

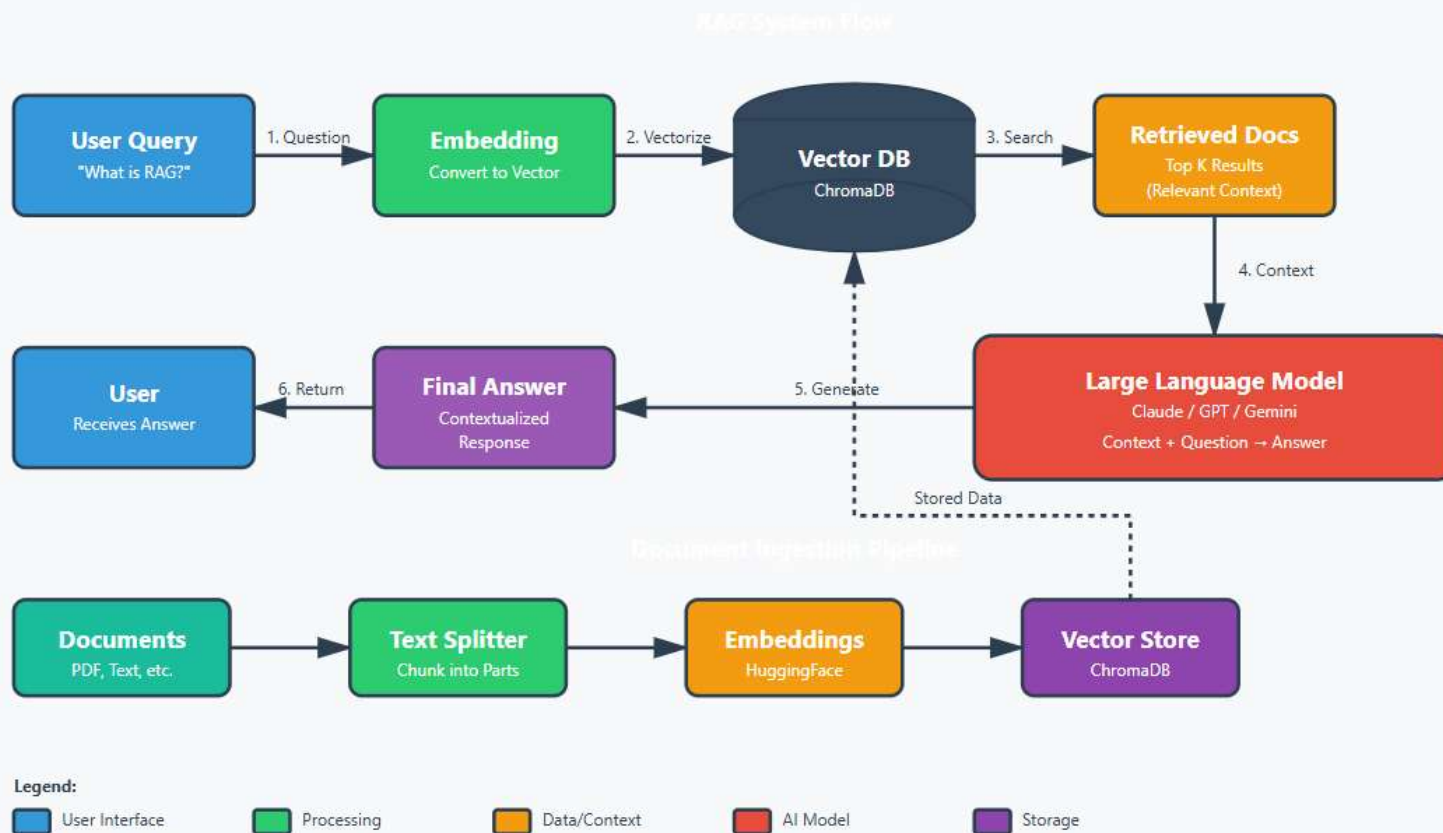
# RAG MCP 서버

With ChromaDB, LangChain, Claude LLC

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## RAG (Retrieval-Augmented Generation) Architecture





## RAG Process Flow

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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

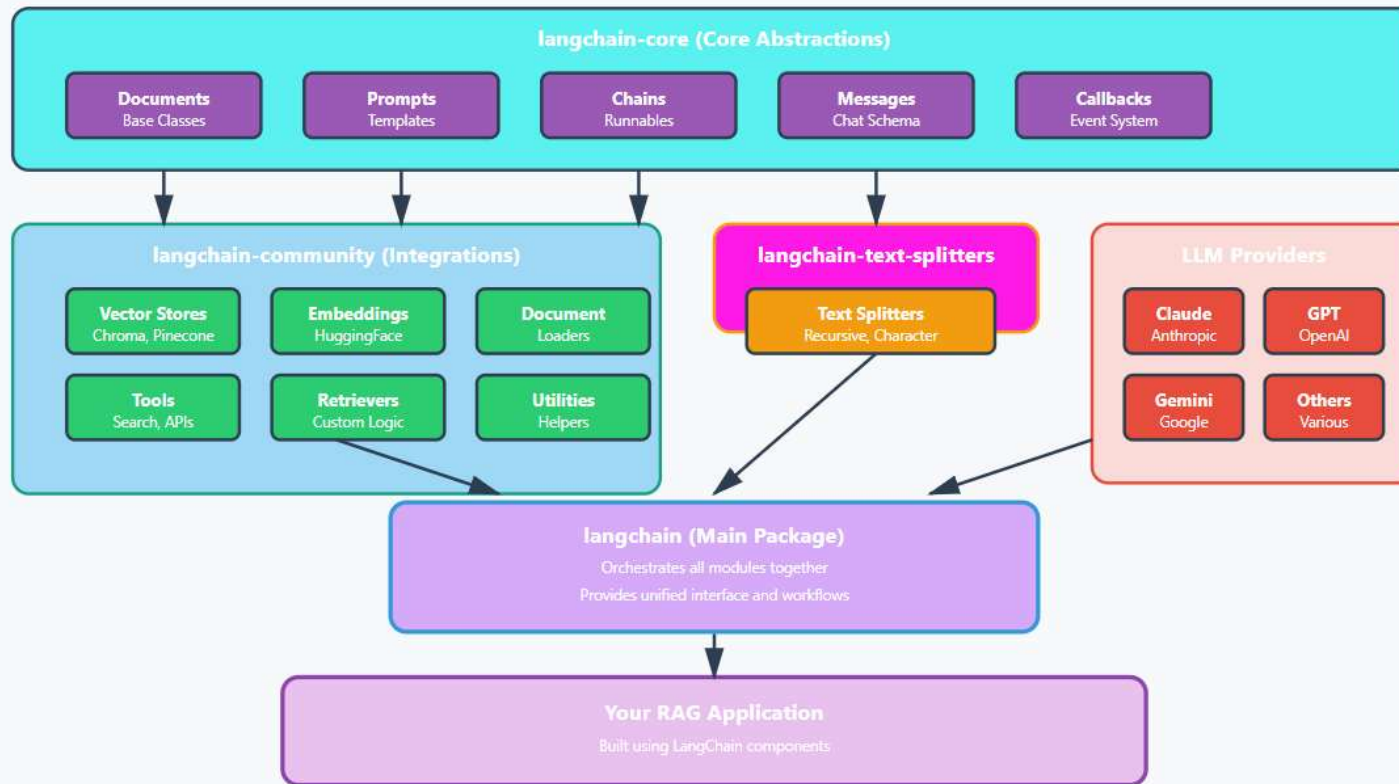
5

### 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



## 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>
```

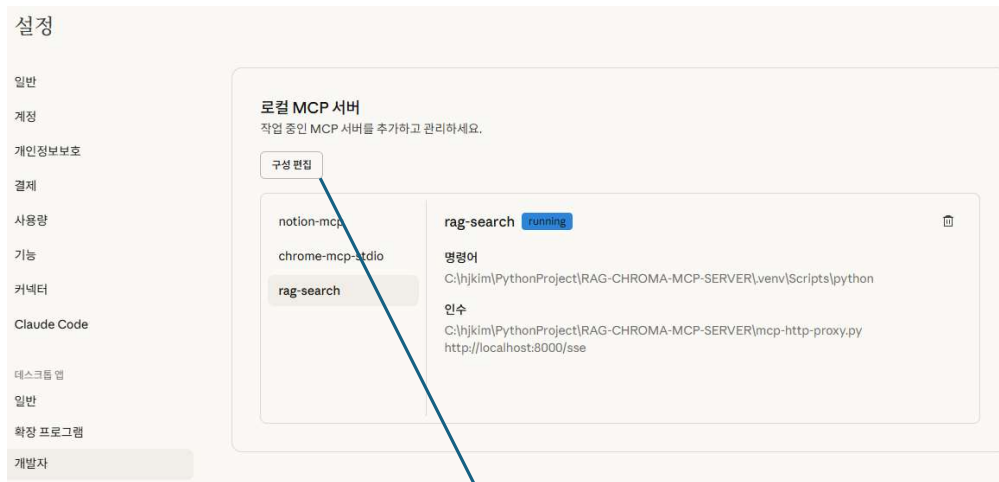
#Check server is working properly

```
(.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\\rag_mcp_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:

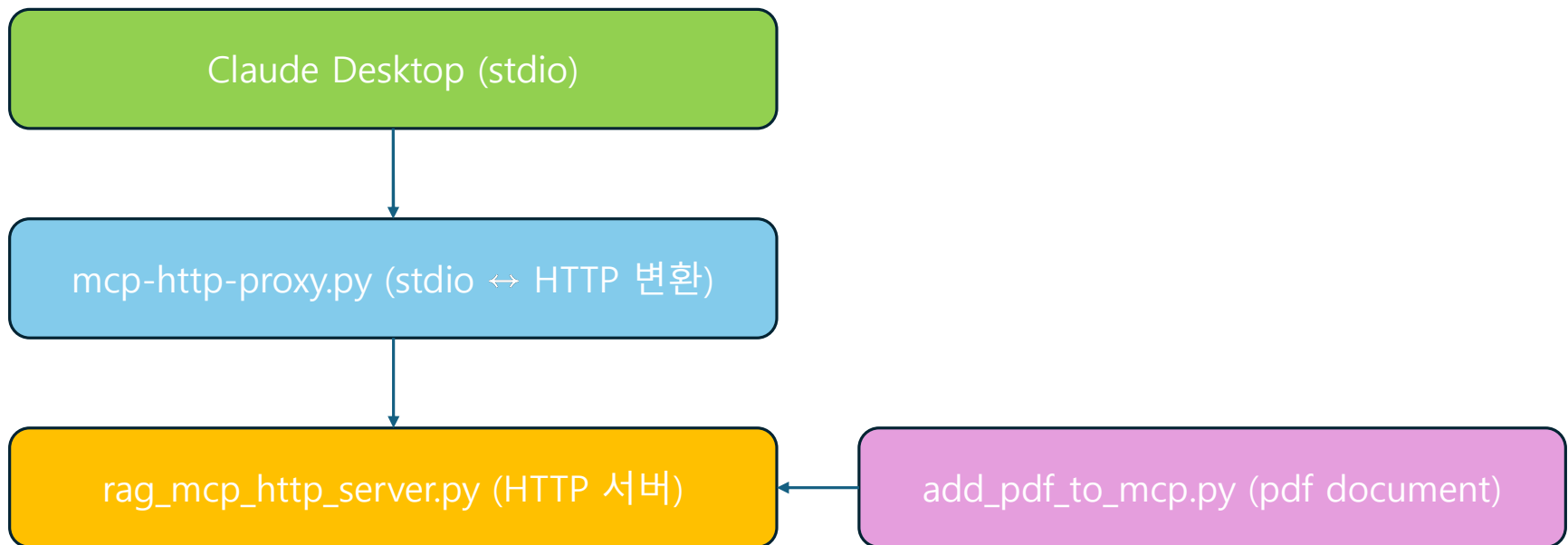
Rag query를 활성화 하려면 .env 파일을 생성한 후 아래와 같이 추가  
ANTHROPIC\_API\_KEY=your-api-key-here



# 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

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### 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

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**LangChain**

v1.0.7

**ChromaDB**

v1.3.5

**Claude AI**

Sonnet 4

**HuggingFace**

Transformers v5.1.2

**MCP**

v1.21.2

**Python**

v3.12