

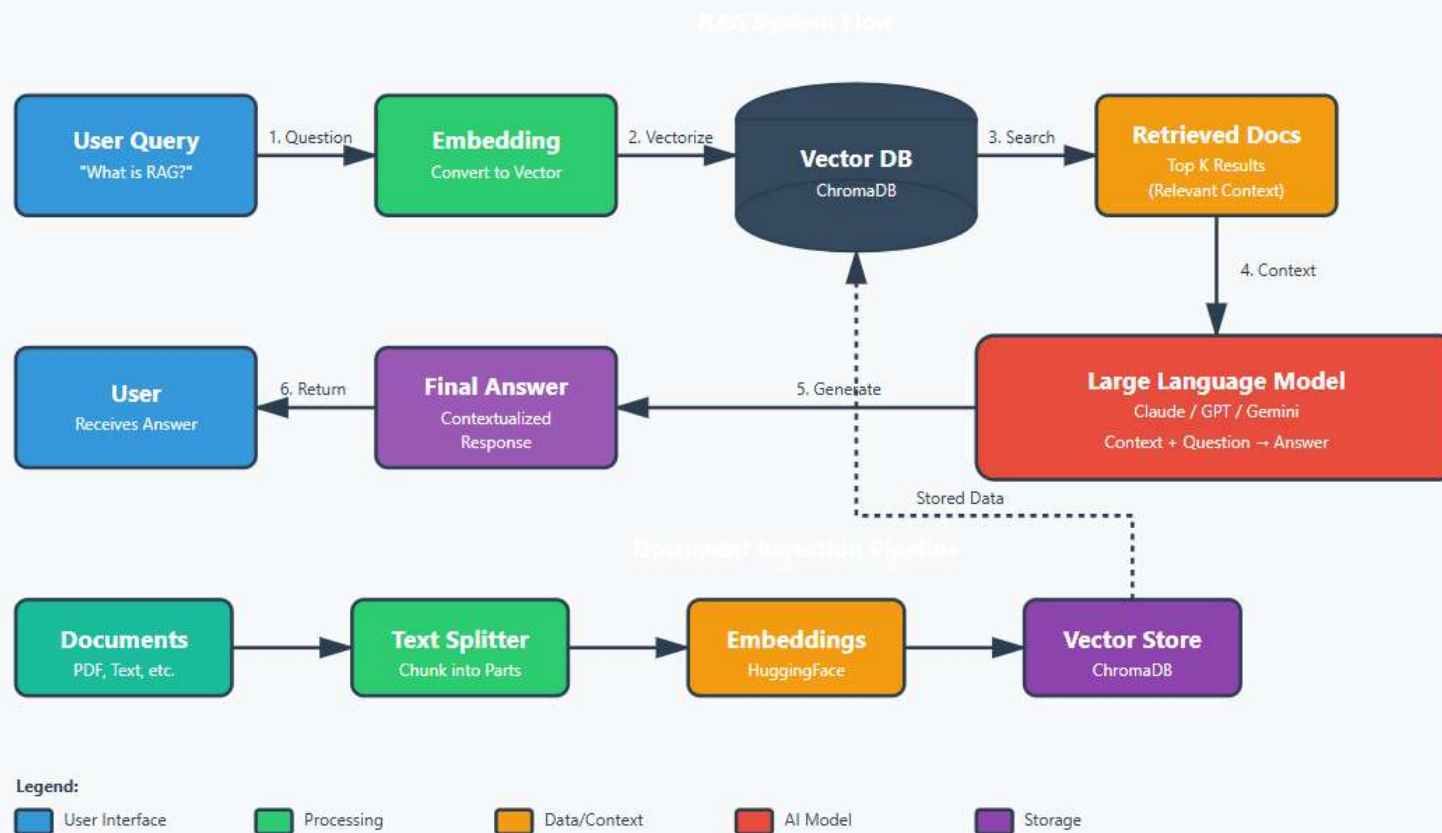
RAG MCP 서버

With ChromaDB, LangChain, Claude LLC

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🔍 RAG (Retrieval-Augmented Generation) Architecture





RAG Process Flow

1

Document Ingestion (add_documents)

Receive source documents → Split into chunks using RecursiveCharacterTextSplitter → Convert to vectors using HuggingFace embeddings → Store in ChromaDB with metadata

2

Vector Search (search_documents)

Convert user query to embedding → Search ChromaDB using cosine similarity → Return top k documents (default: 4)

3

Context Construction

Merge retrieved documents into unified context → Maintain document order and source information → Generate prompt template

4

AI Answer Generation (rag_query)

Send context + question to Claude API → Claude Sonnet 4 model generates answer → Return result with source citations

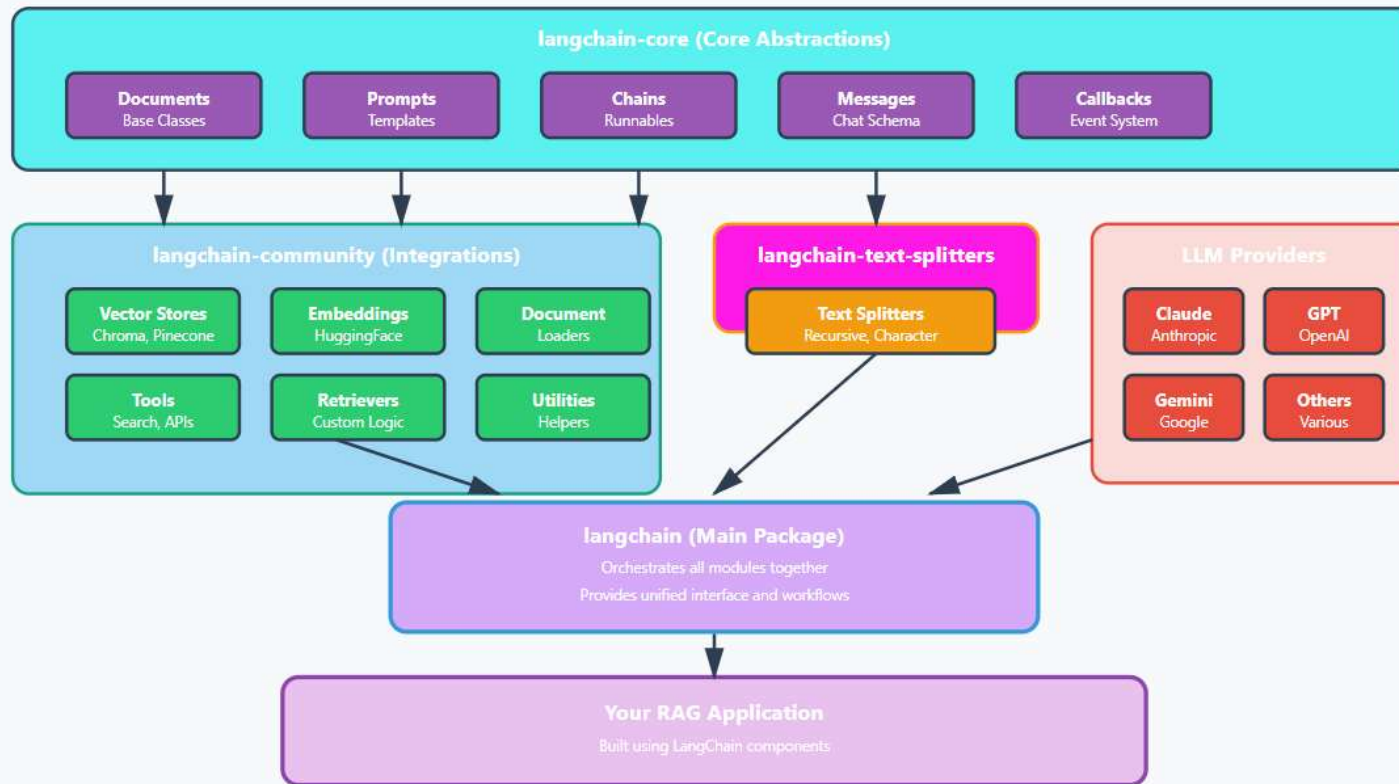
5

Result Formatting & Return

Format answer text + reference document info → Include metadata (source, chunk position, etc.) → Deliver final result to user

LangChain 1.0 Architecture

LangChain Modules Architecture



Module Dependencies: Depends on ■ Main Package ■ Community Integrations ■ LLM Providers ■ Core Abstractions

RAG + LangChain Integration

How LangChain Powers RAG

RAG Components

1. Document Loading

Load PDF, Text, HTML

2. Text Splitting

Chunk documents

3. Embedding

Convert to vectors

4. Vector Storage

Store in database

5. Retrieval

Search similar docs

LangChain Implementation

Document Loaders

PyPDFLoader, TextLoader, UnstructuredLoader
from langchain_community.document_loaders

Text Splitters

RecursiveCharacterTextSplitter
from langchain_text_splitters

Embeddings

HuggingFaceEmbeddings, OpenAIEmbeddings
from langchain_community.embeddings

Vector Stores

Chroma, Pinecone, FAISS, Weaviate
from langchain_community.vectorstores

Retrievers

VectorStoreRetriever, similarity_search()
Built-in retrieval methods

uses

uses

uses

uses

uses

System Architecture

Client Layer

Claude Desktop App

MCP Client

User Interface

Server Layer

MCP Server (http)

Tool Handlers

LangChain Integration

Claude API Client

Storage Layer

ChromaDB

Vector Store

HuggingFace Embeddings

Persistent Storage

Download source

#Download source

\$> **git clone** <https://github.com/hjkim7796/RAG-CHROMA-MCP-SERVER.git>

\$> **dir RAG-CHROMA-MCP-SERVER**

Directory: C:\hjkim\PythonProject\WRAG-CHROMA-MCP-SERVER

Mode	LastWriteTime	Length	Name
d----	11/21/2025 12:06 PM		.env
d----	11/21/2025 11:58 AM		docs
-a----	11/21/2025 11:58 AM	68	.gitignore
-a----	11/21/2025 11:58 AM	227607	abc.pdf
-a----	11/21/2025 11:58 AM	16847	add_pdf_to_mcp.py
-a----	11/21/2025 11:58 AM	5657	mcp-http-proxy.py
-a----	11/21/2025 11:58 AM	383	query-chromaDB.py
-a----	11/21/2025 11:58 AM	20699	rag_mcp_http_server.py
-a----	11/21/2025 11:58 AM	7821	README_HTTP.md
-a----	11/21/2025 11:58 AM	347	requirements.txt
-a----	11/21/2025 11:58 AM	6936	requirements_freeze.txt
-a----	11/21/2025 11:58 AM	11420	test_server.py
-a----	11/21/2025 11:58 AM	2034	test_server_with_proxy.py

Run rag-mcp-server

#Go to the source folder

```
$> cd <user-project-folder>
```

#Config a virtual environment (venv) in VS Code for Python to ensure your project uses the correct isolated environment.

```
$> python -m venv .venv
```

```
$> .venv/Scripts/activate.ps1
```

#Install all python packages

```
(.venv)> pip install -r requirements.txt
```

#Run rag mcp server

```
(.venv)> python rag_mcp_http_server.py
```


Test rag-mcp-server

#Go to the source folder

```
(.venv)> cd <user-project-folder>
```

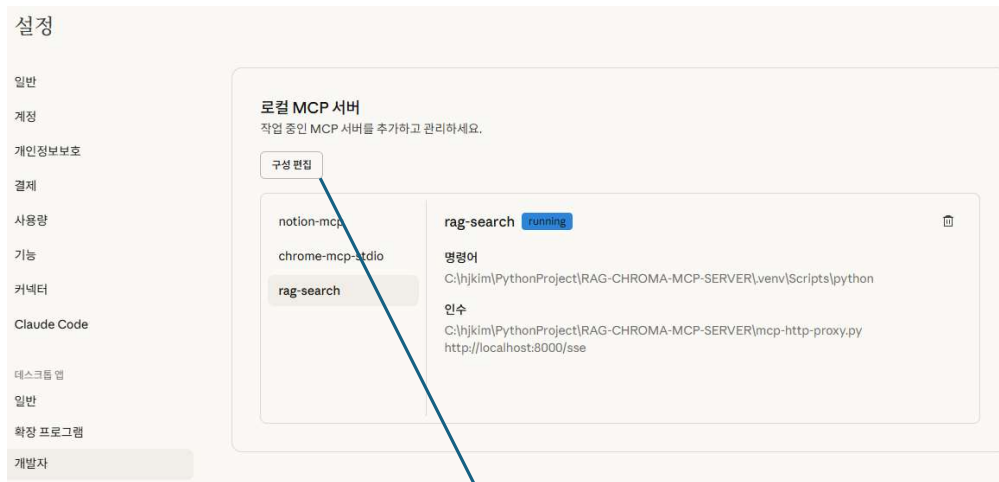
#Check server is working properly

```
(.venv)> python test_server.py
```

#Check proxy server is working properly

```
(.venv)> python test_server_with_proxy.py
```

Config for Claude Desktop integration



File location: %APPDATA%\Claude\claude_desktop_config.json

```
{
  "mcpServers": {
    "rag-search": {
      "command": "C:\\hj\\kim\\PythonProject\\RAG-CHROMA-MCP-SERVER\\venv\\Scripts\\python",
      "args": ["C:\\hj\\kim\\PythonProject\\RAG-CHROMA-MCP-SERVER\\mcp_http_server.py"]
    }
  }
}
```

Test rag-mcp-server

#Go to the source folder

```
(.venv)> cd <user-project-folder>
```

#Check server is working properly

```
(.venv)> python test_server.py
```

#Check proxy server is working properly

```
(.venv)> python test_server_with_proxy.py
```

Add a pdf document

#Go to the source folder

```
(.venv)> cd <user-project-folder>
```

#Add pdf document to RAG system

```
(.venv)> python add_pdf_to_mcp.py abc.pdf
```

...

```
=====
✅ PDF successfully added to RAG system!
=====
```

Test query on claude desktop

Query: rag-search MCP 서버에 접속해서 "EO Patch Wearable Insulin Pump" 에 대해서 요약해줘

L rag-search MCP 서버에 접속해서 "EO Patch Wearable Insulin Pump" 에 대해서 요약해줘

EO Patch에 대해 RAG 검색을 수행하겠습니다.

R Rag query

API 키 설정에 문제가 있는 것 같습니다. 대신 검색 기능을 사용해서 문서를 찾아볼게요.

R Search documents

RAG 검색 결과를 바탕으로 **EO Patch Wearable Insulin Pump**에 대해 요약해드리겠습니다:

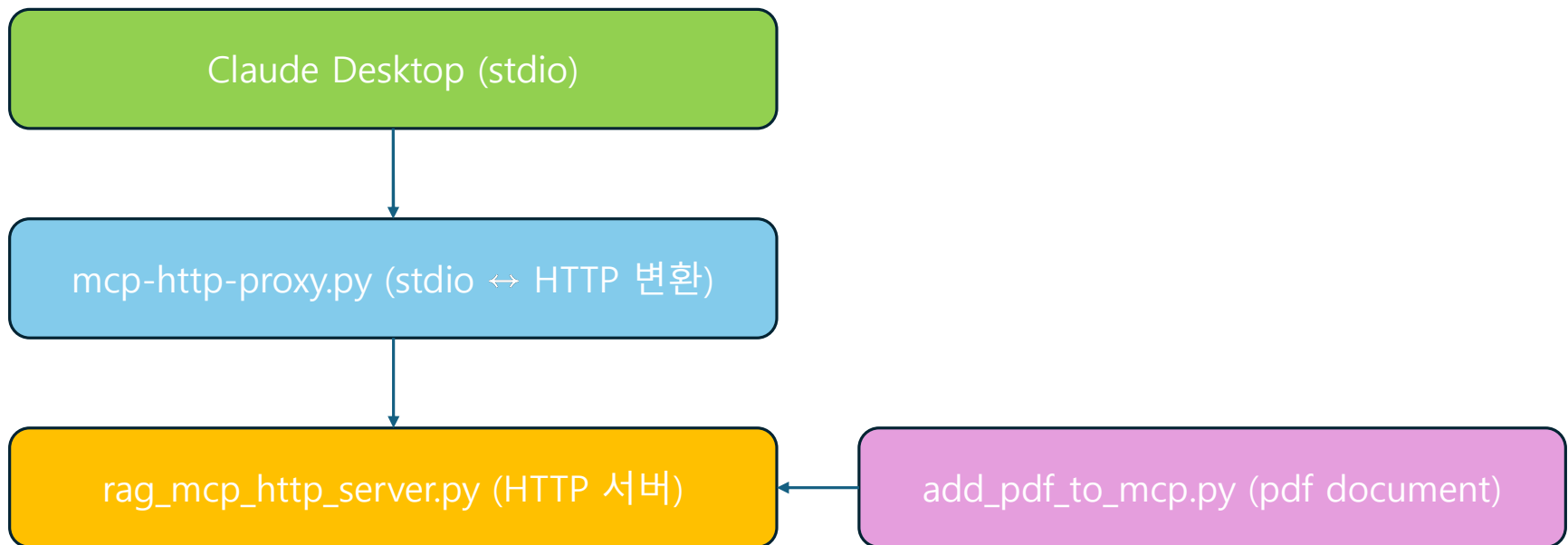
Rag query를 활성화 하려면 .env 파일을 생성한 후 아래와 같이 추가
ANTHROPIC_API_KEY=your-api-key-here



File Description

- [rag_mcp_server.py](#) - Main RAG MCP server
- [add_pdf_to_mcp.py](#) – Example for adding a PDF document to RAG system
- [query-chromaDB.py](#) – Example for querying to chromaDB
- [test_server.py](#) - Example for testing the RAG MCP server
- [test_server_with_proxy.py](#) - Example for testing the Proxy server
- [requirements.txt](#) – All python package list
- [README HTTP.md](#) - Detailed User Guide

How it works





Key Features

- **add_documents** - Adding documents to the vector DB
- **search_documents** - Similarity-based search
- **rag_query** - Query via RAG (using Claude), Need API key
- **delete_collection** - Delete collection

Available MCP Tools

1 add_documents

Add documents to the vector database. Automatically handles chunk splitting and embedding.

Parameters:

- texts: Array of document texts
- metadatas: Array of metadata objects
- chunk_size: Chunk size (default: 1000)
- chunk_overlap: Overlap size (default: 200)

2 search_documents

Search for similar documents in the vector database using cosine similarity.

Parameters:

- query: Search query string
- k: Number of results (default: 4)
- filter: Optional metadata filter

3 rag_query

Answer questions using RAG. Retrieves relevant documents and generates answers with Claude.

Parameters:

- question: Question text
- k: Number of docs to search (default: 4)
- language: Answer language (ko/en)

4 get_collection_info

Retrieve current collection information including document count and storage location.

Parameters:

- None (no parameters required)

5 delete_collection

Delete entire collection. All documents and embeddings will be permanently removed.

Parameters:

- confirm: Confirmation flag (must be true)

Key Features

Automatic Text Chunking

Documents are automatically split into manageable chunks (1000 chars with 200 char overlap) while preserving semantic coherence.

Smart Embeddings

Uses sentence-transformers/all-MiniLM-L6-v2 model for high-quality embeddings. Runs locally without API keys.

Persistent Storage

ChromaDB stores vectors persistently on disk. Data survives server restarts and can be queried anytime.

Similarity Search

Cosine similarity-based vector search efficiently finds the most relevant documents for any query.

Metadata Management

Store and filter documents by custom metadata (source, category, tags, etc.) for precise retrieval.

Multilingual Support

Supports multiple languages through HuggingFace embeddings and Claude's multilingual capabilities.



Technology Stack

LangChain

v1.0.7

ChromaDB

v1.3.5

Claude AI

Sonnet 4

HuggingFace

Transformers v5.1.2

MCP

v1.21.2

Python

v3.12