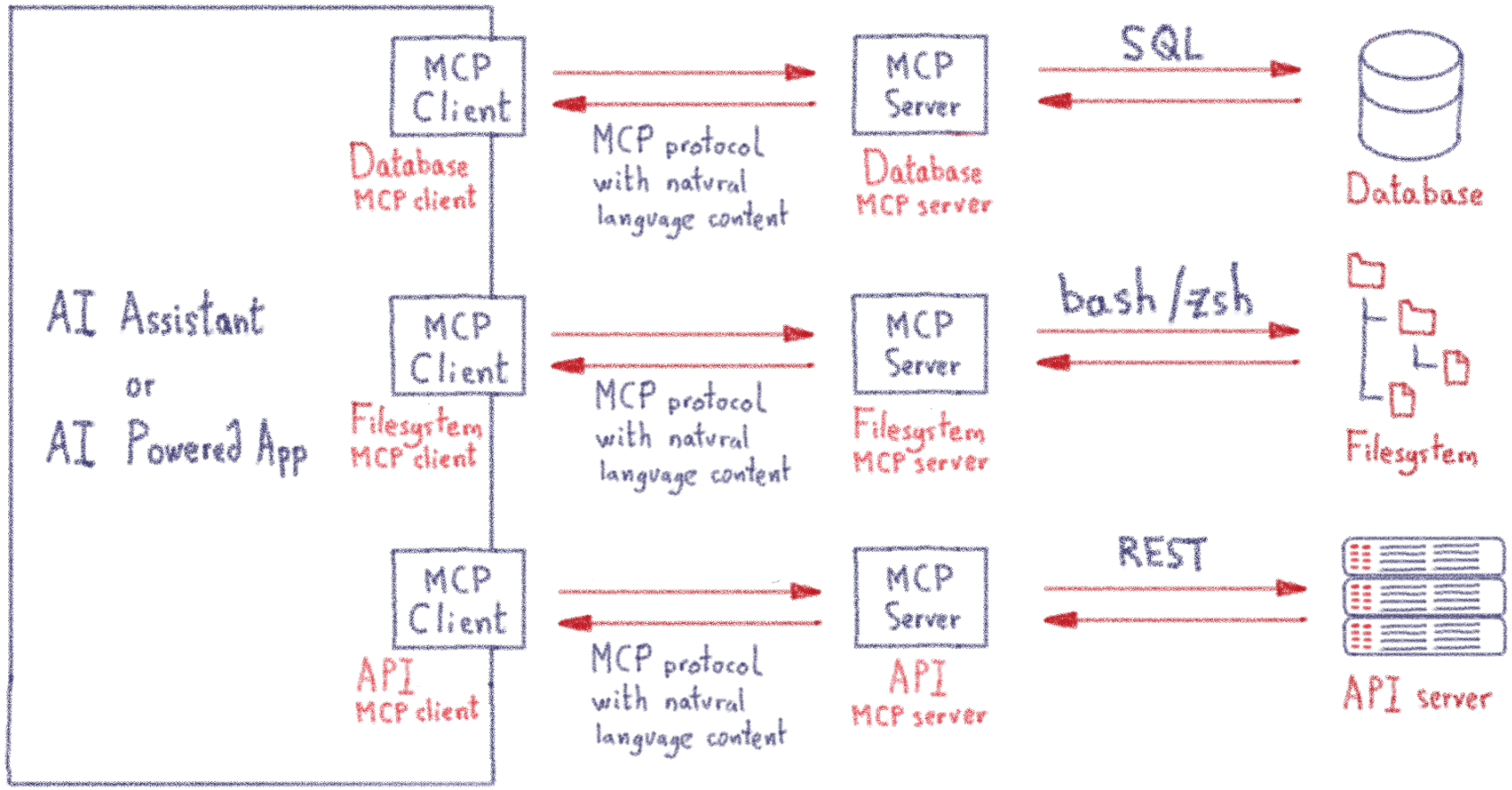


The architecture of MCP

Clients, servers, protocol and transports
Tools, resources and prompts



MCP Clients: on the AI assistant or app side



One MCP client per MCP Server

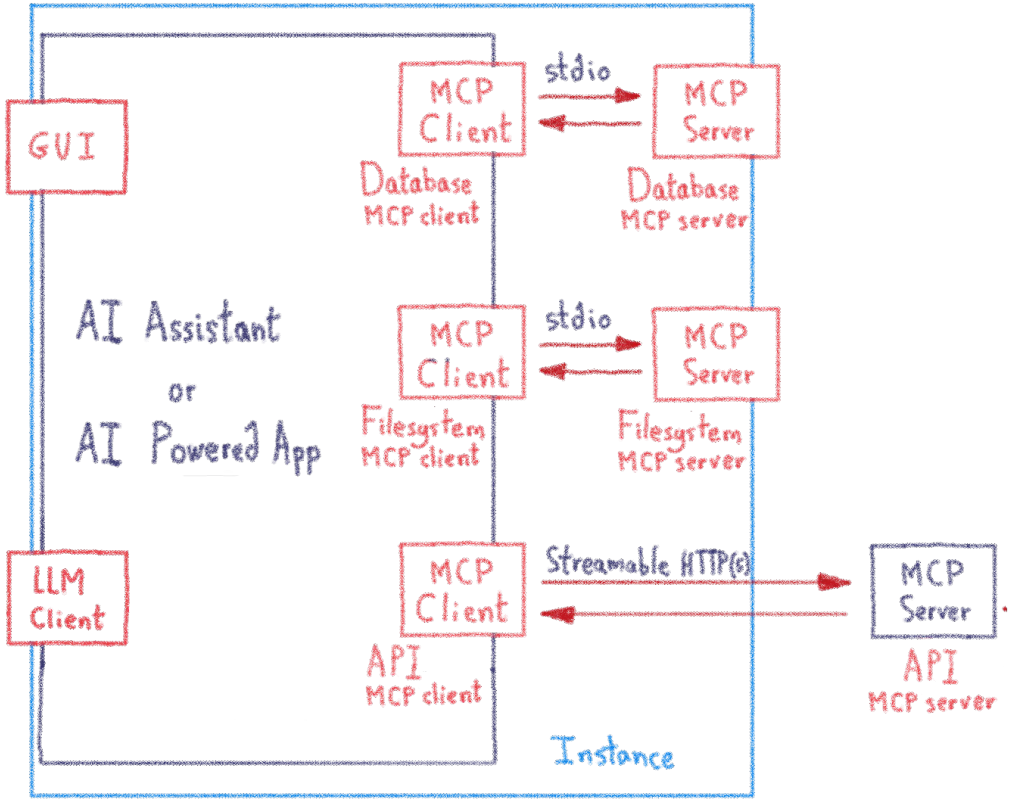
MCP Protocol & Transports

MCP Protocol

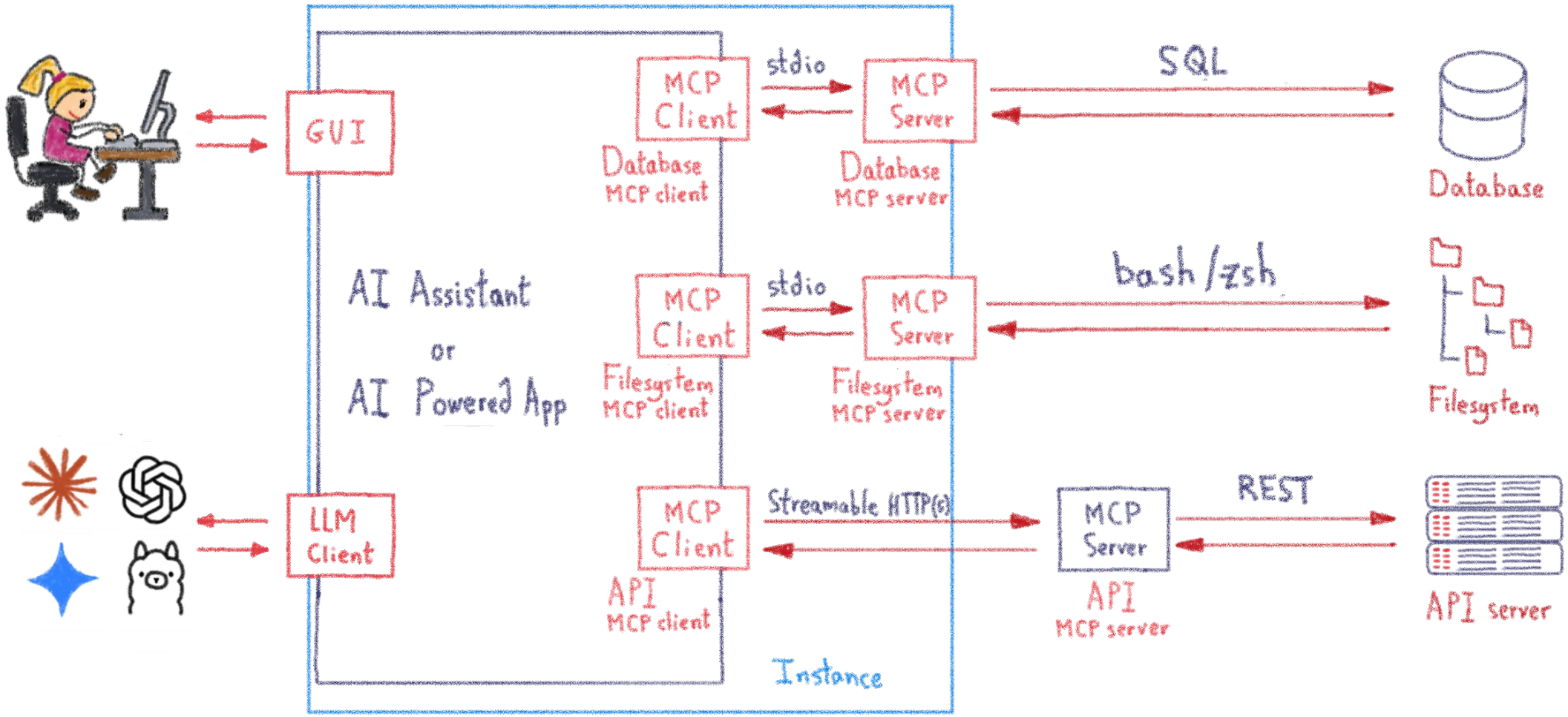
Follow the JSON-RPC 2.0 specification

MCP Transports

- STDIO (standard I/O)
  - Client and server in the same instance
- HTTP with SSE transport (deprecated)
- Streamable HTTP
  - Servers SHOULD implement proper authentication for all connections

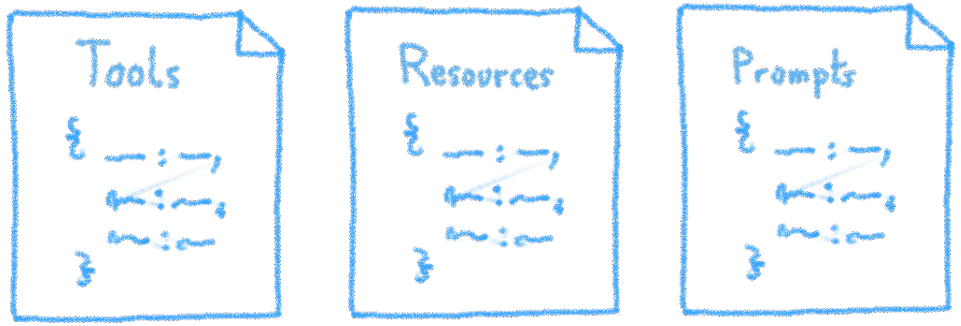


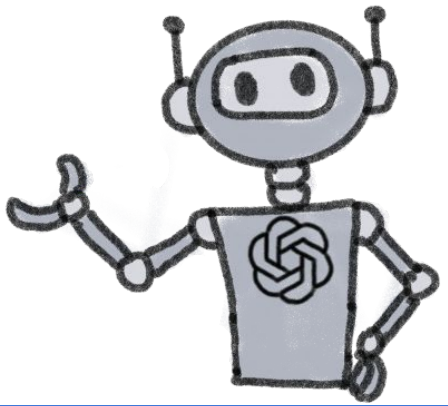
Full MCP architecture



Services: tools, resources & prompts

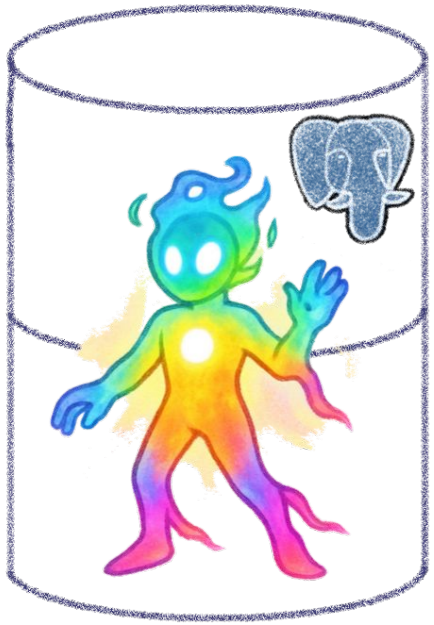
- Tools
  - Standardized way to expose functions that can be invoked by clients
- Resources
  - Standardized way to expose resources to clients
  - Each resource is uniquely identified by a URI
- Prompts
  - Standardized way to expose prompt templates to clients
  - Structured messages and instructions for interacting with LLMs





MCPs are APIs

And they should be architectured in a similar way



Developer Expectations Have Shifted

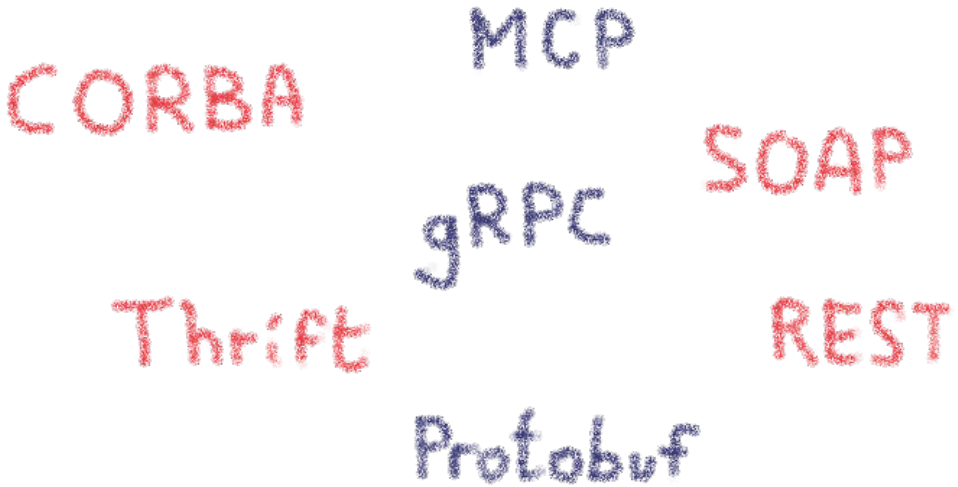
Winter 2024–2025

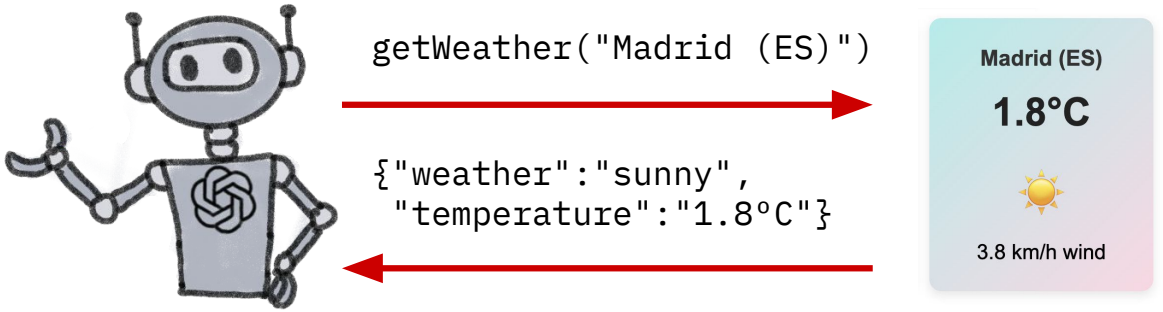
- “What is MCP?”
- “How do I connect my DB?”

Summer 2025

- “How do I build smarter MCP servers?”
- “How do I secure them?”
- “How do they fit into agent workflows?”

MCP Servers: APIs for LLMs





All those API technologies define protocols for communication between systems

Let's use an example: RAGmonsters

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🧩 RAGmonsters Dataset

Overview

The RAGmonsters dataset is a collection of 30 fictional monsters created specifically for demonstrating and testing Retrieval-Augmented Generation (RAG) systems. Each monster is completely fictional and contains detailed information that would not be found in an LLM's training data, making it perfect for showcasing how RAG can enhance an LLM's knowledge with external information.

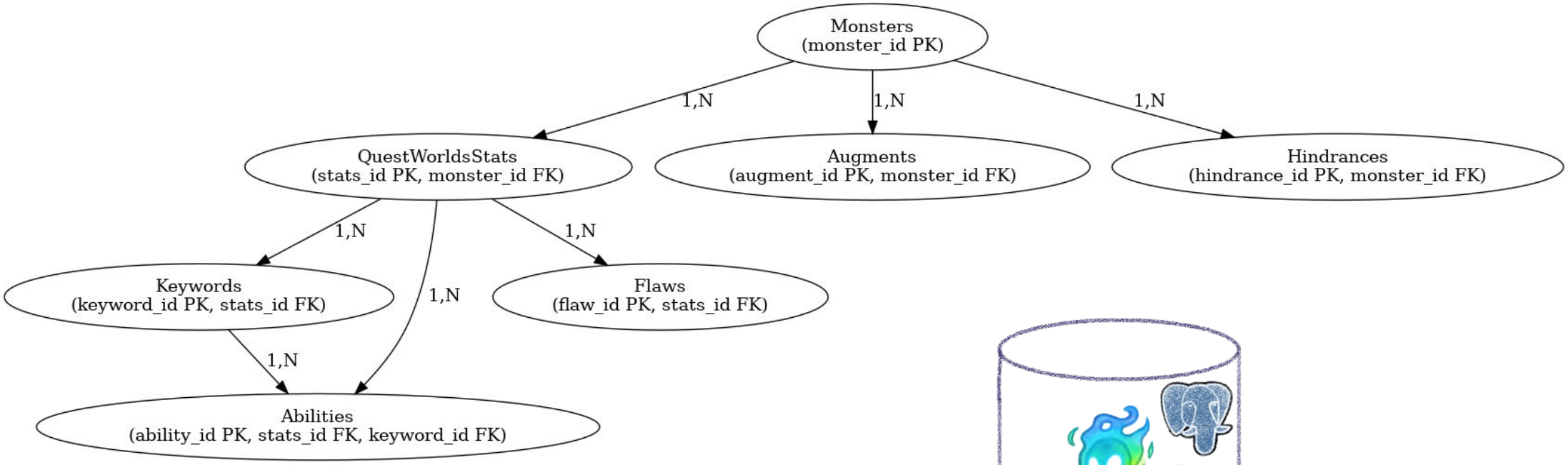
Purpose

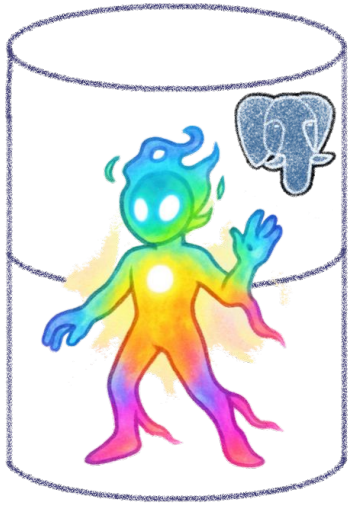
This dataset serves several educational purposes:

1. **Demonstrates RAG Value**: Shows how RAG can provide accurate answers about topics not in the LLM's training data
2. **Tests Retrieval Quality**: The varied attributes and relationships allow testing of different retrieval methods
3. **Supports Advanced Features**: Perfect for demonstrating filtering, re-ranking, and hybrid search techniques
4. **Provides Engaging Content**: Makes learning RAG concepts more fun and memorable

https://github.com/LostInBrittany/RAGmonsters

RAGmonsters PostgreSQL Database



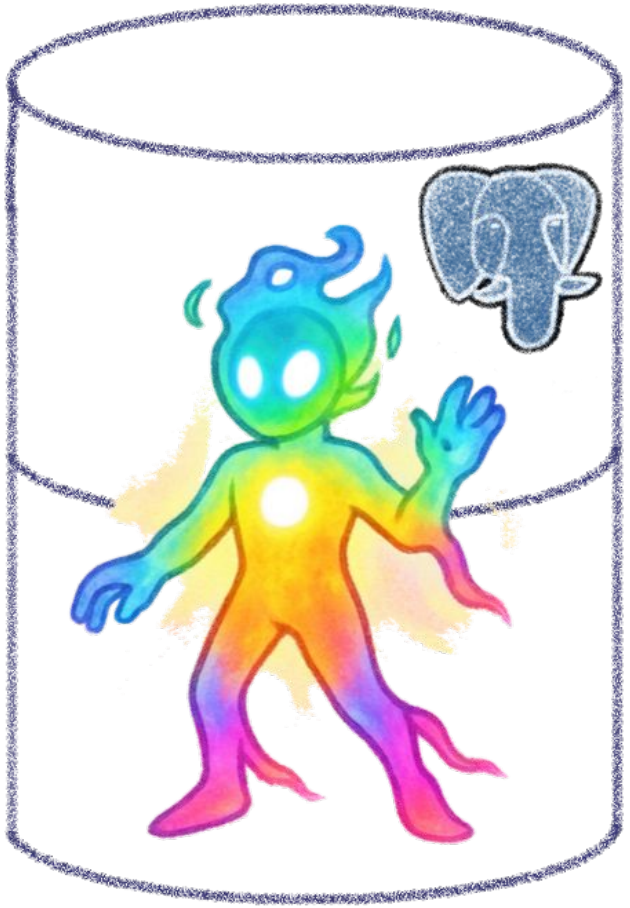


We want to allow LLM request it

Two options:

- A quick and dirty MCP server based on PostgreSQL MCP server
- A custom-made MCP server tailored for RAGmonsters

Which one to choose?

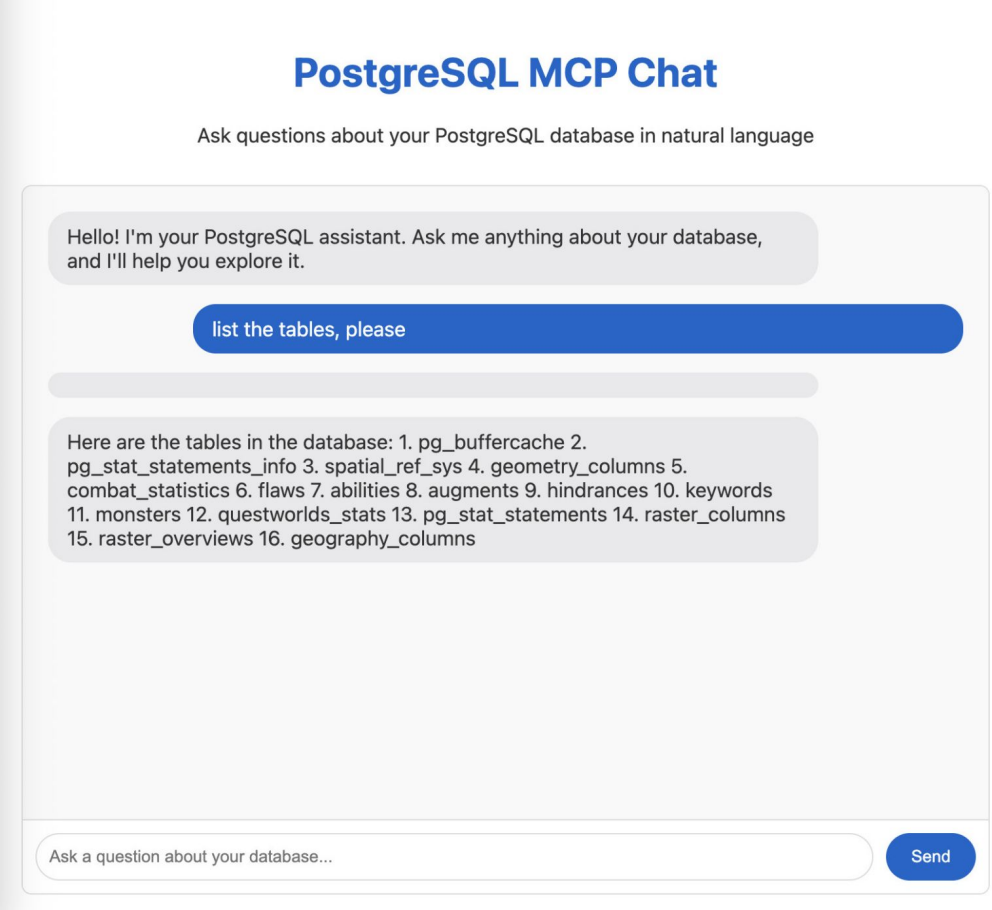


Generic PostgreSQL MCP server

Using PostgreSQL MCP Server

- A Resource that give the table schema for tables: `/schema`
- A Tool that allows to do SQL queries: `query`

LLM can know what tables do we have and what is their structure, and it can request them



Implementation:
https://github.com/CleverCloud/mcp-pg-example
PostgreSQL MCP Server:
https://github.com/modelcontextprotocol/servers/tree/main/src/postgres

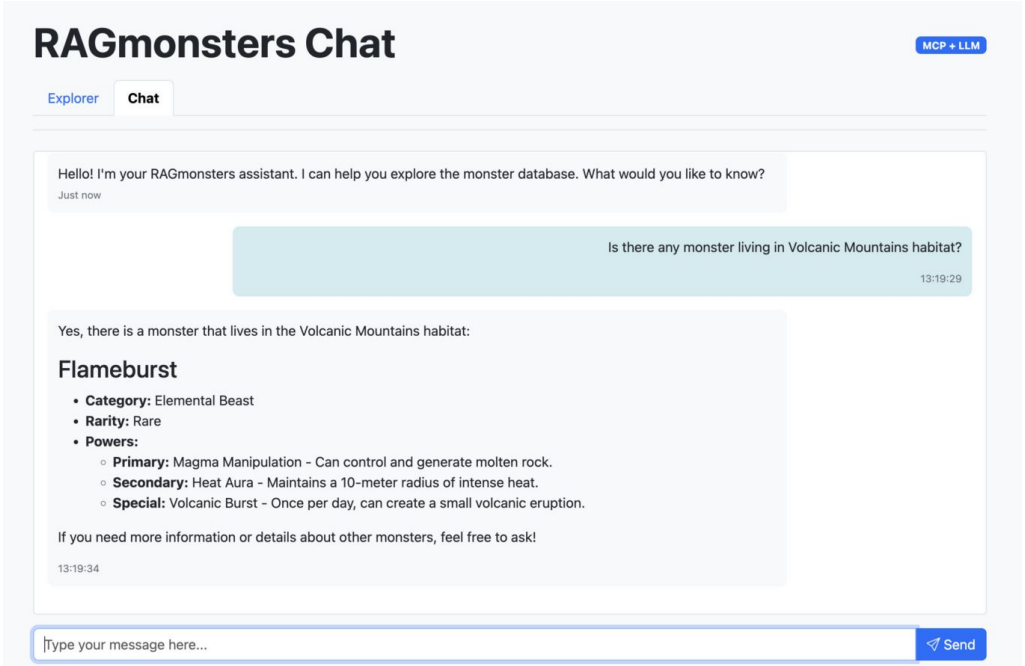
Custom-made RAGmonsters MCP server

Coding a MCP server for it. It offers targeted tools:

- `getMonsterByName`: fetches detailed information about a monster.
- `listMonstersByType`: Lists monsters of a given type.

- Easy, intuitive interactions for LLMs.
- Optimized for specific use cases.
- Secure (no raw SQL).

Implementation:
https://github.com/LostInBrittany/RAGmonsters-mcp-pg

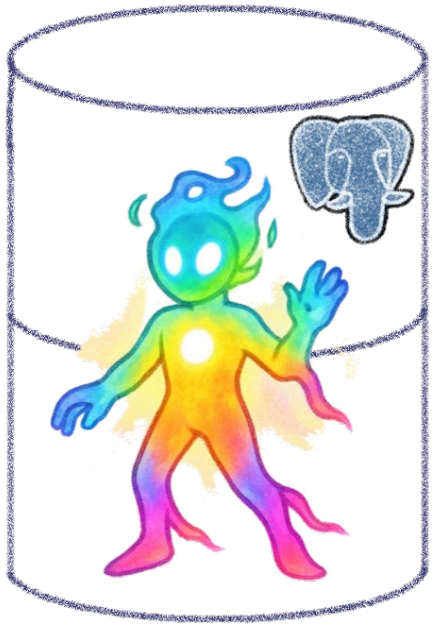


How to choose?

Aspect	Generic MCP Server	Domain-Specific MCP Server
Setup Speed	Fast, minimal configuration	Slower, requires planning
Efficiency	Lower, LLM must explore schema	High, optimized for specific tasks
Security	Risk of SQL injection	Secure, predefined tools
Flexibility	Adapts to any schema	Needs updates with schema changes
User Experience	Complex, LLM must learn	Simple, guided interactions

But how to do it?

Some down-to-Earth, practical advices



Design principles

What “good” looks like

- **Narrow, named capabilities**
  each tool should read like a product verb: `getMonsterByName`, `listMonstersByType`, `compareMonsters`.
- **Stable types in/out**
  explicit schemas (IDs, enums, unions) so the agent can plan reliably.
- **Deterministic behavior**
  same inputs → same outputs; include idempotencyKey when making state changes.
- **Least privilege**
  tools do one thing; internal queries/side-effects are not exposed.
- **Guardrails at the edge**
  validate inputs, clamp result sizes, redact PII, enforce authZ inside the server.

Capability modeling

Turn “tasks” into MCP tools/resources/prompts

Tools (actions)

- Read: `getMonsterByName(name) -> Monster`
- List: `listMonstersByType(type, limit=25, cursor?) -> {items:[Monster], nextCursor}`
- Search: `searchMonsters(q, limit=10) -> [MonsterSummary]`

Resources (documents/URIs the client can browse/fetch)

- `ragmonsters://schema/Monster` (JSON schema for types)
- `ragmonsters://docs/query-tips` (compact usage notes)
- `ragmonsters://images/{monsterId}` (read-only asset stream)

Prompts (reusable instructions/templates)

- prompt://ragmonsters/answering-style (tone, do/don’t)
- prompt://ragmonsters/disambiguation (ask for missing fields first)

Input contracts

Make the LLM succeed on the first try

- Refer enums & unions for fields the model tends to invent
  `type ∈ {BEAST, ELEMENTAL, UNDEAD,…}`
- Add optional "`reason`"/"`intent`" fields that your server ignores functionally but logs for eval
- Hard limits at the boundary: `limit ≤ 50, name.length ≤ 64, q.length ≤ 120`

```
{
 "type": "object",
 "required": ["type"],
 "properties": {
   "type": {"enum": ["BEAST","ELEMENTAL","UNDEAD","CELESTIAL","HUMANOID"]},
   "limit": {"type":"integer","minimum":1,"maximum":50},
   "cursor": {"type":"string"}
 }
}
```

Output shape

Make it composable

Always return a machine part and a human part:

- `data`: typed payload the client/agent can chain.
- `summary`: 1–2 lines the model can quote.
- `next`: cursors or suggested follow-ups.

```
{
 "data": { "items": [ { "id":"glowfang", "type":"BEAST", "danger":3 } ],
"nextCursor":"abc123" },
 "summary": "Found 1 beast: Glowfang (danger 3)." ,
 "next": ["getMonsterByName('glowfang')" ]
}
```


Security & governance

Baked into the server

- **AuthN**: accept a caller token; map to user/roles inside your server.
- **AuthZ**: per-tool role checks (viewer, editor, admin).
- **Data scope**: inject row-level filters (tenant, project) before hitting storage.
- **Rate limits**: e.g., 60 rpm per user; lower for heavy tools.
- **Redaction**: never return secrets; hash IDs in logs.
- **Explainability**: include source/policy notes in responses where relevant.

Observability & evaluation

From the beginning

- **Structured logs**
  `{tool, userId, durationMs, ok, errorCode}`

- **Traces**
  around datastore calls; record row counts

- **Golden tasks**
  keep a small suite (10–20) of representative prompts; run nightly

- **Safety tests**:
  prompt-injection set, over-broad queries, boundary limits

Conclusion

- Generic MCP servers
  Quick to set up, flexible, but less efficient and more error-prone.
- Domain-specific MCP servers
  Safer and faster for targeted tasks, but need more upfront design.

Choose wisely
Use generic for exploration, domain-specific for production.

A bit like for REST APIs, isn't it?

The road ahead

- MCP is quickly becoming the *lingua franca* of agents.
- We’re still early — best practices are being shaped right now.
- Your design choices today will set the tone for secure, scalable agent ecosystems tomorrow.

That's all, folks!

Thank you all!