

BOOK-M@TE

Building an AI-assistant

ABOUT ME

nupsea.github.io



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WHY?

- Hard to remember
 - Summarization & Insights
 - Multi book Comparison
 - Self - hosted
-

AGENDA

Chapter 1 – Foundations

- LLMs
- RAG & Search

Chapter 2 – AI Assistant

- Agents & MCP
- Eval & Monitoring

Chapter 3 – Book-mate App

- Architecture & Components
- Demo

Chapter 4 – Future

- Learnings, Next Steps
- Q n A

CHAPTER 1

FOUNDATIONS

LLMs

Strengths

- Natural language generation
- Reasoning and summarization
- Fast inference with modern models

Limitations

- Missing Context
- Outdated knowledge
- Hallucinations
- Token window limits

LLMs

Foundational Models (*Innovation*)

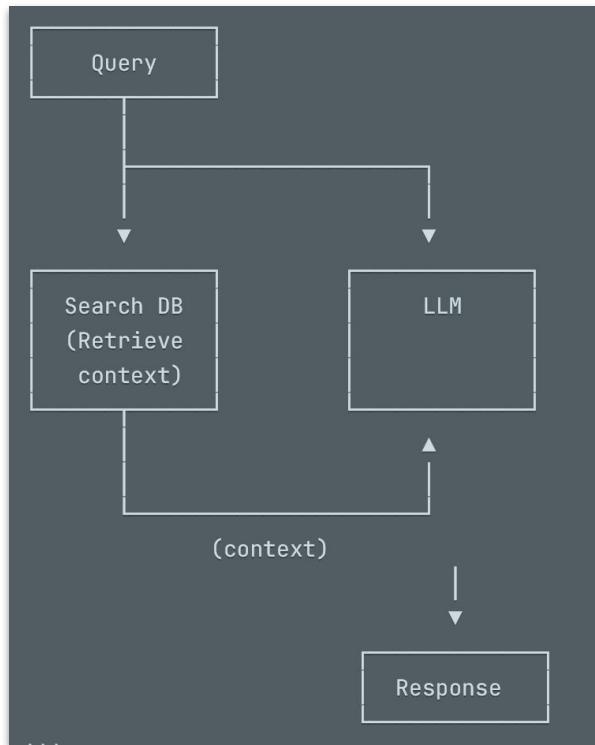
Pretraining

Adoption Approaches (*Engineering*)

- Fine-tuning: Adapt to your domain
- RAGs + Prompt Engineering + Agentic : Widely used

RAG

Retrieval Augmented Generation



Advantages:

- Simple architecture and safe
- No retraining required
- Adds latest knowledge

Challenges:

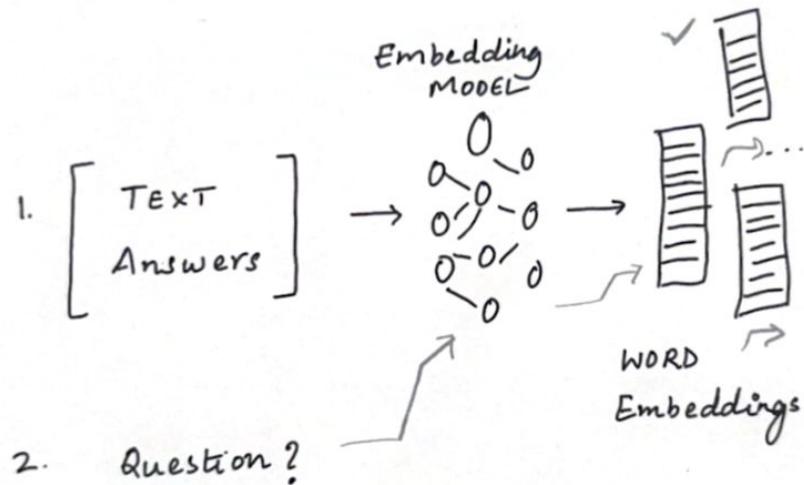
- Limited to the model's skills.
- Heavy reliance on search.

SEARCH

Query: "Who is Cheshire?"

- **✗ Keyword search:** Fails if user says "grinning cat"
- **✓ Vector search:** Understands semantic meaning

Vectorization.



SEARCH

-   **Hybrid: Best of both**

```
def hybrid_search(query):
    bm25_hits = keyword_index.search(query)
    vec_hits = vector_index.search(encode(query))

    return rrf_merge(bm25_hits, vec_hits)
```

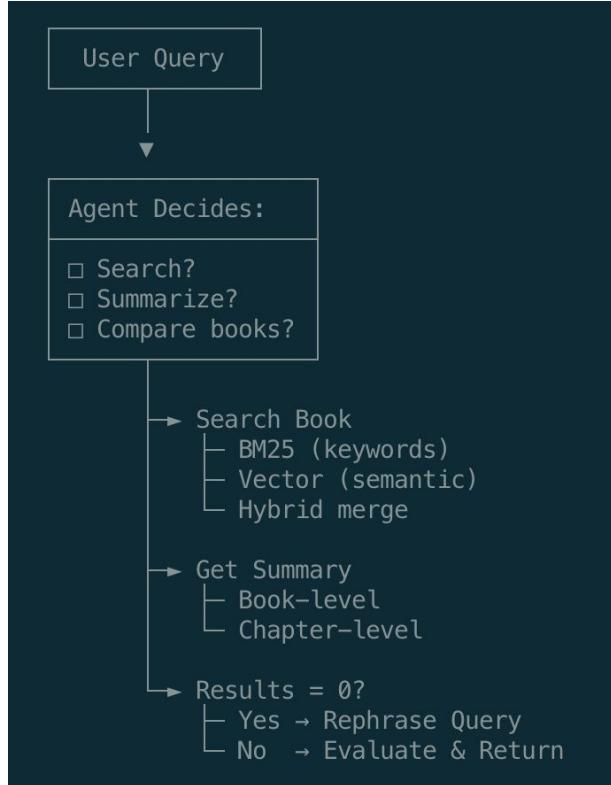
CHAPTER 2

ASSISTANT

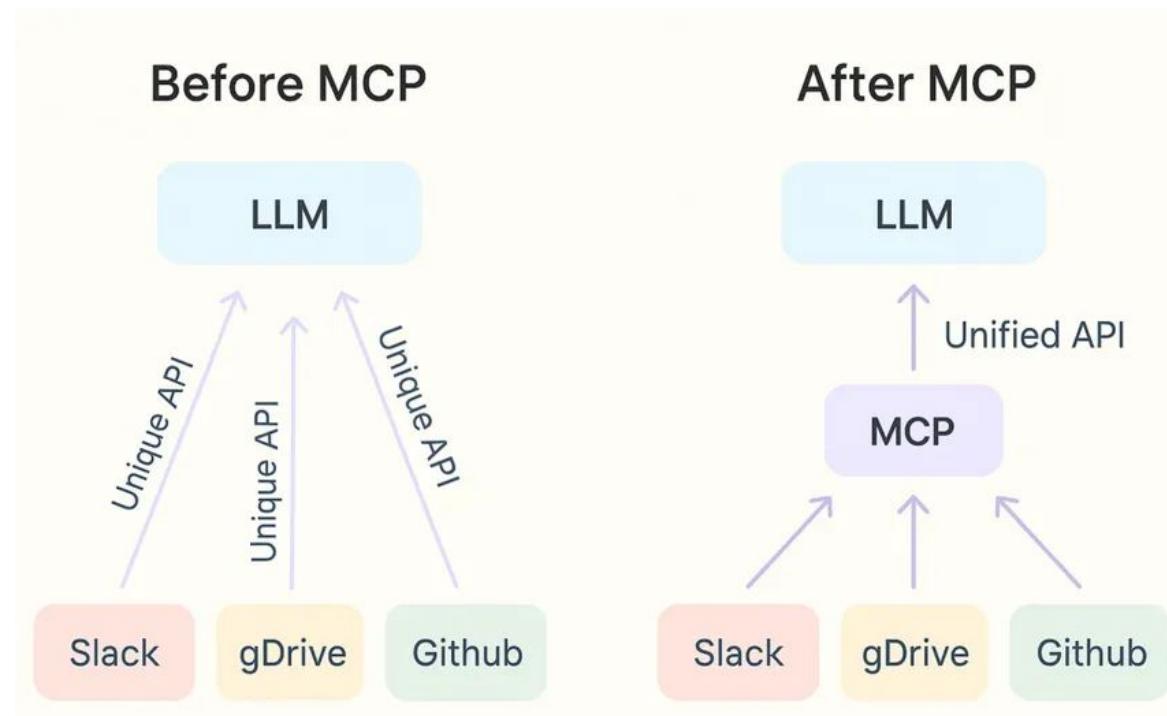


AGENTS

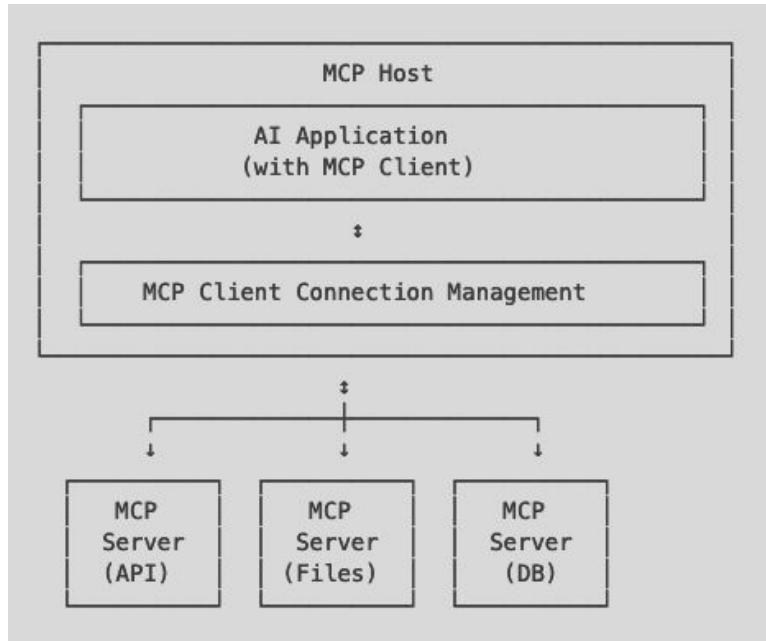
- Complex, multi steps
- External integration
- Dynamic Decision Making
- Accuracy & Reliability
- Stateful Operations
- Task Automation



MODEL CONTEXT PROTOCOL



MCP ARCHITECTURE



- **MCP Server:** Exposes Tools [search_book, get_summary..]
 - Resources
 - Prompts
- **MCP Client:** Agent (makes decisions)
- **MCP Host:** Manages connections

MCP TOOLS

```
Tool(  
    name="search_book",  
    description="Search for content within a book using hybrid (BM25 + Vector) retrieval. Returns relevant text chunks.",  
    inputSchema={  
        "type": "object",  
        "properties": {  
            "query": {"type": "string", "description": "The search query"},  
            "book_identifier": {  
                "type": "string",  
                "description": "The book SLUG from the available books list (e.g., 'abc', 'xyz'). "  
                "MUST use the slug shown in [square brackets], NOT the full title.",  
            },  
            "limit": {  
                "type": "integer",  
                "description": "Number of results to return",  
                "default": 5,  
            },  
        },  
        "required": ["query", "book_identifier"],  
    },  
)
```

EVAL & MONITORING

Search Eval: Golden data + HR & MRR

Book	Hit Rate@5	MRR@5
Hegel	70.3%	0.58
Marcus	68.1%	0.55

LLM Response Eval: LLM as a judge.

```
prompt = f"""
Rate the relevance of the answer to the question (0-5):
Q: {question}
A: {answer}
"""

score = llm(prompt)
```

Observability:

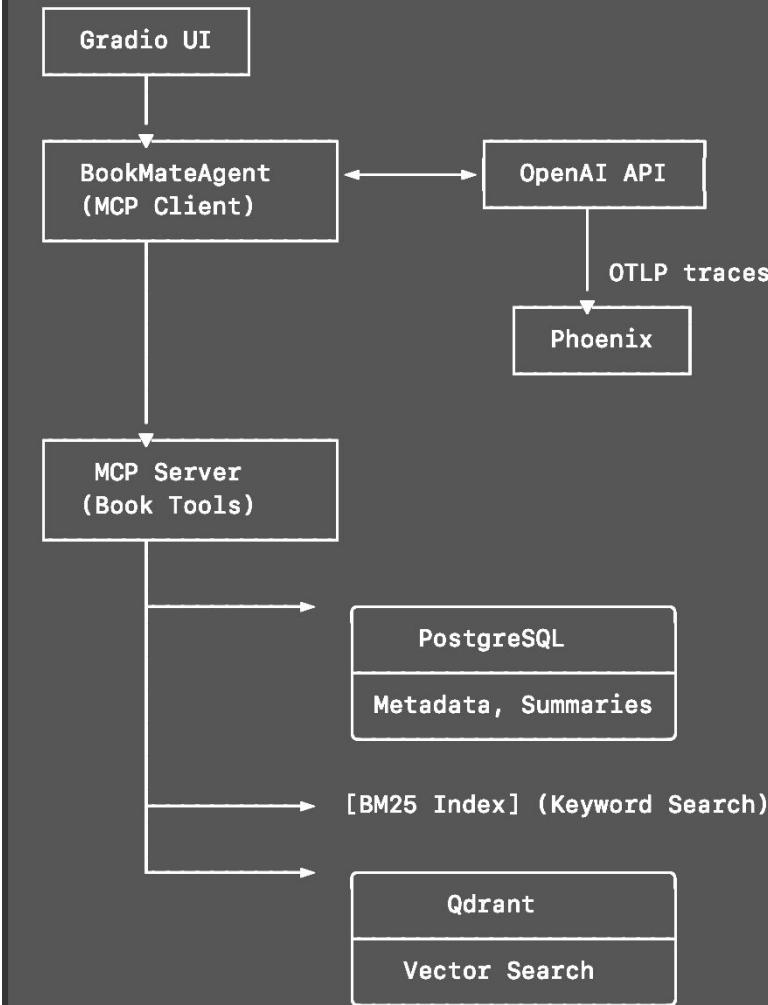
- **Metrics** with User Ratings
- Distributed **Tracing**
- Automated Tests

CHAPTER 3

BOOK-MATE APP

ARCHITECTURE

BOOK - MATE



DEMO

CHAPTER 4

FUTURE

LEARNINGS & CHALLENGES

- Search optimizations to improve eval metrics
- Non deterministic workflows
- Real world: different book sources.

Next Steps:

- Supporting more formats and expanding ecosystem.
- Security and Essential Guard Rails with Alerting
- Comprehend images and extend to technical-docs
- Data collection and analytics.

Q&A

Suggestions:

- Try it yourself.
github.com/nupsea/book-mate
- ⭐ the repo if you found it useful
- Highly appreciate contributions

References:

- LLM Zoomcamp - Alexey
datatalks.club/courses/llm-zoomcamp
- Books - Gutenberg Project
<https://www.gutenberg.org/>



Thank you!



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