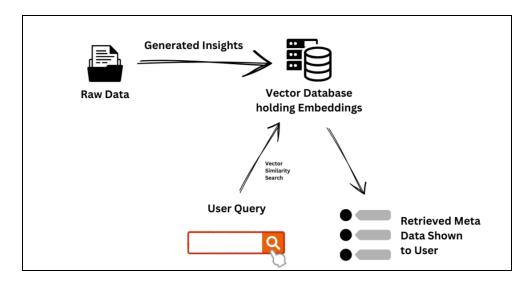
Solution Overview: Advanced File Search System



To address the task of creating an advanced file search system using an agent-based approach, I implemented a solution that leverages both file metadata and dynamically generated insights to intelligently respond to user queries. The system works as follows:

1. Data Representation and Storage:

- Each file in the system is represented with metadata fields such as the author, source (e.g., Google Drive), file title, size, type, location, creation date, last update date, and URL.
- To enhance search capabilities, a brief description or "insight" is generated for each file using either static templates (based on file type, title, or source) or through a generative model.
- The insights are converted into embeddings and stored in a vector database (Chroma DB). This allows efficient similarity-based search during user gueries.

2. User Query Processing:

- When a user submits a query, the system extracts specific constraints from the input, such as file type (e.g., PDF), file source (e.g., Google Drive), time constraints (e.g., files uploaded in June), or size constraints (e.g., files smaller than 1MB).
- These constraints help refine the search scope and filter the files accordingly.
- The user's query is also summarized and converted into an embedding,

which is then used to find the most similar file descriptions in the vector database.

3. Matching and Results:

- Based on the similarity between the query embedding and the stored file embeddings, the system retrieves the most relevant files, along with their corresponding metadata (such as file title, size, date, and URL).
- The metadata is displayed to the user, allowing them to quickly access or download the files that best match their search criteria.

By combining metadata extraction, embedding-based search, and intelligent filtering, this approach provides a robust and efficient solution for advanced file search queries.

Possible Advancements:

1. Stronger Vectorization Techniques:

- Advanced Embedding Models: Instead of basic vectorization methods, more powerful models like BERT, Sentence Transformers, or OpenAI's embedding models could be used for richer, context-aware representations of file insights and user queries. These models would capture deeper semantic relationships between queries and stored descriptions, improving the system's ability to understand and match complex queries.
- Fine-tuning on Domain-Specific Data: Fine-tuning the vectorizer on domain-specific datasets (e.g., document types and query patterns unique to the business) could further enhance the accuracy and relevance of search results.