

# SNOWCAMP



## MCP Servers: Good Practices, Design Choices and Consequences

Horacio González  
2026-01-16



@Lost In Brittany



# Who are we?

Introducing myself and  
introducing Clever Cloud

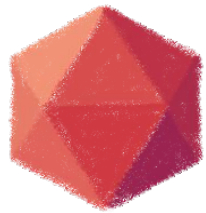


# Horacio Gonzalez

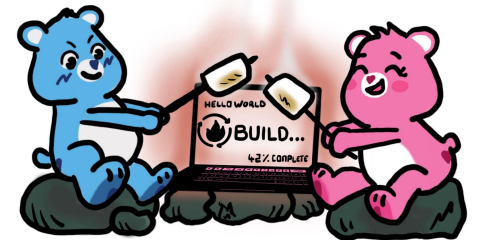
@LostInBrittany

Spaniard Lost in Brittany

Head of DevRel



clever cloud





# Clever Cloud

From Code to Product



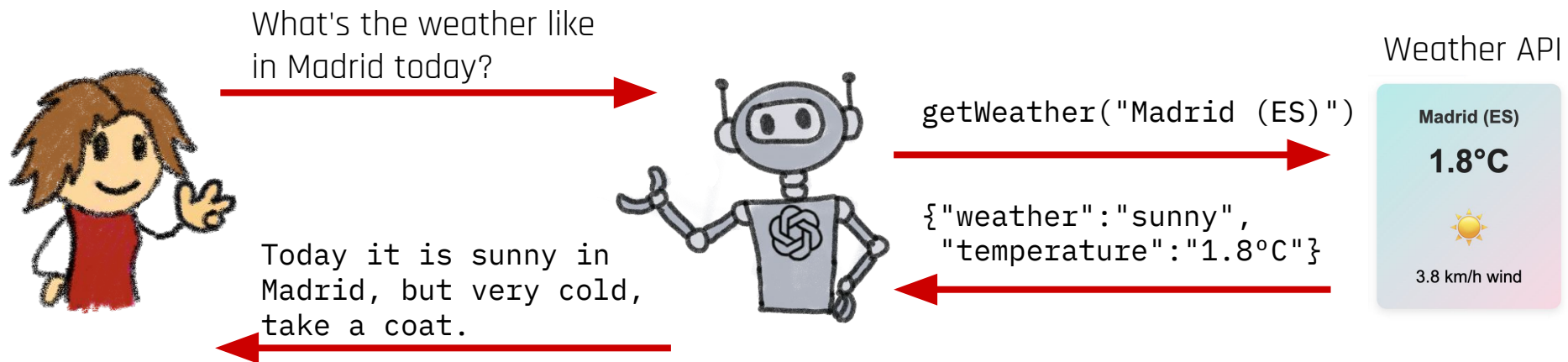
clever cloud





# LLM evolution

From simple chat to tool-enhanced agent!

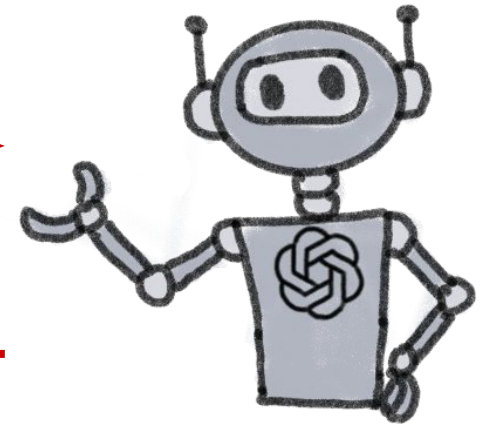


# LLM are only language models



What's the weather like in Madrid today?

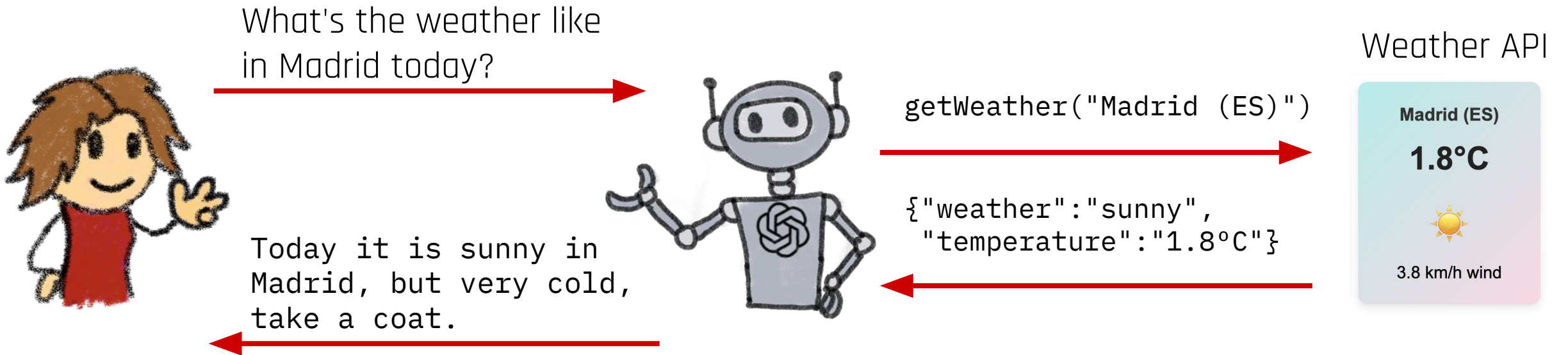
I'm unable to provide real-time  
information or current weather updates.



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



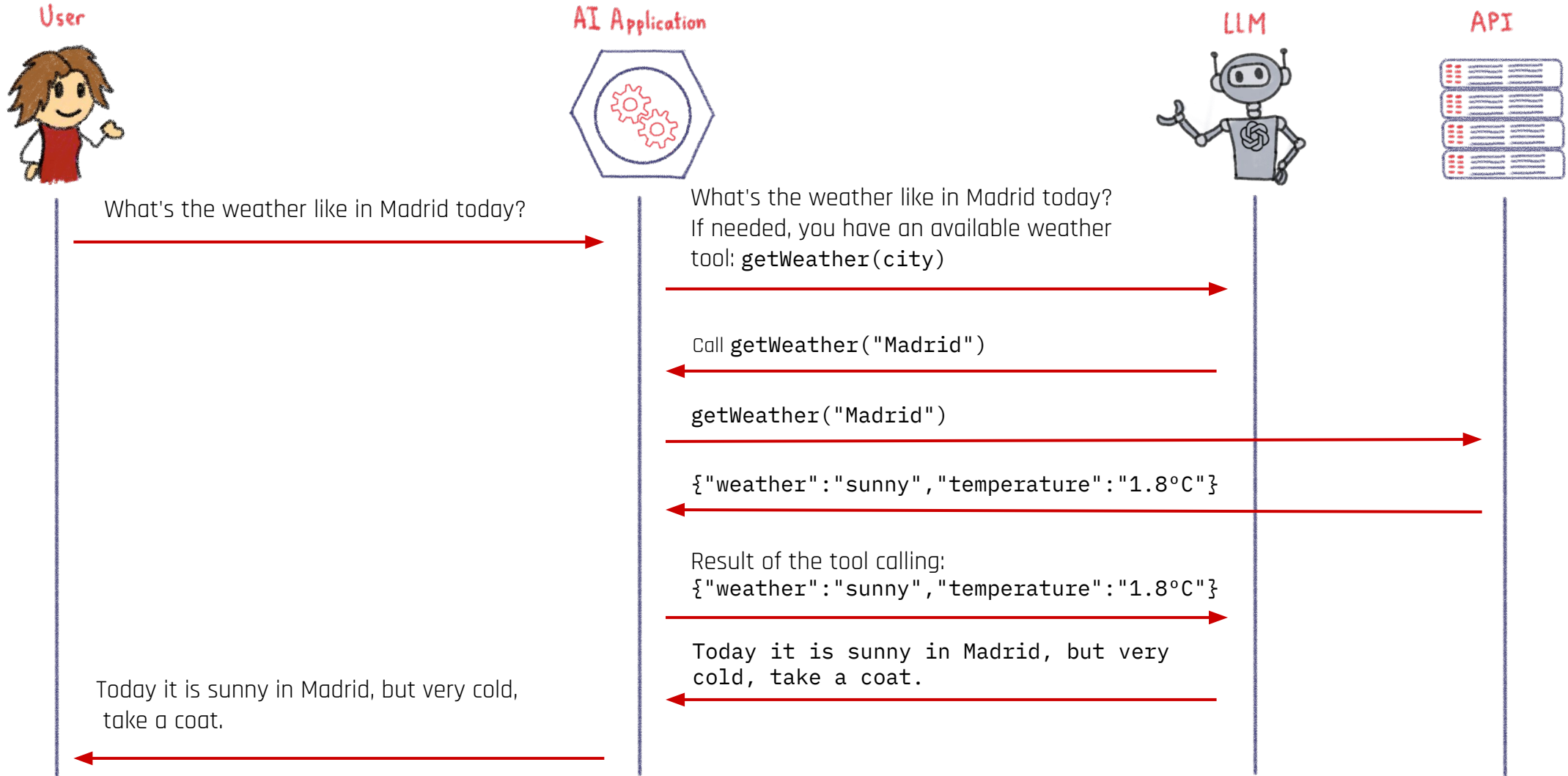
# Tools and plugins were added



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

//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?

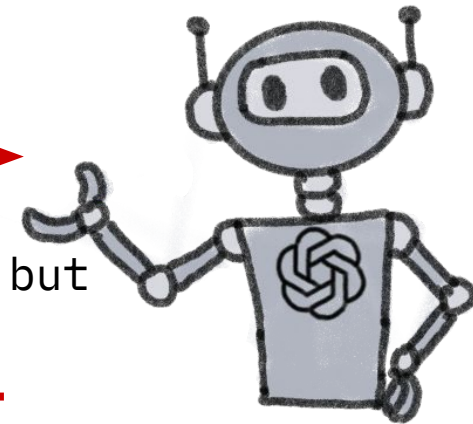


- Moves LLMs from static text generation
  - dynamic system components
- Increases accuracy & real-world usability
- Allows developers to control what the LLM can access

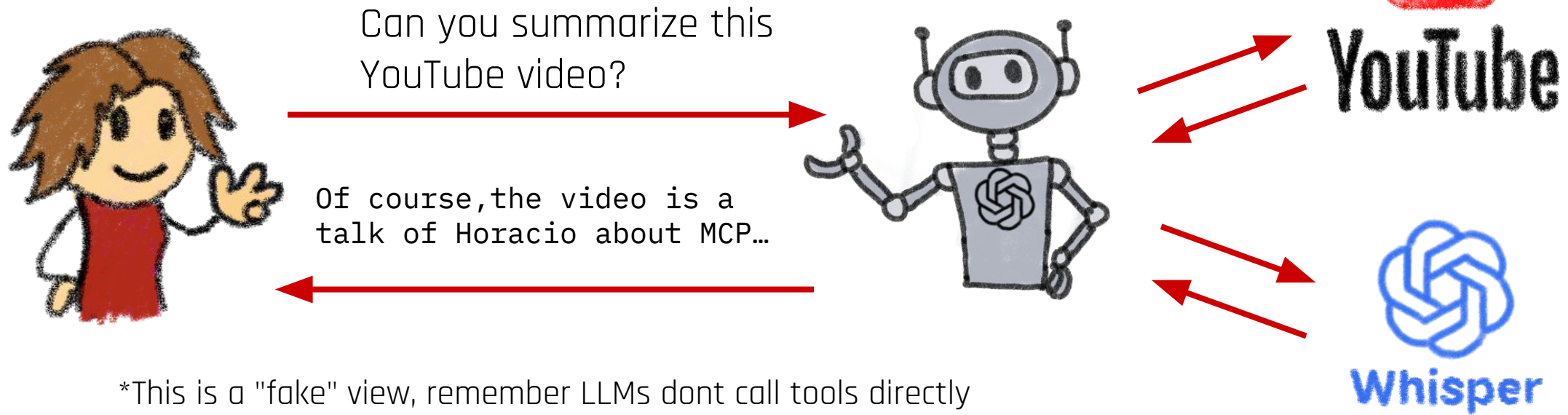


What's the weather like  
in Madrid today?

Today it is sunny in Madrid, but  
very cold, take a coat.



# From LLM chats to LLM-powered agents

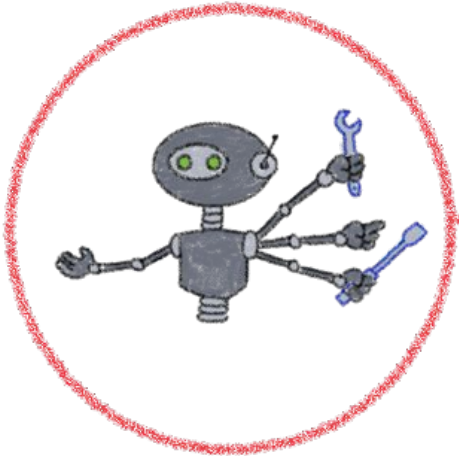


\*This is a "fake" view, remember LLMs don't call tools directly  
But it's the view from the Point of View of the user

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



# The rapid evolution of agents



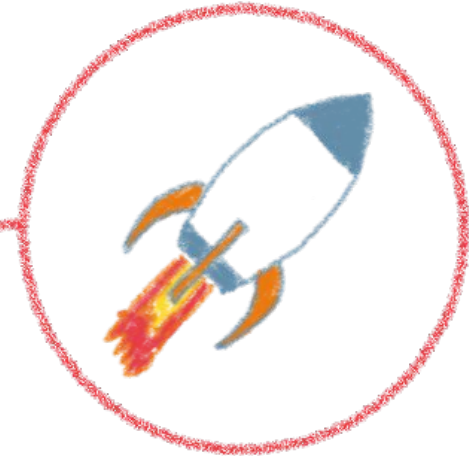
## Before MCP (2023–November 2024)

- 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.

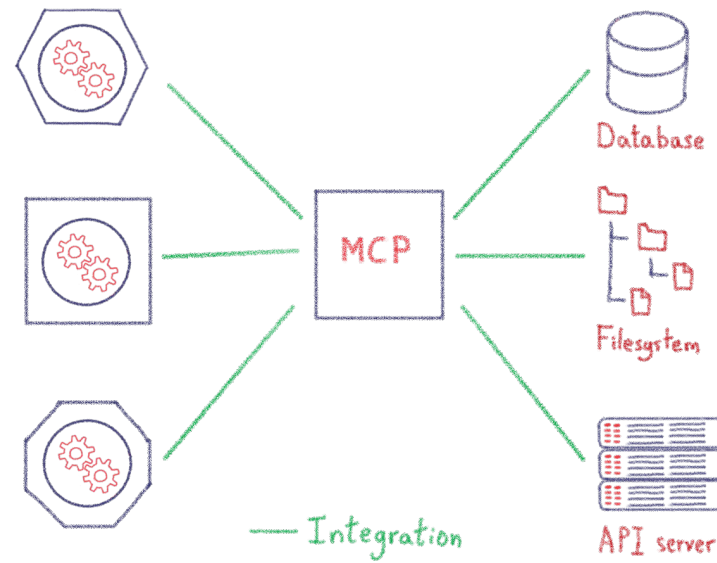




# Model Context Protocol (MCP): The missing link



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



# Why Do We Need MCP?



Function calling is powerful,  
why do I need another concept?



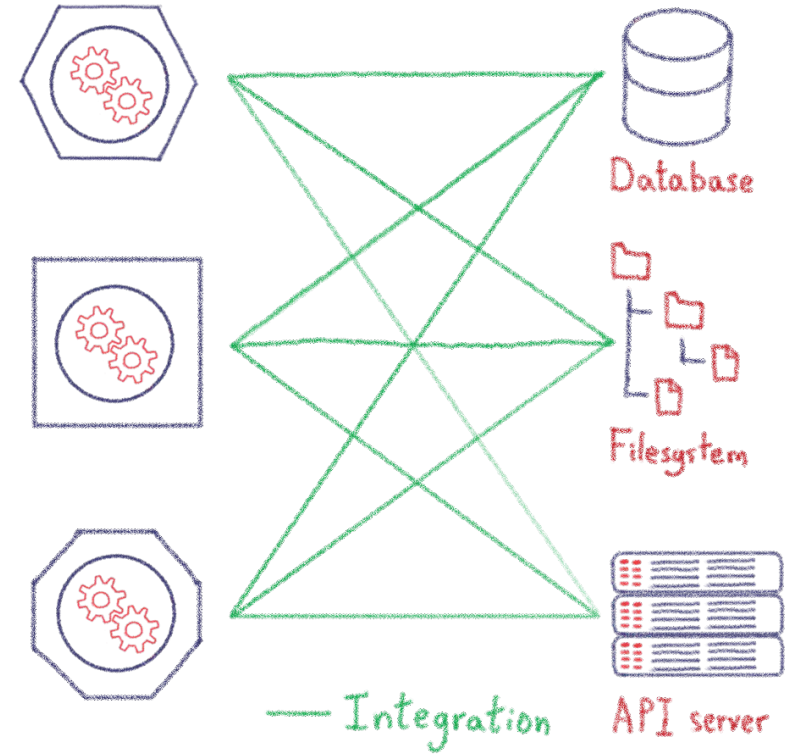
LLM function calling is useful,  
but it lacks structure

# 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

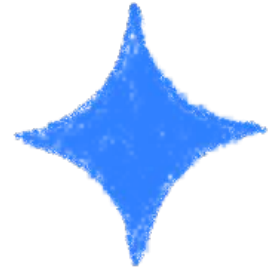
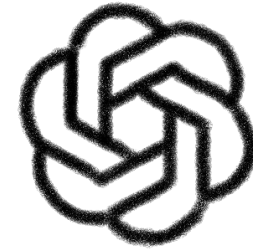
<https://modelcontextprotocol.io/>

# The MCP Landscape Today



Major players adopted MCP:

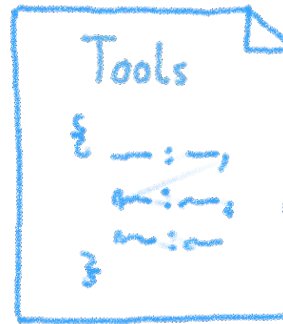
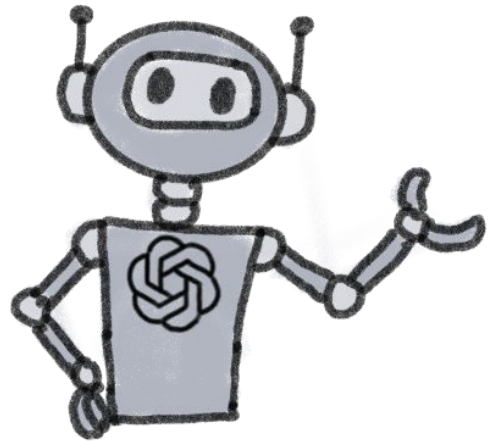
- **Anthropic** – Originator and tool provider (Claude Desktop, SDKs).
- **OpenAI** – Agent SDK, ChatGPT Desktop, Responses API.
- **Google DeepMind** – Gemini support and tooling.
- **Microsoft / GitHub** – Copilot Studio, Azure, Office integration, C# SDK.
- **Developer Platforms** – Replit, JetBrains, 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.



Weather  
MCP Server



# MCP is provider-agnostic



Works with any LLM provider

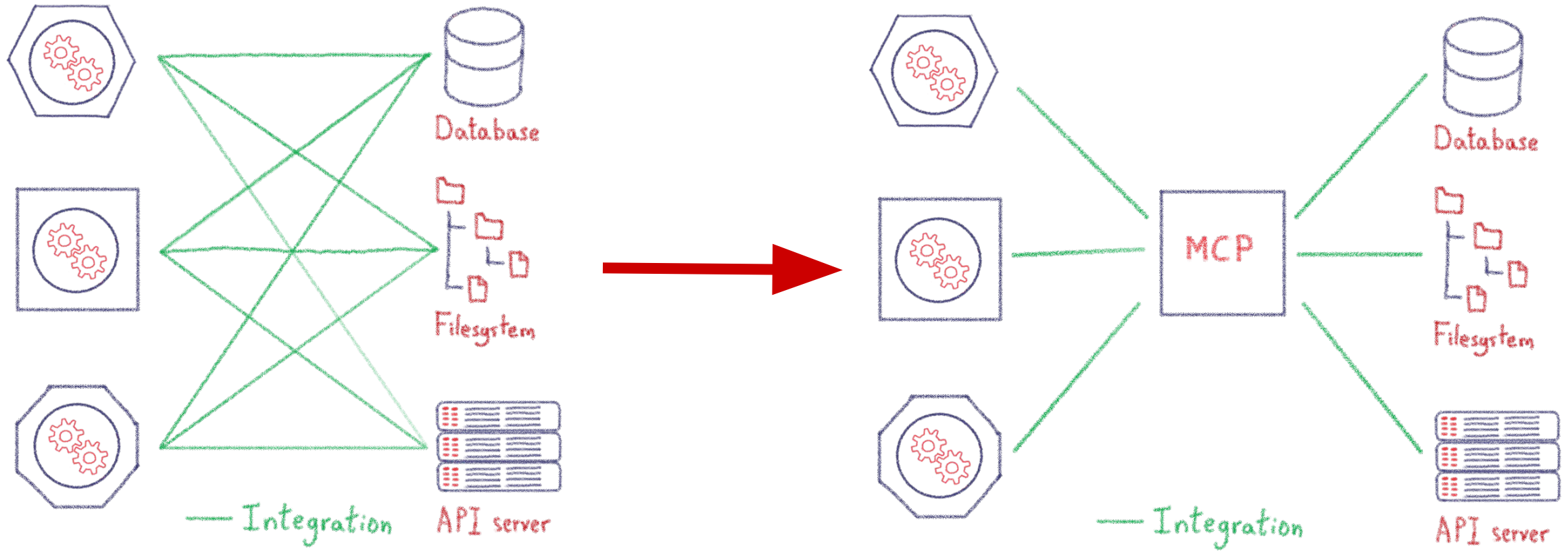


Ensures standardized function exposure  
across platforms





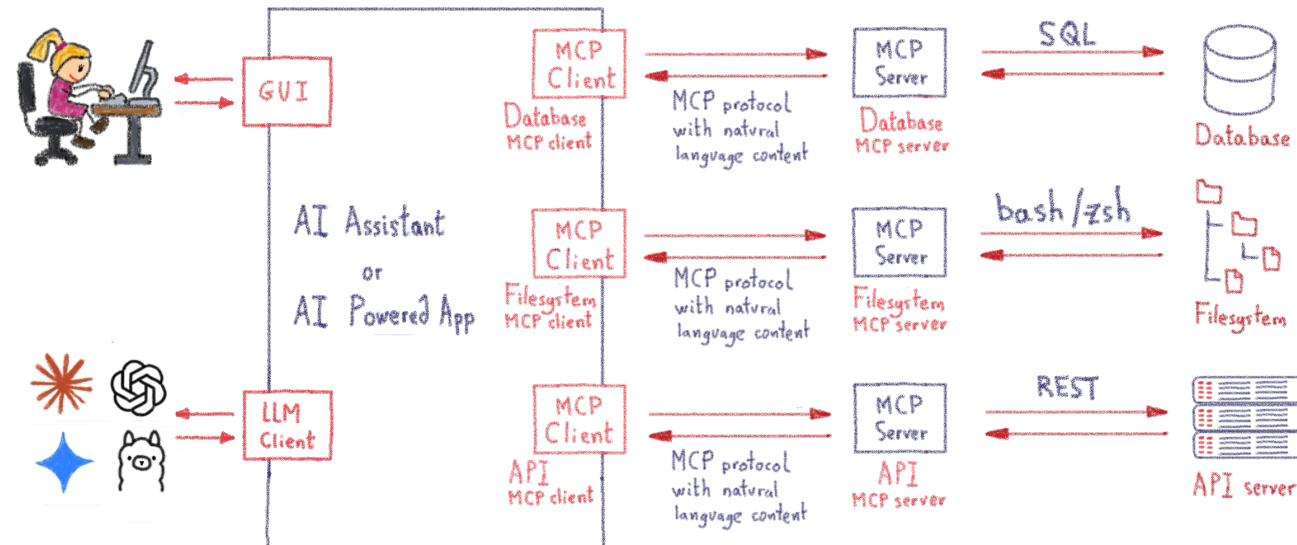
# MCP solves integration spaghetti



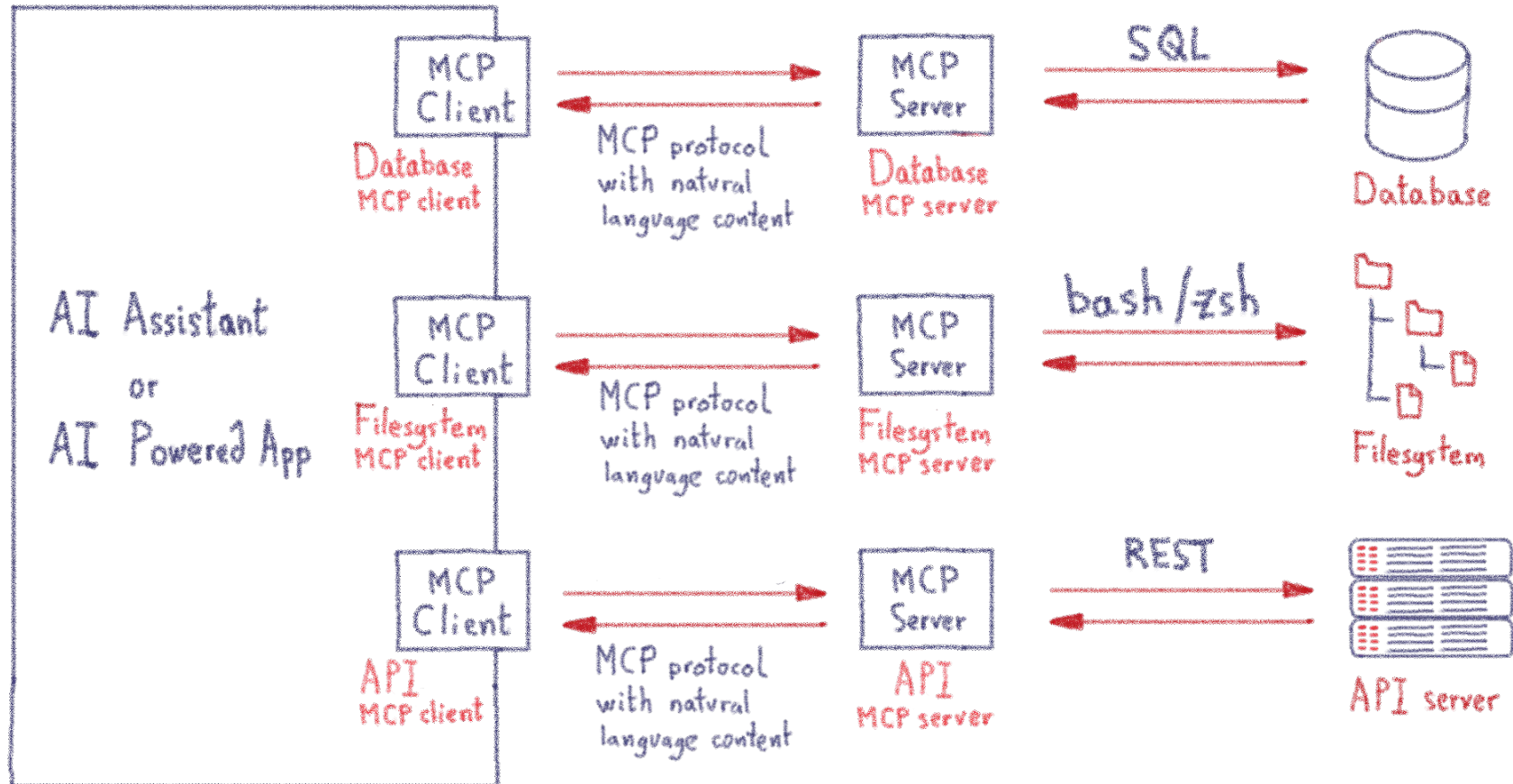


# The architecture of MCP

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

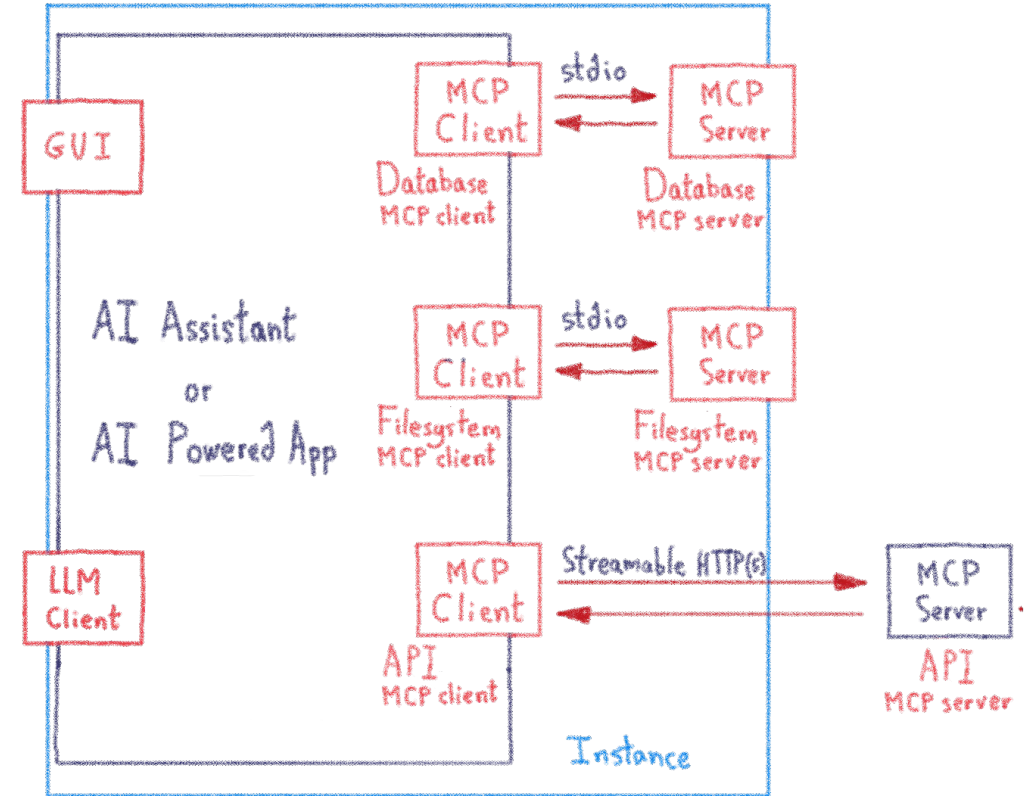


## MCP Protocol

Follow the JSON-RPC 2.0 specification

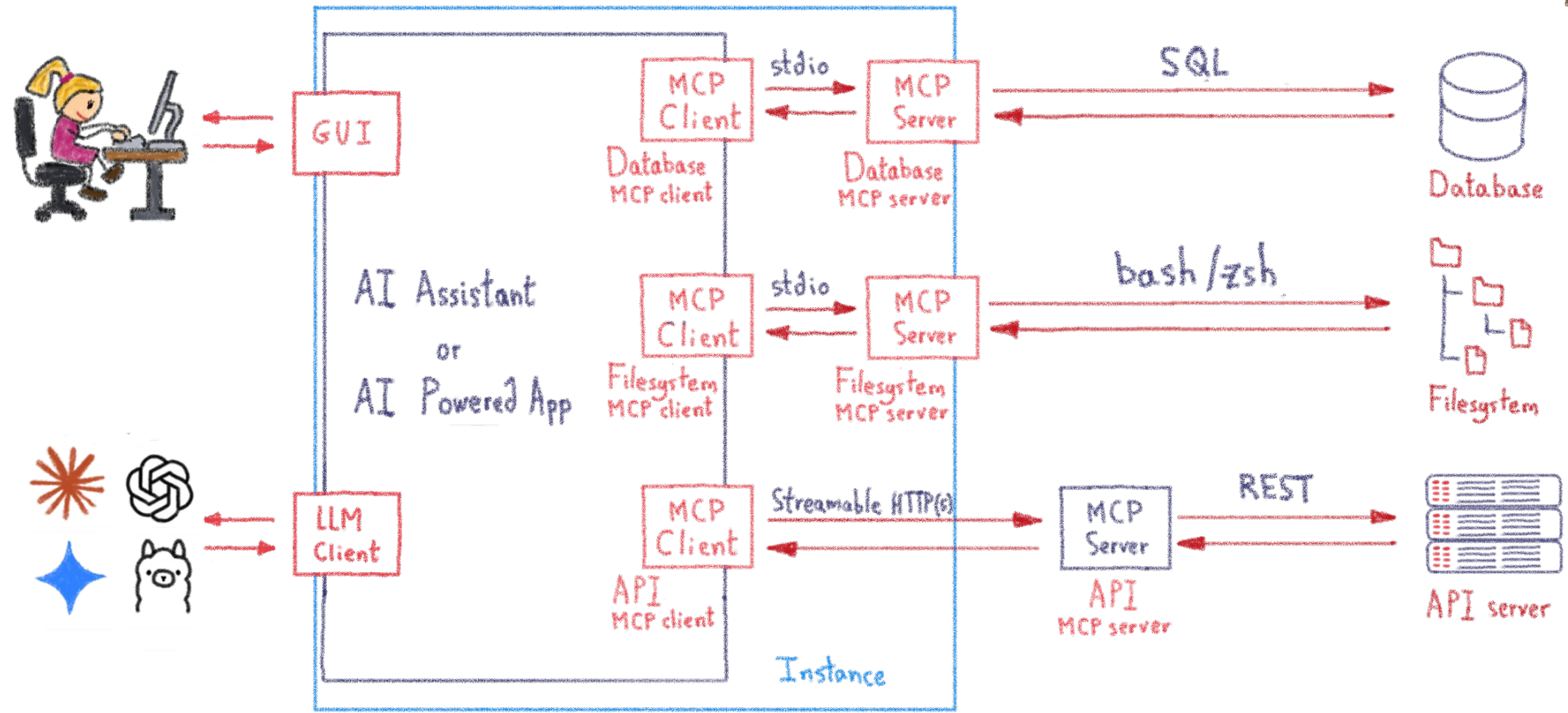
## MCP Transports

- 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





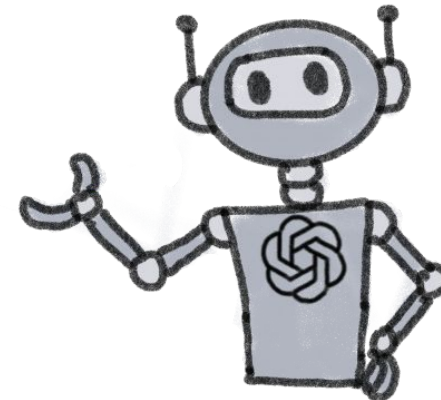
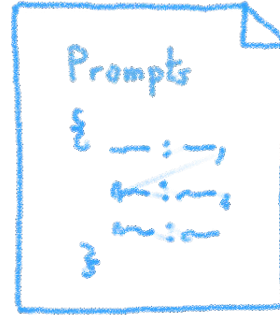
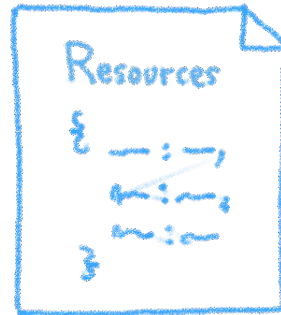
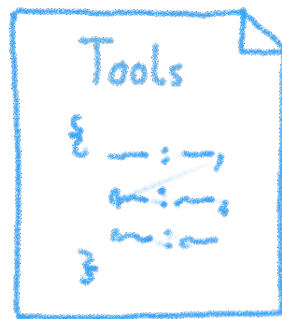
# Full MCP architecture



# Services: tools, resources & prompts



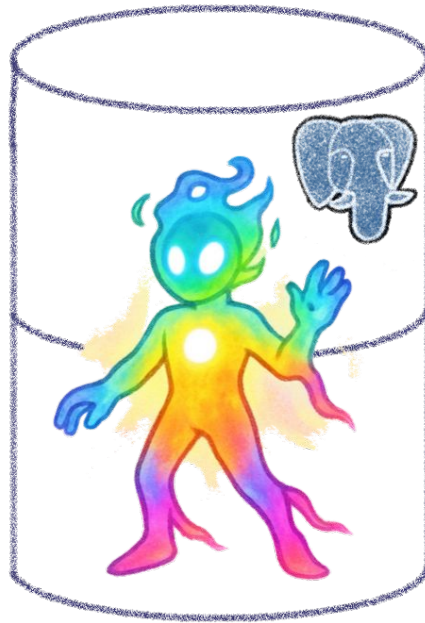
- Tools
  - Standardized way to expose functions that can be invoked by clients
- Resources
  - Standardized way to expose resources to clients
  - Each resource is uniquely identified by a URI
- Prompts
  - Standardized way to expose prompt templates to clients
  - Structured messages and instructions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





# MCPs are APIs

And they should be architected in a similar way



# Developer Expectations Have Shifted



## Winter 2024–2025

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



## Summer 2025

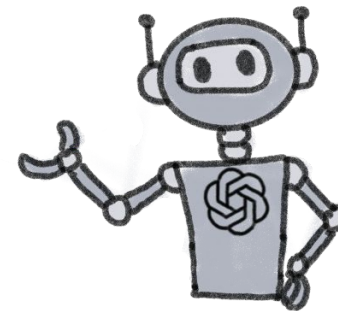
- “How do I build smarter MCP servers?”
- “How do I secure them?”
- “How do they fit into agent workflows?”



# MCP Servers: APIs for LLMs



CORBA  
MCP  
gRPC  
SOAP  
Thrift  
REST  
Protobuf



`getWeather("Madrid (ES)")`

`{"weather": "sunny",  
"temperature": "1.8°C"}`

Weather API





All those API technologies define protocols for communication between systems





# Let's use an example: RAGmonsters



 README  License

## 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:

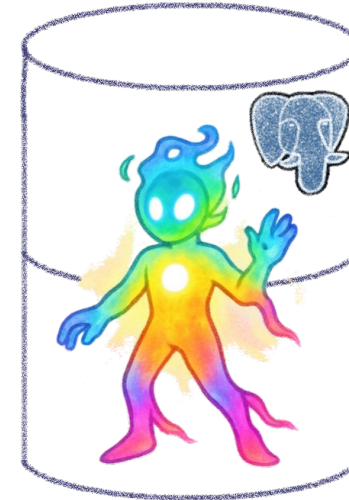
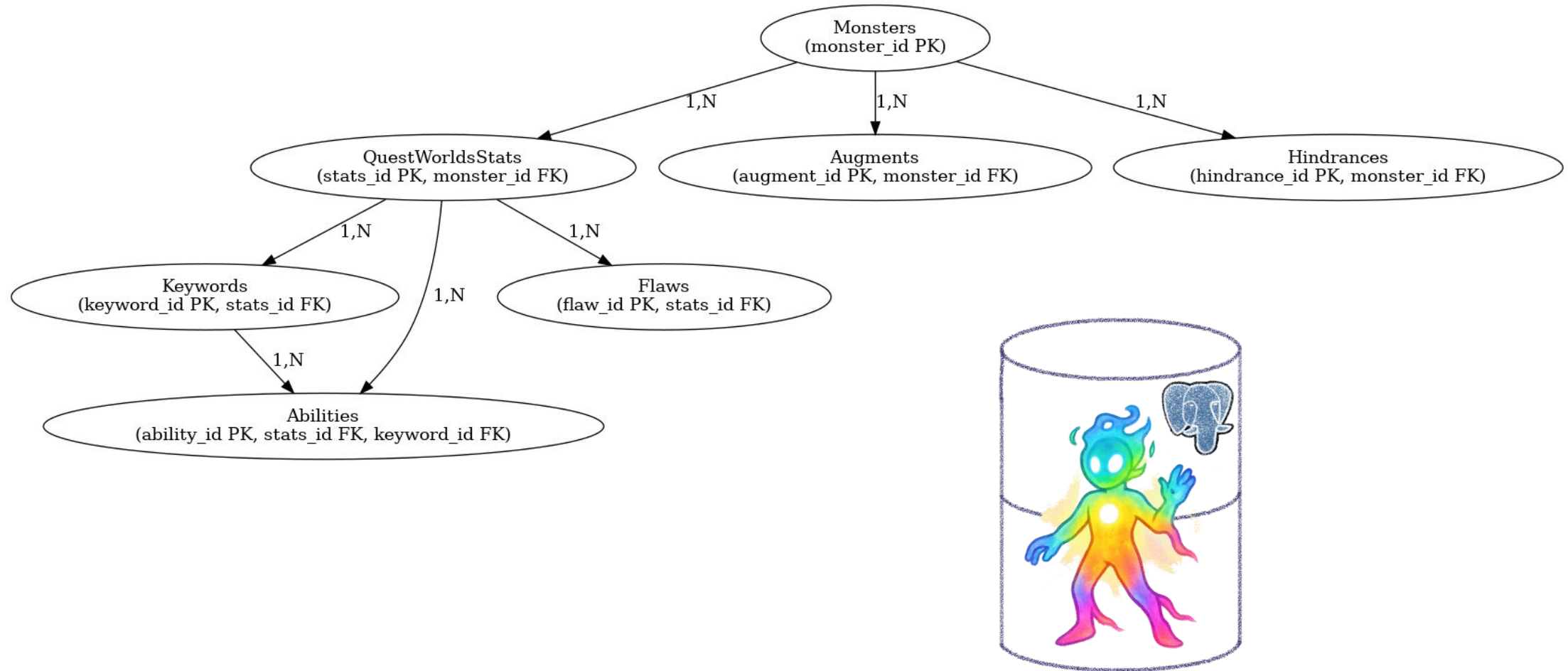
1. **Demonstrates RAG Value:** Shows how RAG can provide accurate answers about topics not in the LLM's training data
2. **Tests Retrieval Quality:** The varied attributes and relationships allow testing of different retrieval methods
3. **Supports Advanced Features:** Perfect for demonstrating filtering, 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](#)

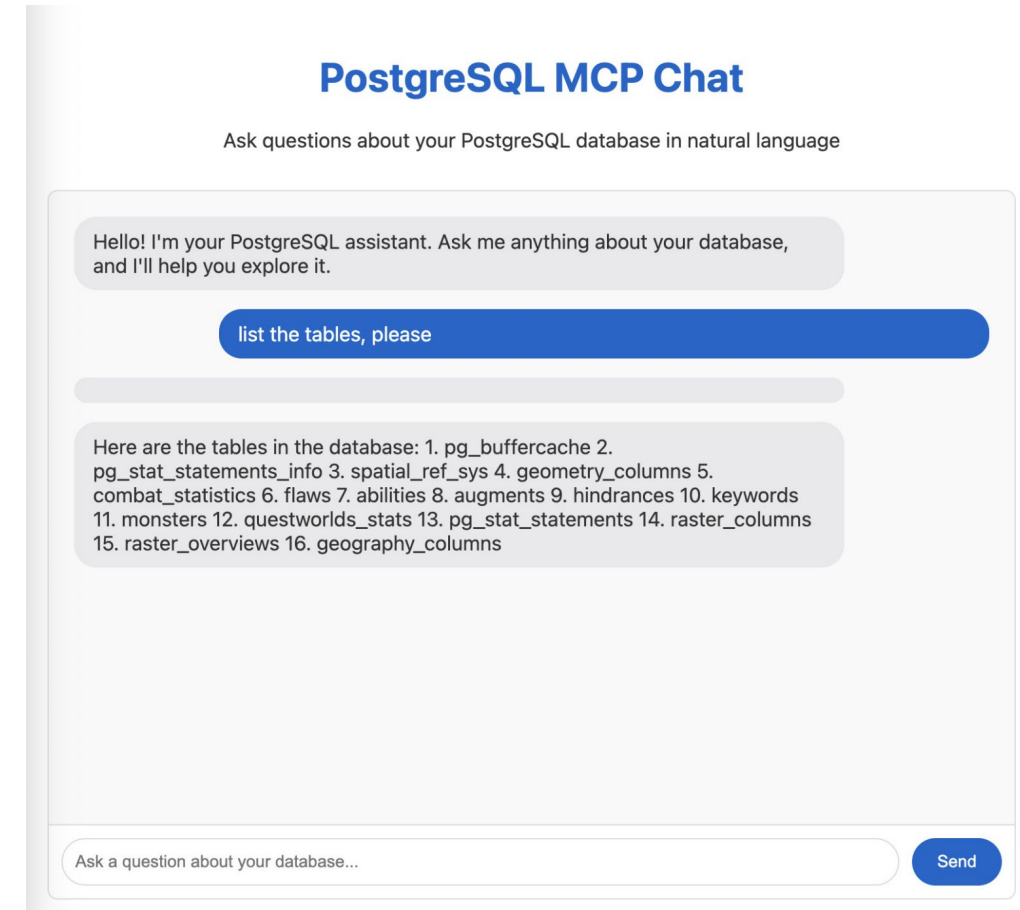
LLM can know what tables do we have and what is their structure, and it can request them

Implementation:

<https://github.com/CleverCloud/mcp-pg-example>

PostgreSQL MCP Server:

<https://github.com/modelcontextprotocol/servers/tree/main/src/postgres>



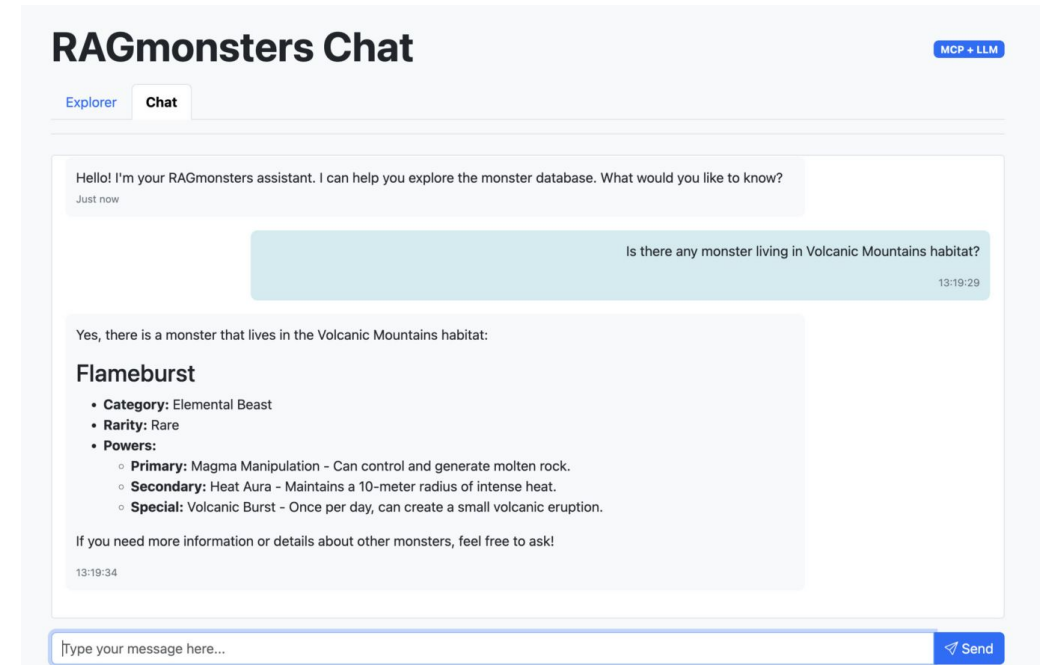


# Custom-made RAGmonsters MCP server



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/URLs 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!



*Please leave your  
feedback!*

