Why MCP?

AI development is often fragmented and brittle. MCP solves this by:

- Reducing fragmentation between models, tools, and data sources.
- Recognizing that LLMs are only as effective as the context they are given.
- Enabling smoother, consistent integration of tools, APIs, and resources into LLM workflows.

What is MCP?

MCP (Model Context Protocol) is a standard interface between language models and external context providers, such as:

- Tools (functions)
- Data sources (resources)
- Prompt templates

Just like:

- REST standardizes application communication
- LSP standardizes IDE-language tooling
- MCP standardizes LLM-context communication

Core Ideas:

- MCP is a standardization layer, not a reinvention.
- Everything possible with MCP can be done without it—just in a fragmented, errorprone way.
- Anyone can author MCP servers. Many reusable open-source servers are already available.

Isn't MCP just an API wrapper?

In some ways, yes—but it's more structured and LLM-native.

Key Benefits:

- Unified interface for tools, resources, and templates
- Strong typing, better error handling
- Easy integration with tools like LangChain or OpenDevin

MCP Architecture

MCP uses a Client-Server model built around three primitives:

1. Tools

Functions the model can invoke with parameters.

Example:

```
@mcp.tool()
def add(a: int, b: int) -> int:
    "Add two numbers"
    return a + b
```

Use Cases: Database updates, API calls, internal logic

2. Resources

Read-only content accessible by the model.

Static Resource Example:

```
@mcp.resource("config://version")
def static_greeting():
    return "Hello"
```

Dynamic Resource Example:

```
@mcp.resource("users://{name}")
def dynamic_greeting(name: str):
    return f"Hello {name}"
```

Use Cases: Config files, user profiles, DB records

3. Prompt Templates

Reusable templates for common LLM prompts.

Example:

```
@mcp.prompt()
def custom_prompt(name: str) -> str:
    "Generate a prompt asking for a summary."
    return f"Hello from prompt, {name}"
```

Use Cases: Q&A, summaries, structured JSON responses

Client-Server Initialization

To connect a client with an MCP server:

- 1. Client sends an initialization request
- 2. Server replies with an initialization response
- 3. Client sends an initialization notification

This ensures that the client understands what tools/resources are available.

MCP Transports

Depending on the setup, different transports are used:

| Mode | Protocol | Use Case |
|--------|-----------------|------------------------------------|
| Local | Stdio | CLI or test setups |
| Remote | http + sse | Streaming responses |
| Remote | streamable http | Stateless or stateful HTTP servers |

Creating an MCP Server

Minimal server using fastmcp:

```
from fastmcp import FastMCP
mcp = FastMCP("Demo")
if __name__ == "__main__":
    mcp.run()
```

After running, tools, resources, and prompts can be accessed by any compatible MCP client.

Complete Server Implementation

```
# server.py
from fastmcp import FastMCP
mcp = FastMCP("Demo")
@mcp.tool()
def add(a: int, b: int) -> int:
    """Add two numbers"""
    return a + b
@mcp.resource("config://greeting")
def static_greeting():
    return "Hello"
@mcp.resource("users://{name}")
def dynamic_greeting(name: str):
    return f"Hello {name}"
@mcp.prompt()
def custom_prompt(name: str) -> str:
    """Generate a prompt asking for a summary."""
    return f"Hello from prompt, {name}"
if __name__ == "__main__":
    mcp.run()
```

Complete Client Implementation

```
import asyncio
from fastmcp import Client
async def main():
    async with Client("server.py") as client:
        tools = await client.list tools()
        print(f"Available tools: {tools}\n")
        result = await client.call_tool("add", {"a": 5, "b": 3})
        print(f"Result: {result[0].text}\n")
        static_resources = await client.list_resources()
        print(f"Available static resources: {static_resources}\n")
        static_resource_response = await
client.read_resource("config://greeting")
        print(f"Result: {static_resource_response[0].text}\n")
        dynamic resources = await client.list resource templates()
        print(f"Available dynamic resources: {dynamic_resources}\n")
        dynamic_resource_response = await
client.read_resource("users://jainil")
        print(f"Result: {dynamic resource response[0].text}\n")
        prompts = await client.list prompts()
        print(f"Available Prompts: {prompts}\n")
        prompt = await client.get_prompt("custom_prompt",
                                        arguments={"name": "Jainil"})
        print(f"Result: {prompt}")
if __name__ == "__main__":
    asyncio.run(main())
```

Example Use Case: Al Support Agent

Using MCP to build a support chatbot:

- Tools: fetch_ticket, escalate_issue
- Resources: faqs://product-x, files://docs/manual.pdf
- Prompts: summarize_transcript, reply_template

This architecture reduces prompt complexity and helps scale context-rich AI features with ease.

Summary

| MCP Components | Purpose |
|------------------------|--|
| Tool | MCP Server Primitive, Invokable function |
| Resource | MCP Server Primitive, Read-only data source |
| Prompt Template | MCP Server Primitive, Standardized prompt creation |
| Transport | Communication method |
| Server | Hosts all MCP primitives |
| Client | Connects and consumes server features |