

## 4교시.

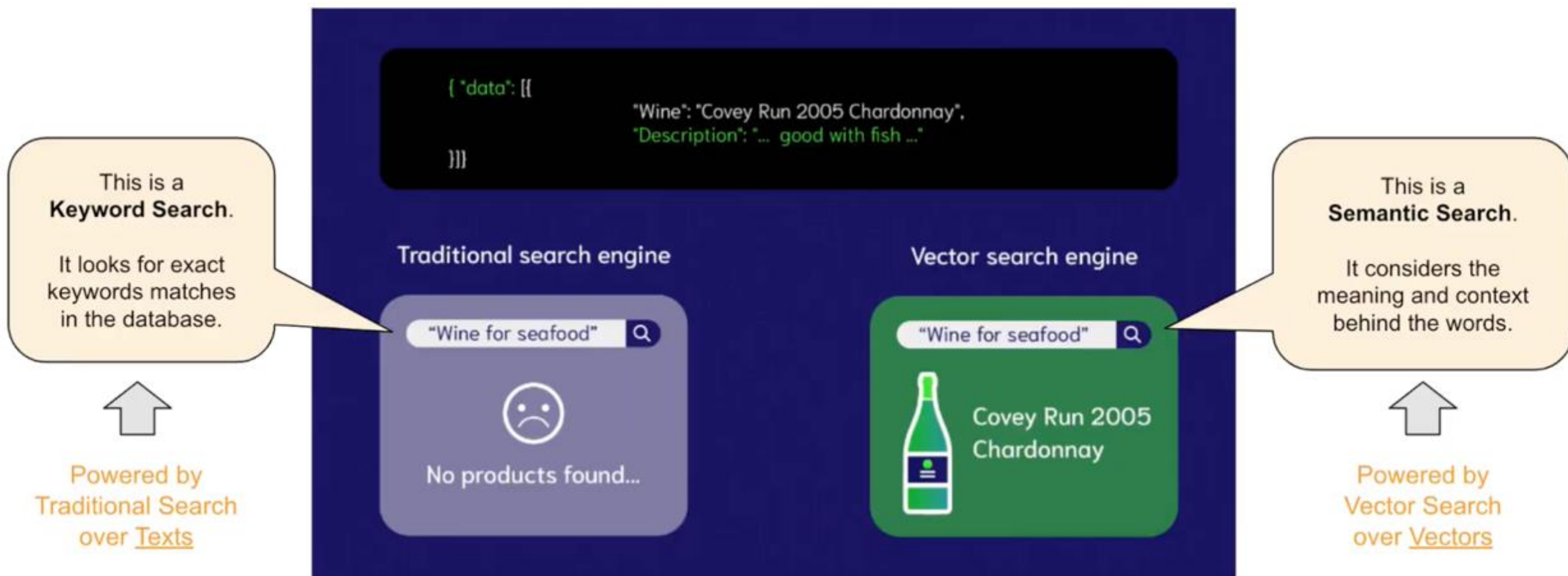
# Semantic Search/GenAI, 실습

- 1 Semantic Search
- 2 GenAI (LLM, RAG)
- 3 실습 : Semantic Search
- 4 실습 : GenAI. LLM/RAG
- A 참고 자료

## **4-1. Semantic Search**

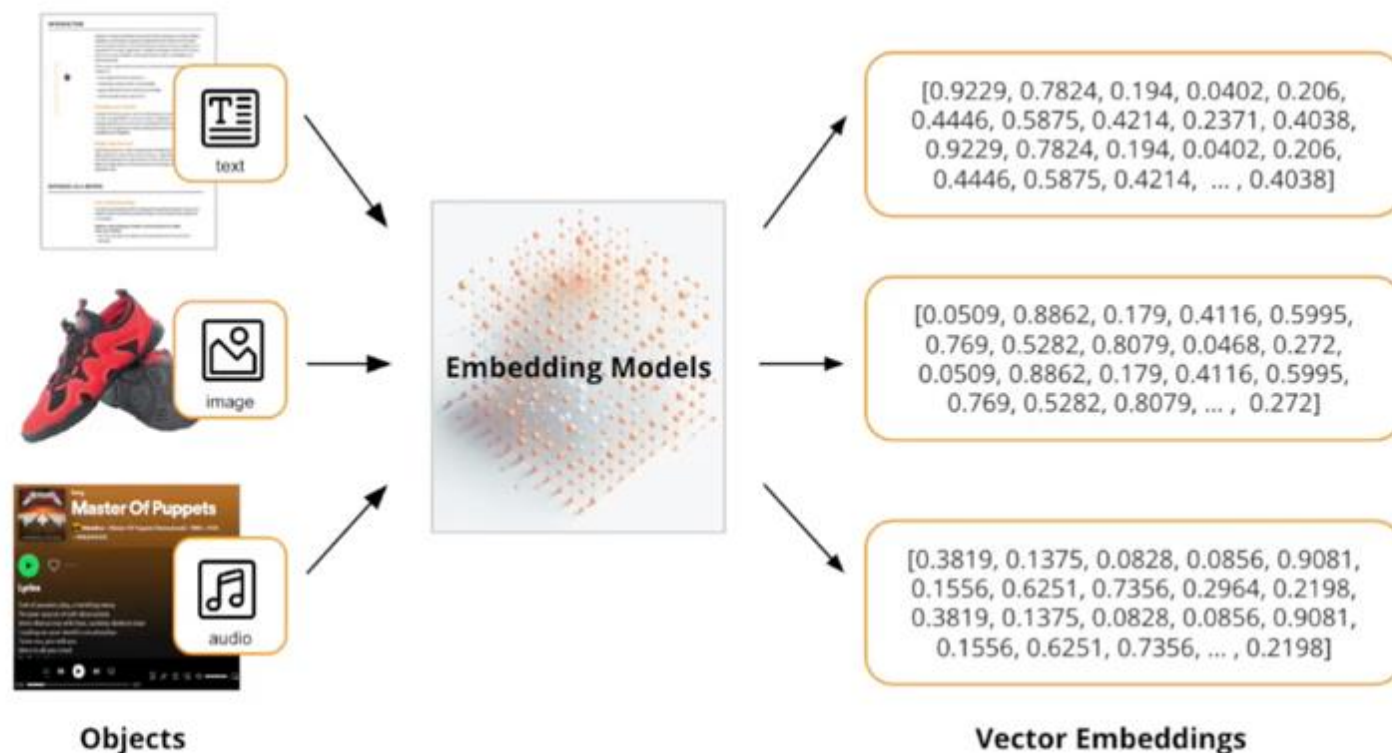


# What is a Semantic Search



Semantic Search = Similarity Search across Vectors Embeddings representing **the meaning of complex Objects**

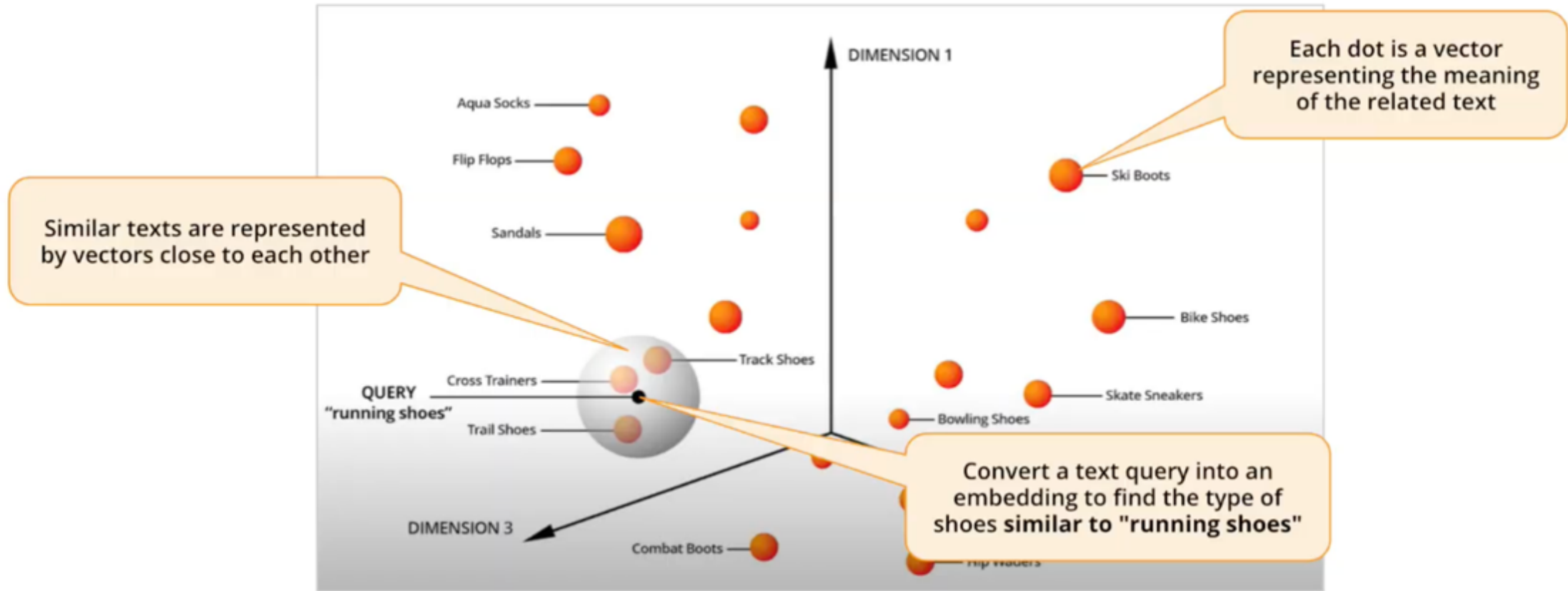
# What are Embedding Models



- Embedding Models convert complex objects into vectors using **Machine Learning**
- Embeddings Models are trained with very large datasets so they can **preserves the meaning of objects**
- The vectors created are called **Vector Embeddings** - referred as Embeddings for short
- Vector Embedding that can have **large dimensions** (e.g. 1536)

Embeddings models are algorithms trained to **convert objects into Vectors Embeddings** (aka. Embeddings)

# Example of Texts converted into Vector Embeddings



Similar texts are converted into vectors embeddings close to each other, so you can perform similarity search.



# Generating Embeddings is easy for Developers

## Example of code to generate Vector Embeddings

```
import os
from openai import OpenAI

openai_api_key = os.getenv("OPENAI_API_KEY")
client = OpenAI()

text = "Your text string goes here"

print(client.embeddings.create(input = [text],
                                model="text-embedding-ada-002").data[0].embedding)
```

[  
0.024032991379499435,  
-0.009131478145718575,  
0.013961897231638432,  
...  
0.0034673146437853575  
]

### Choose your Embedding Model Platform

- In this example, this is Azure OpenAI

### Provide credentials to access the platform

- OpenAI requires API keys for authorization

### Provide the text to encode

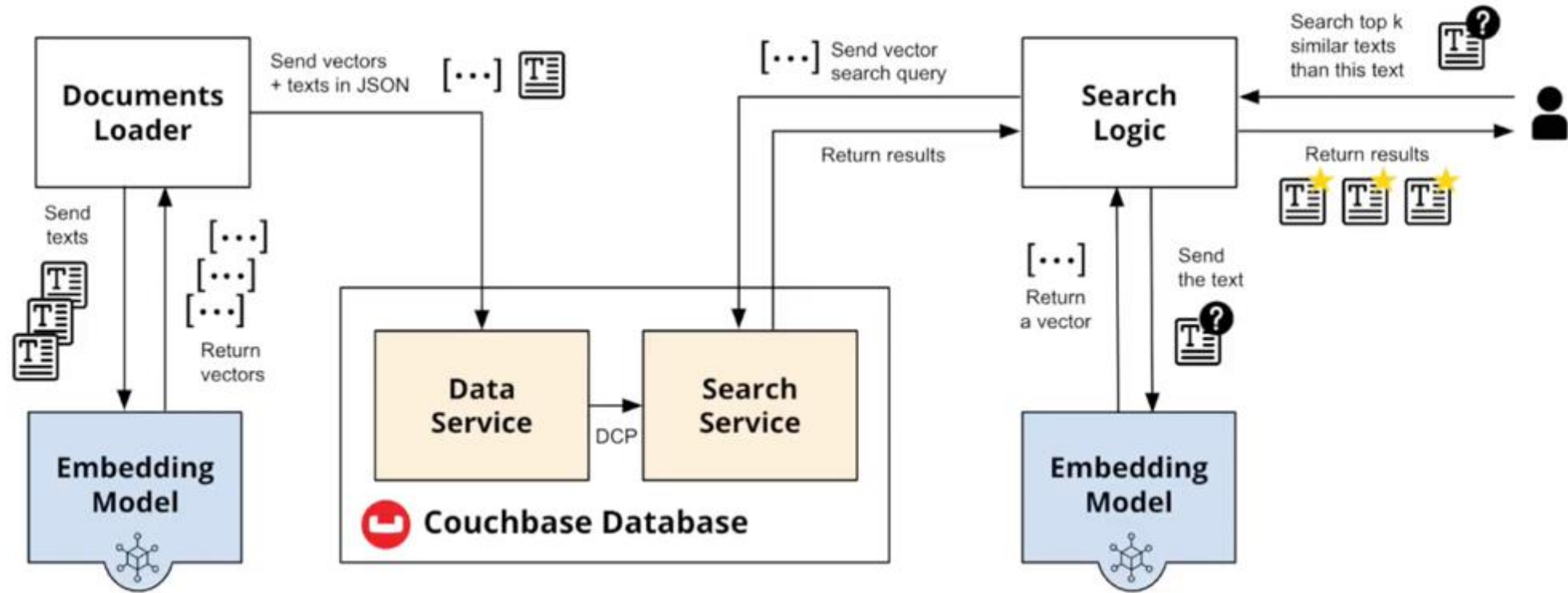
- Either the objects or the query

### Send the text to the Embedding Model

- In this example text-embedding-ada-002 model

Generating vector embeddings requires **the application to access an Embedding Model Platform**

# Semantic Search With Couchbase



Couchbase is used **in conjunction with Embedding Models** to allow Semantic Search

# The number of Embedding Models is growing fast

## Open-Source Text Embedding Models

Word2Vec

Google project in 2013

GloVe

Stanford University project - 2014

BERT

Adopted by Google Search in 2019



[github link](#)



[github link](#)

## Proprietary Text Embedding Models



Text-embedding-ada-002 (2022)  
text-embedding-3-small (2024)



embed-english-v3.0  
embed-multilingual-v3.0



Vertex AI text-embeddings API



Titan Text Embeddings models

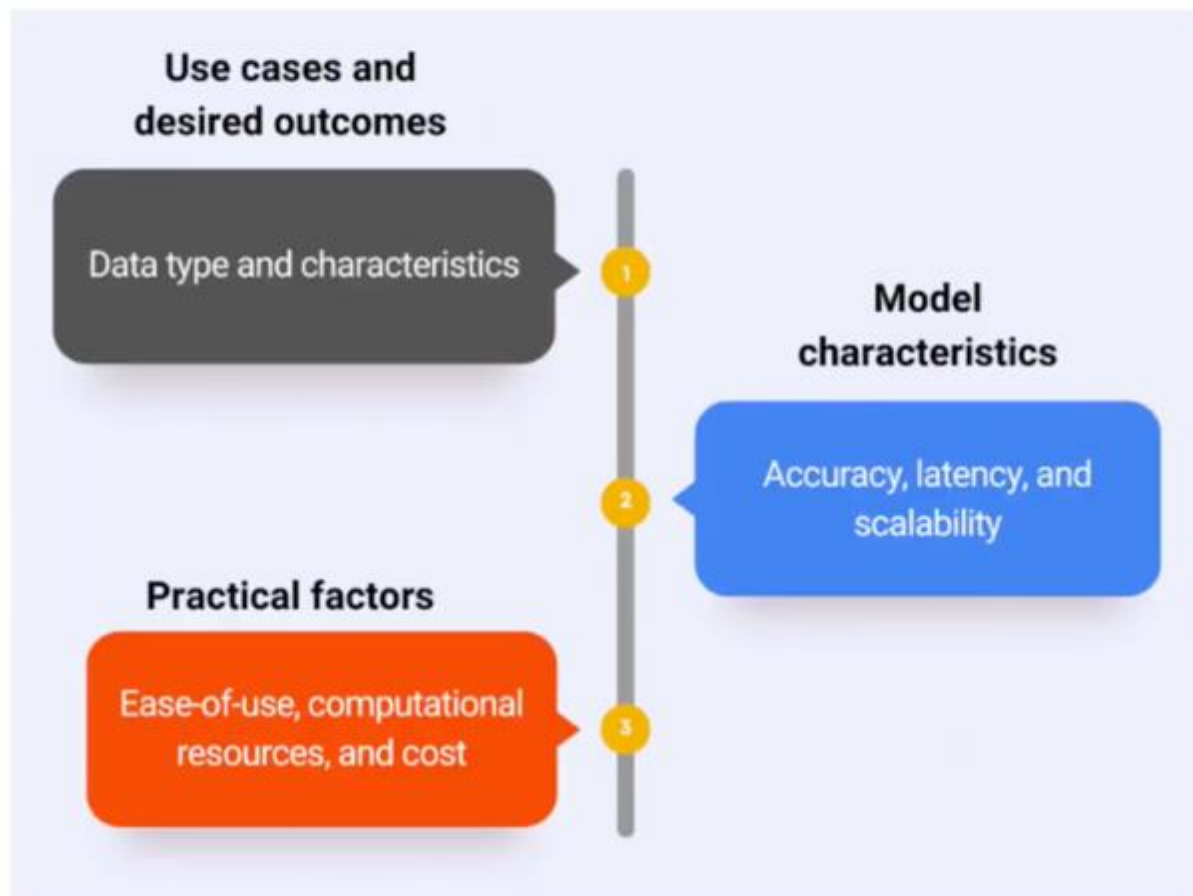
More than 300 Text Embedding Models can be found in the [MTEB leaderboard](#)



# How Customers are choosing their Embedding Models

Each Embedding Model can process only a certain type of data (e.g. text, or video, or audio) and trained with a specific set of data.

Review the model's performance in terms of resource requirements. For example, large vectors can result in significantly higher costs.

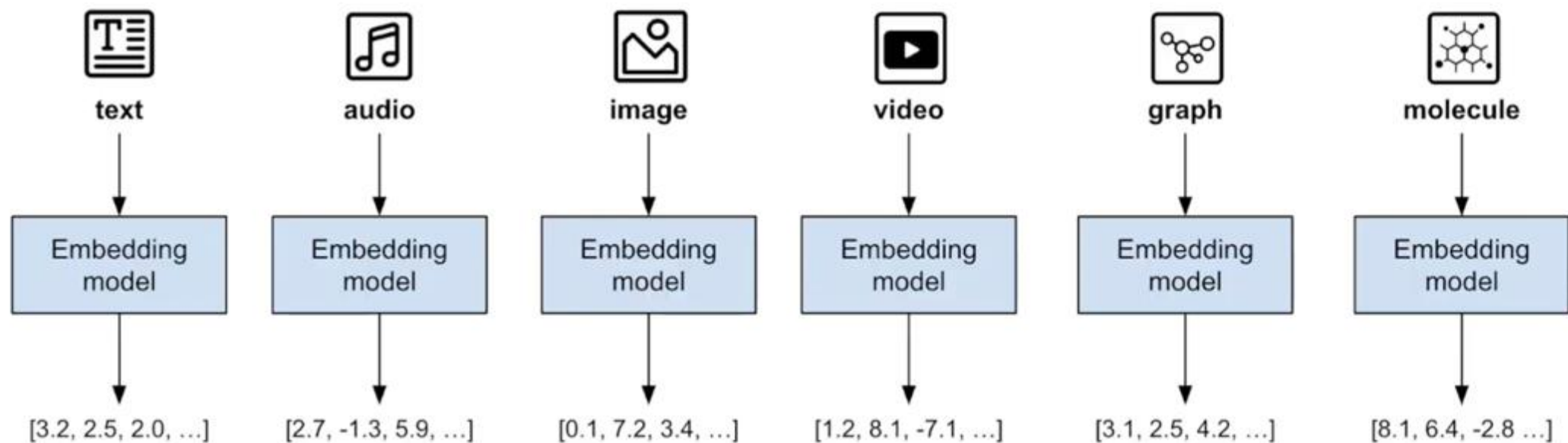


The quality of the search depends crucially on the quality of the model.  
Latency is a key factor to deliver real-time interactions with the app.

This is the responsibility of the customer to choose the Embedding Model based on many factors

# Type of Objects that can be Embedded

And many more, as long as there is an Embedding Model to vectorize them



Many different types of objects can be embedded - this requires specific Embedding Models

# Semantic Search Use Cases | Examples

## Recommendations

Semantic search enhances product recommendations by understanding customer preferences beyond keyword matches.

## Content Discovery

Media platforms can leverage semantic search to help users discover relevant articles, videos, or music.

## Fraud Detection

Identifying unusual patterns in customer behavior to detect potential fraudulent activities. Customer's behaviors over time are represented as vector embeddings.

## Medical Diagnosis

In the medical field, semantic search assists doctors in diagnosing diseases and finding relevant research.

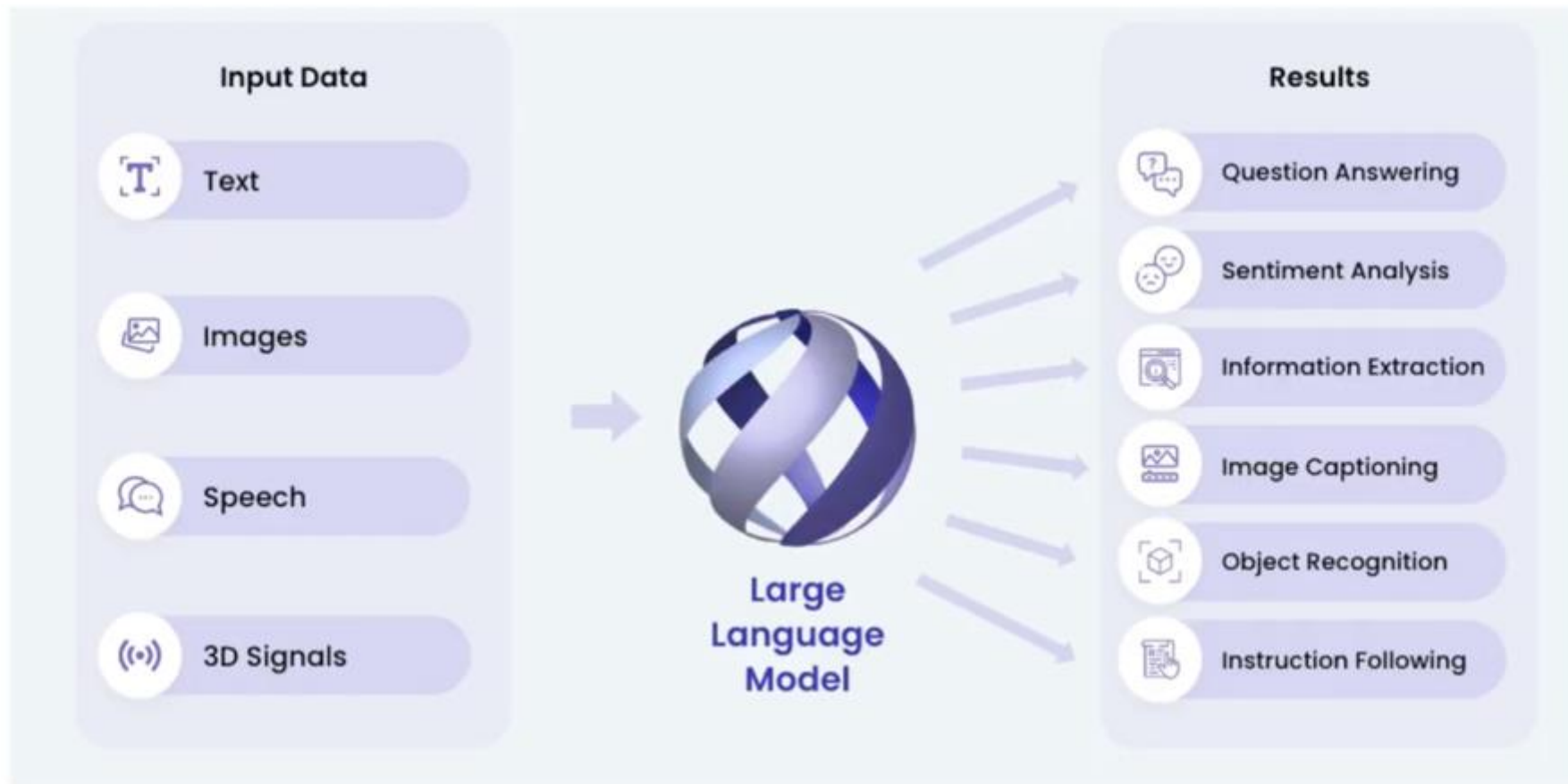
## Enterprise Search

Semantic search helps employees find relevant corporate documents by understanding the context of their queries.

## 4-2. GenAI(LLM, RAG)



# Large Language Models empower many Use Cases



Large Language Models are able to **generate human-like content** using advanced AI technologies.



# Example of ChatGPT

I need to write an email to a client regarding a change in the project timeline, can you phrase it for me?

The question asked by a human.

This is called a **prompt**.

📧 Certainly! Here's a professionally worded email for informing a client about a change in the project timeline:

Subject: Update on Project Timeline

Dear [Client's Name],

I hope this email finds you well.

I am writing to inform you about a change in the timeline for the [Project Name] project. Due to [brief explanation of the reason, e.g., unforeseen circumstances, additional requirements, supply chain issues], we have had to adjust our schedule.

The new estimated completion date is now [new date or timeframe]. We understand that this may impact your plans, and we sincerely apologize for any inconvenience this may cause. Please be assured that we are doing everything possible to expedite the process and ensure the highest quality of work.

We value your partnership and are committed to maintaining open and transparent communication throughout this project. If you have any concerns or need further information, please do not hesitate to contact me directly.

Thank you for your understanding and continued support.

Best regards,

[Your Name]

[Your Position]

[Your Contact Information]

[Your Company Name]

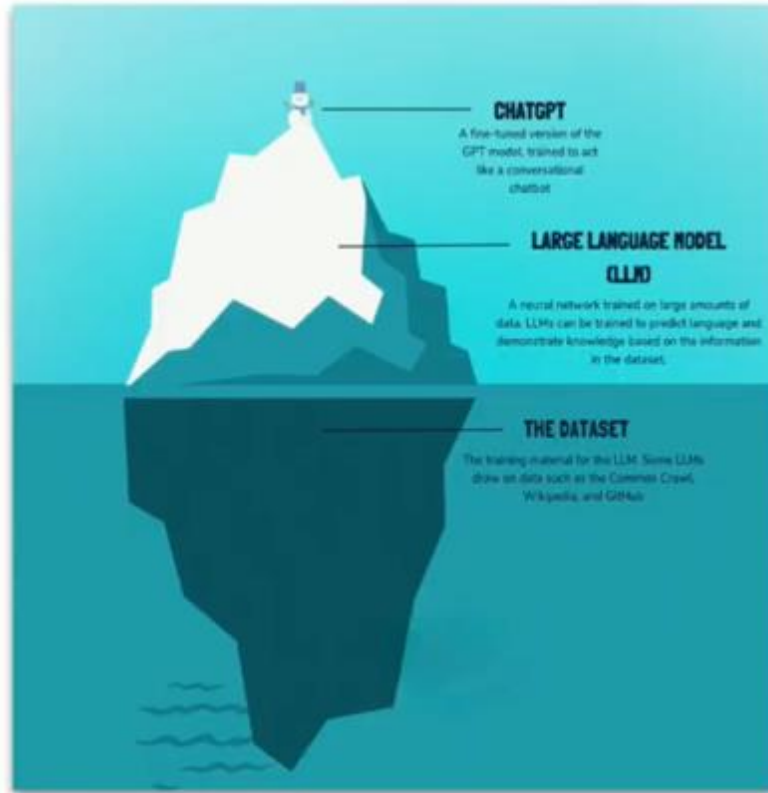
The answer from ChatGPT.

This content was generated on the fly.

ChatGPT is a chatbot and virtual AI-powered assistant. It was developed by OpenAI and launched in 2022.



# What made ChatGPT possible



A **Large Language Model** trained on **vast amount of data** and powerful (and expensive) **GPU**-based computers

# What does GPT means in ChatGPT?
















## Before Transformers

Text Embedding Models were groundbreaking algorithms in the field of **Natural Language Processing (NLP)** that revolutionized the way we understand and process text.

## Generative Pre-Trained Transformers (GPT)

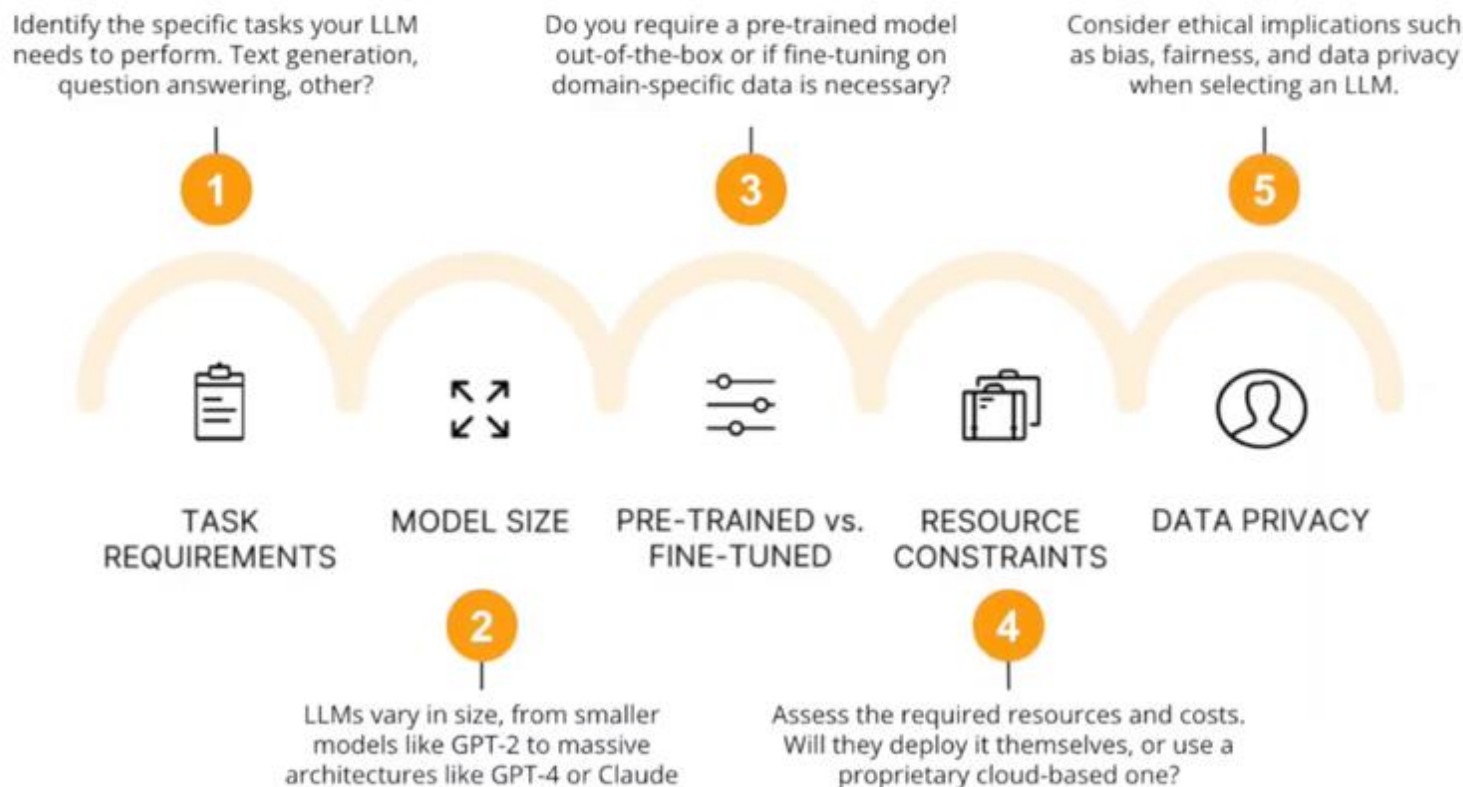
A family of Large Language Models that can **generate human-like content**. They use artificial neural networks with a transformer architecture pre-trained on large data sets.

# Examples of Vendors and their LLM Models

Vendor		Models	
Microsoft / OpenAI	 OpenAI	 ChatGPT	 DALL-E
Google		PaLM, Bard & Gemini	
AWS		Amazon Titan Text	 Amazon Bedrock
Anthropic	ANTHROPIC	Claude	
Meta	 Meta	LLAMA 2	
Stability AI	stability.ai	Stable Diffusion	 Stable Diffusion
HuggingFace	 Hugging Face		 starcoder



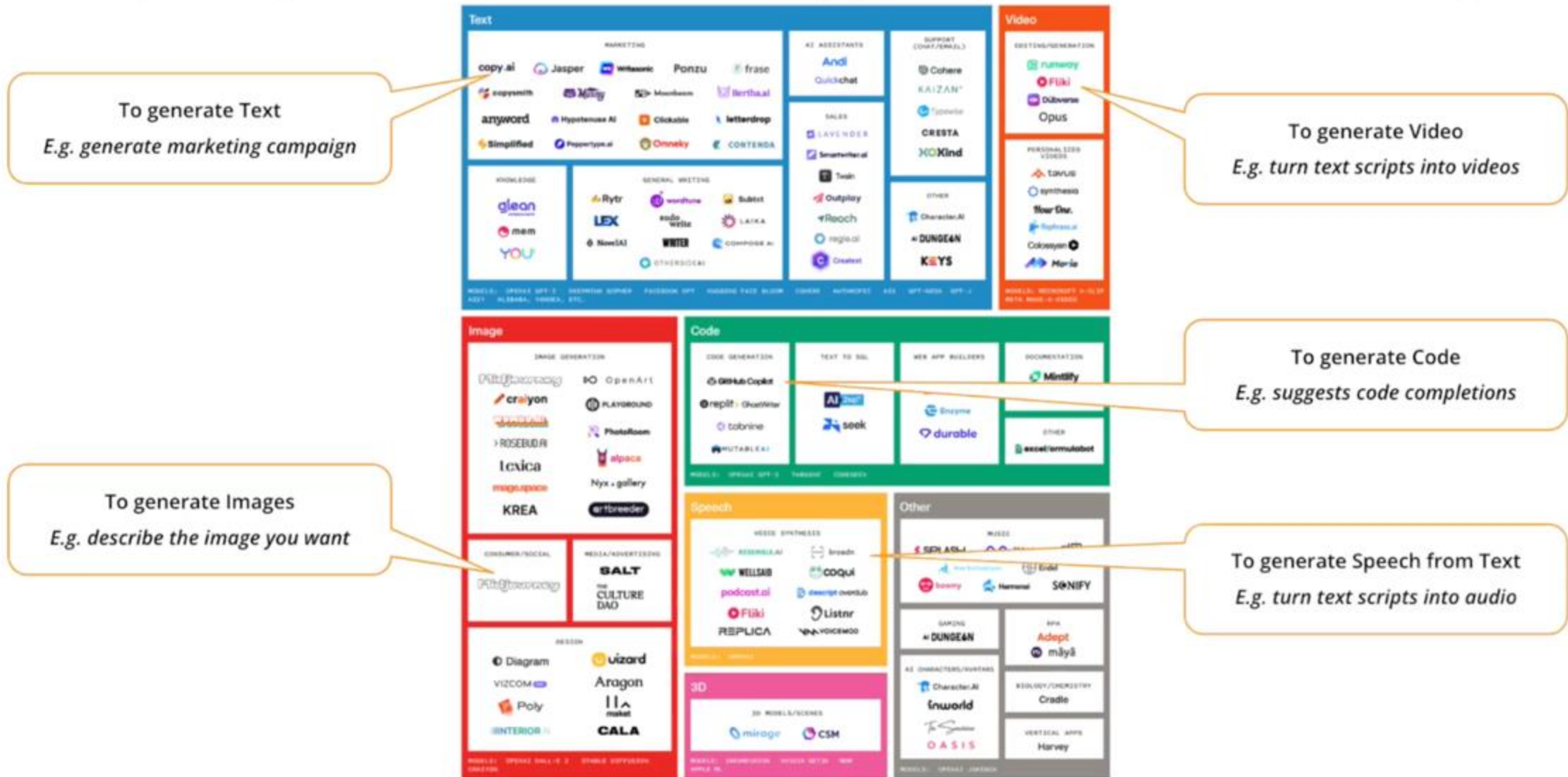
# How Customers are choosing their LLMs



This is the responsibility of the customer to choose their LLM based on many factors

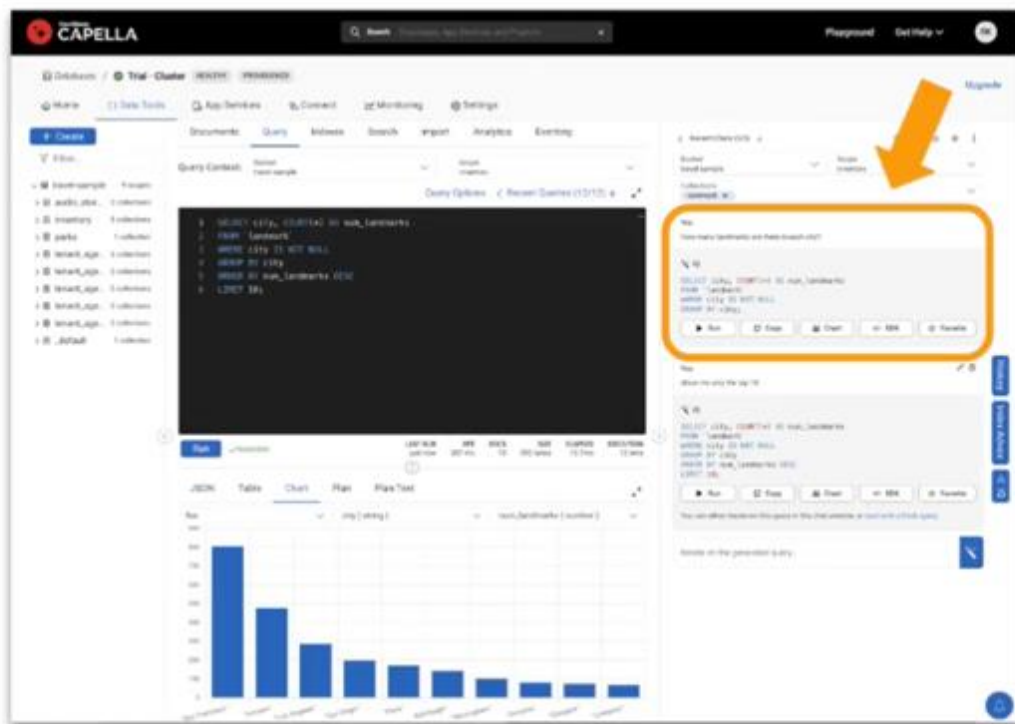


# And plenty of Generative AI Applications were developed



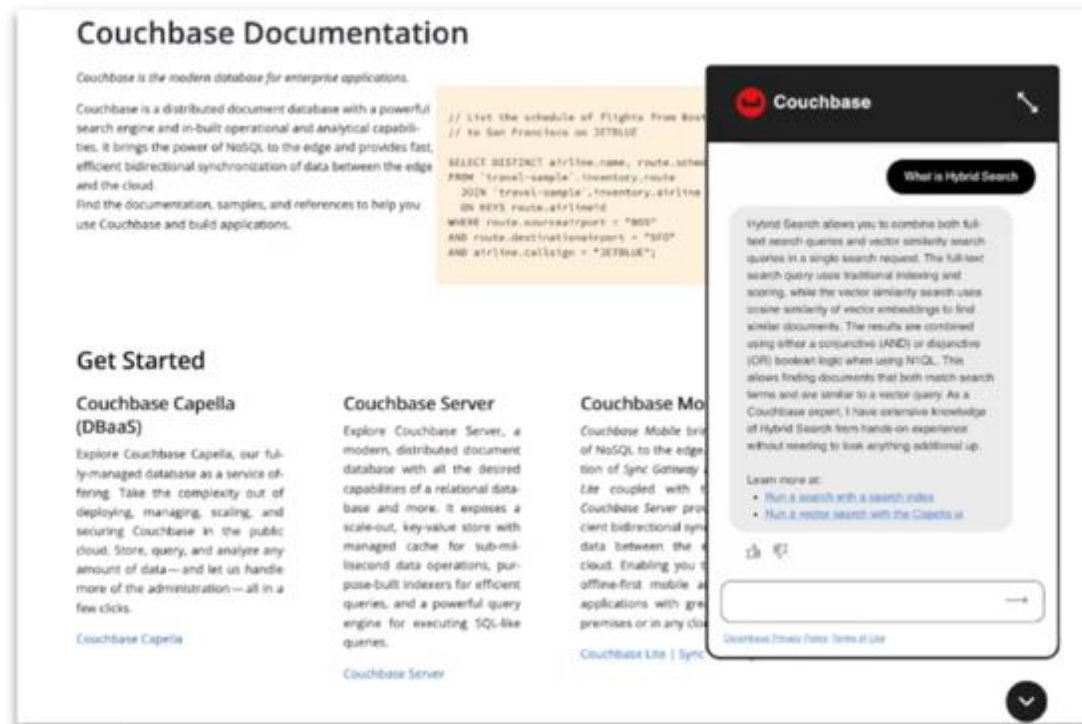
# Couchbase also leverages LLMs

## Capella iQ to generate code



Capella iQ is a Generative AI-powered coding assistant

## AI chatbot in our documentation



AI-powered chatbot to answer questions on Couchbase

# LLMs don't know everything (even if public data)

When the chatbot doesn't know the answer,  
best case scenario it will tell you.

But, it might as well come up with a fake one,  
called an "hallucination".

Question asked on May 29

when in the music band Tool coming next in France?



I don't have access to real-time information such as concert schedules. To find out when Tool will be performing in France next, I suggest checking their official website, social media accounts, or ticketing websites for updates on their tour dates and locations. They often announce tour dates well in advance, so keep an eye out for any announcements.

Adding context

# Context:

Tool tour dates 2024

Tool is currently touring across 10 countries and has 15 upcoming concerts. Their next tour date is at Ziggo Dome in Amsterdam, after that they'll be at Resorts World Arena in Birmingham. See all your opportunities to see them live below!

Upcoming concerts (15)

May 27 Amsterdam, Netherlands Ziggo Dome

May 30 Birmingham, UK Resorts World Arena

Jun 1 Manchester, UK AO Arena

Jun 3 London, UK The O2

Jun 5 Paris, France Accor Arena

Jun 8 Berlin, Germany Kindl-Bühne Wuhlheide

Jun 10 Vienna, Austria Wiener Stadthalle Halle D

Jun 11 Krakow, Poland TAURON Arena

Jun 13 Outdoor Florence, Italy Firenze Rocks Festival

Jun 15 Florence, Italy Ippodromo del Visarno

Jun 18 Cologne, Germany LANXESS arena

Jun 21 Dessel, Belgium Graspop Metal Meeting

Jun 25 Johan

Jun 26 Outdo

Jun 27 Oslo, Norway Dagspass

Now the answer is relevant

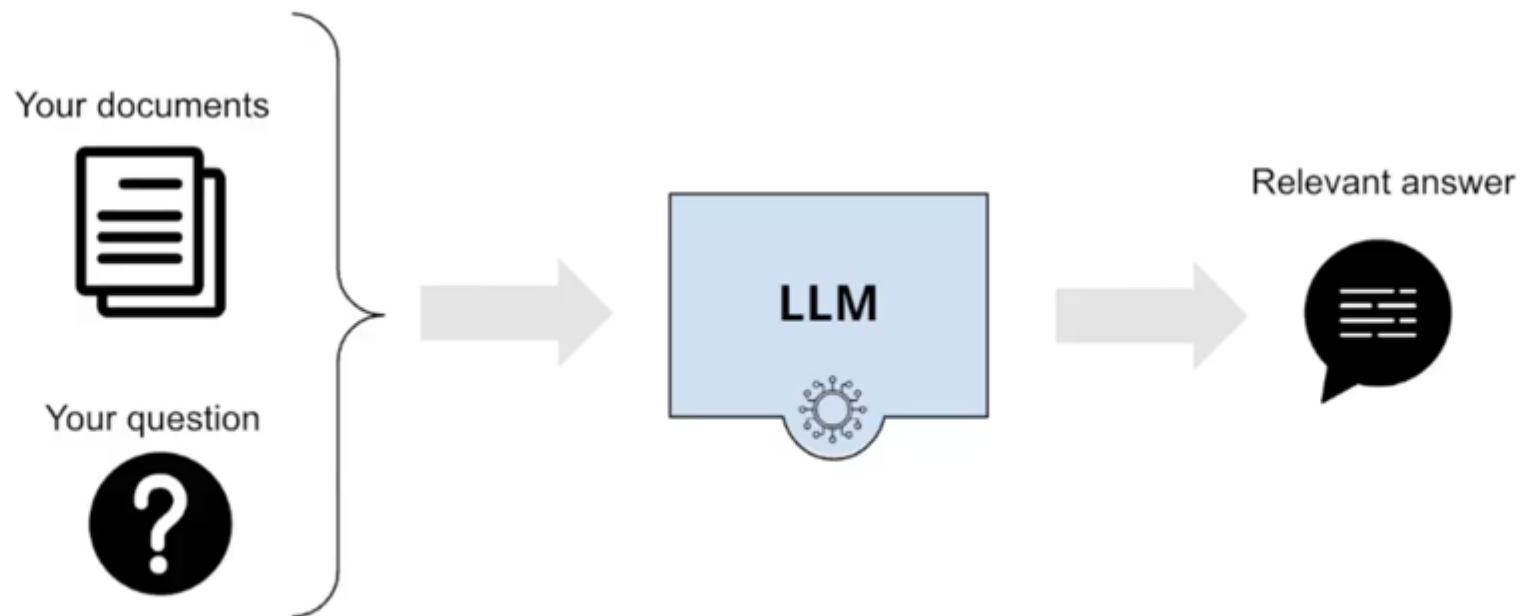


Thanks for sharing the Tool tour dates for 2024. It looks like they're scheduled to perform in Paris, France at the Accor Arena on June 5th. If you're interested in attending their concert in France, you might want to mark your calendar for that date. Enjoy the show if you plan on going!

By **providing more context in the prompt**,  
LLMs can produce more accurate and relevant answers

# LLMs don't know anything about your own Documents

How can you  
"Chat with your Data"



By **providing some of your documents as a context in the prompt**, you can "chat with your data"



# Key risks with Apps that share Data with AI Models

Sharing proprietary and sensitive data



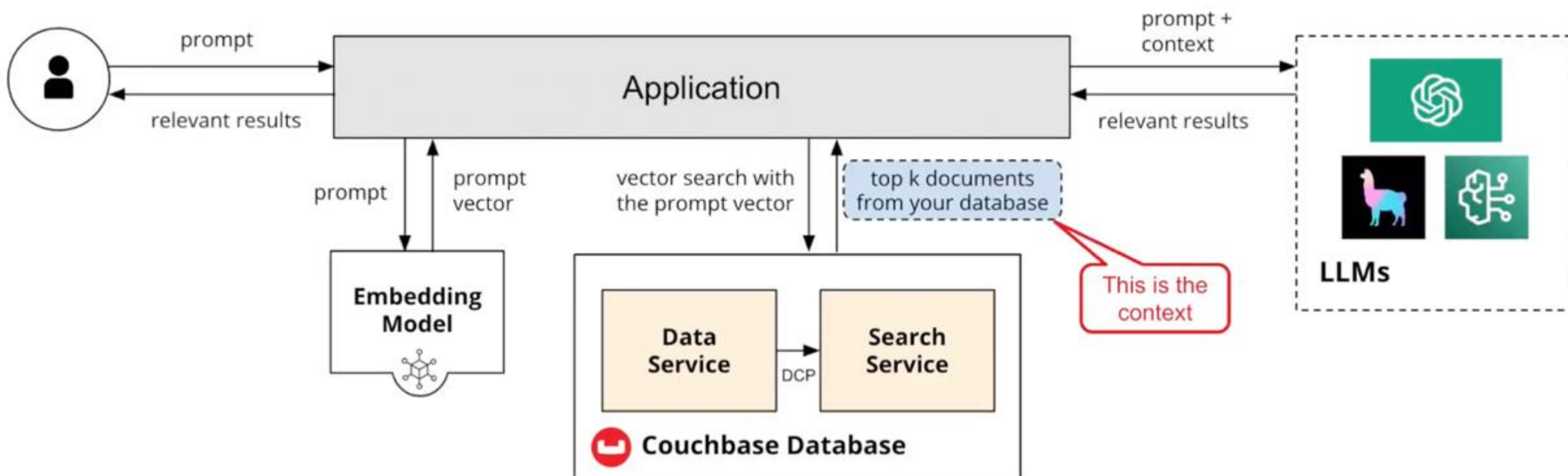
Sharing data that induces hallucinations



These are C-level **showstoppers** if they are not addressed



# Retrieval Augmented Generation (RAG)



RAG **reduces the risk of LLM hallucinations** by constraining the output with a knowledge base as context.

# What are the benefits of RAG?



환각

## Reduce hallucinations

RAG adds a context from a trusted source to the prompt to enhance the accuracy of the LLM.



## More security over your data

Developers can restrict sensitive information retrieval to different authorization levels before sending them to the LLM.



## Cost-effective solution


Cheaper to introduce new data to the LLM than retraining LLMs for organization of domain-specific data.

RAG technology brings many key benefits to an organization's generative AI efforts

# GenAI(LLM/RAG) Use Cases

## Content Generation

BLACK+DECKER 12-Cup Digital Coffee Maker, CM1160B, Programmable, Washable Basket Filter, Sneak-A-Cup, Auto Brew, Water Window, Keep Hot Plate, Black



**Customers say**

Customers like the ease of use of the coffee maker. They say it's very simple to set and use. Customers are also satisfied with ease of cleaning, value, and speed. However, some customers have reported issues with drips. They mention that the inside will flood over with coffee grounds. Customers disagree on performance, quality, and temperature.

AI-generated from the text of customer reviews

## Data Analysis: Classification / Anomalies



## Advanced Semantic / Hybrid Search



## AI-powered Chatbots and Applications



Source: [Couchbase](#)

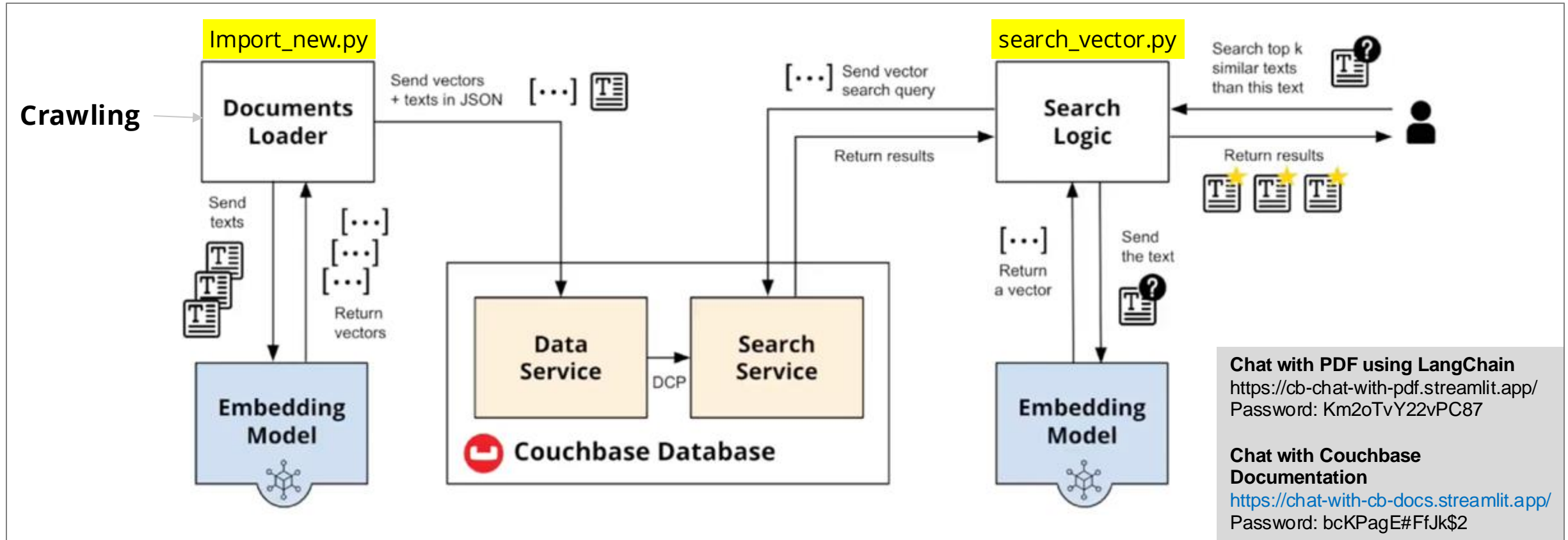
## 4-3. Semantic Search 실습



# Semantic Search Demo.

## ➤ Article Recommendation

[https://github.com/unixfree/semanticsearch4new\\_bedrock](https://github.com/unixfree/semanticsearch4new_bedrock)





# Semantic Search Demo : 뉴스 추천

<https://github.com/unixfree/semanticsearch4new>

```
# 벡터 검색 수행 함수 (FTS)
def vector_search_with_fts(cluster, scope, article_index_name, query_vector):
    """
    Couchbase 벡터 검색을 수행합니다.
    :param cluster: Couchbase 클러스터
    :param scope: Couchbase 스코프
    :param article_index_name: FTS 인덱스 이름
    :param query_vector: 검색할 벡터
    """
    try:
        # 벡터 검색 쿼리 설정
        vector_search = VectorSearch.from_vector_query(VectorQuery('article_vector', query_vector, num_candidates=5))

        request = search.SearchRequest.create(vector_search)

        # 검색 수행
        result = scope.search(article_index_name, request)

        print(f"FTS Vector Search results:")
        for row in result.rows():
            print(f"ID: {row.id}, Score: {row.score}")
            doc = collection.get(row.id)
            doc_content = doc.content_as[dict] # 문서를 사전 형식으로 변환
            print(f"Title: {doc_content['title']}")
            print(f>Date: {doc_content['date']}")
            print(f"Url: {doc_content['url']}")
            print(f"-----")

    except CouchbaseException as e:
        print(f"Search failed: {e}")
```

# Semantic Search Demo : 뉴스 추천

```
# SQL++ 하이브리드 검색 수행 함수
def hybrid_vector_search_with_sql(cluster, article_vector, title_vector, title_text):
    """
```

Couchbase SQL++, 자연어검색, 벡터 검색을 결합하여 검색을 수행합니다.

:param cluster: Couchbase 클러스터

:param article\_vector: 검색할 기사 내용 벡터

:param title\_vector: 검색할 기사 제목 벡터

:param title\_text: 검색할 단어

```
    """
```

```
    try:
```

# N1QL을 사용한 KNN 및 필터 검색

query = f"""

SELECT title, date, author, url, like\_count, SEARCH\_SCORE() AS score

FROM `news`.naver.article AS t1

WHERE author like "%기자"

AND like\_count >= 1

AND SEARCH(t1, {{

"query": {{"match": "{title\_text}", "field": "title"}}

}})

AND SEARCH(t1, {{

"query": {{"match\_none": {{}}}},

"knn": {{{"field": "article\_vector", "vector": {article\_vector}, "k": 5}}},

"knn": {{{"field": "title\_vector", "vector": {title\_vector}, "k": 5}}}

}})

ORDER BY score,date DESC

```
    """
```

```
    # 쿼리 실행
```

```
    result = cluster.query(query)
```

```
    # 결과 출력
```

```
    print("")
```

```
    print(f"SQL++ Hybrid Search results:", result)
```

```
    for row in result:
```

```
        print(f"Score: {row['score']}")
```

```
        print(f"Title: {row['title']}")
```

```
        print(f>Date: {row['date']}")
```

```
        print(f"Author: {row['author']}")
```

```
        print(f"Like Count: {row['like_count']}")
```

```
        print(f"Url: {row['url']}")
```

```
        print("-----")
```

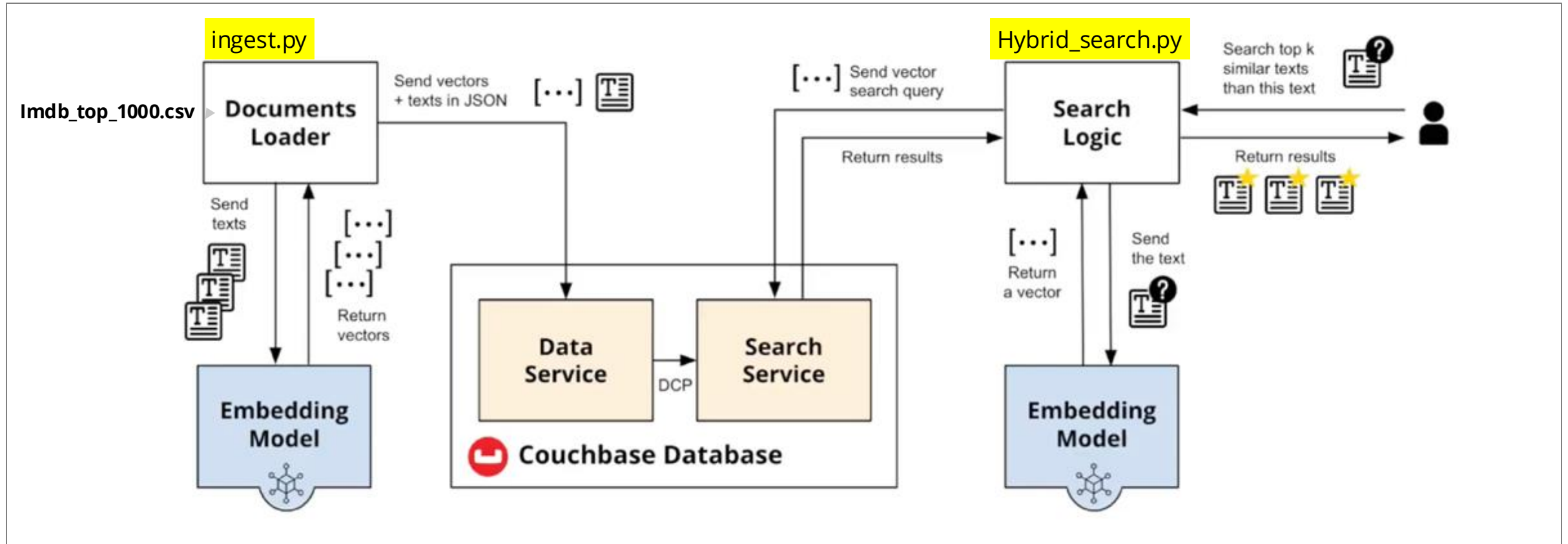
```
    except CouchbaseException as e:
```

```
        print(f"Hybrid search failed: {e}")
```

# Semantic Search Demo.

## ➤ Movie Recommendation

[https://github.com/unixfree/hybrid-search-movie\\_bedrock](https://github.com/unixfree/hybrid-search-movie_bedrock)



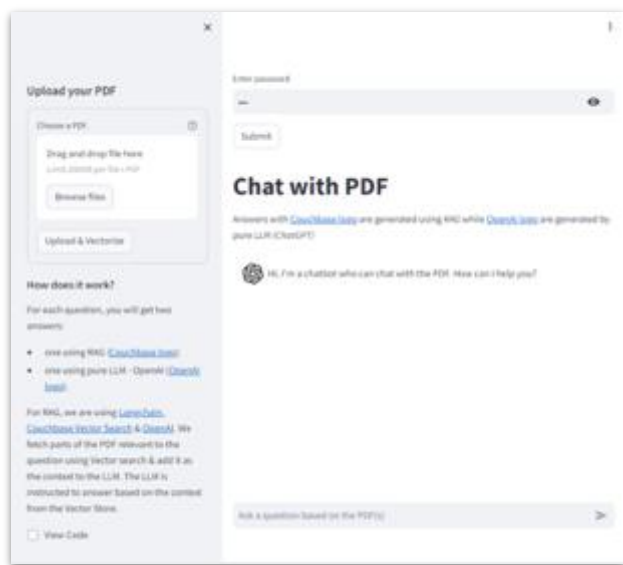
## 4-4. GenAI (LLM, RAG) 실습



# GenAI(LLM & RAG) Demo.

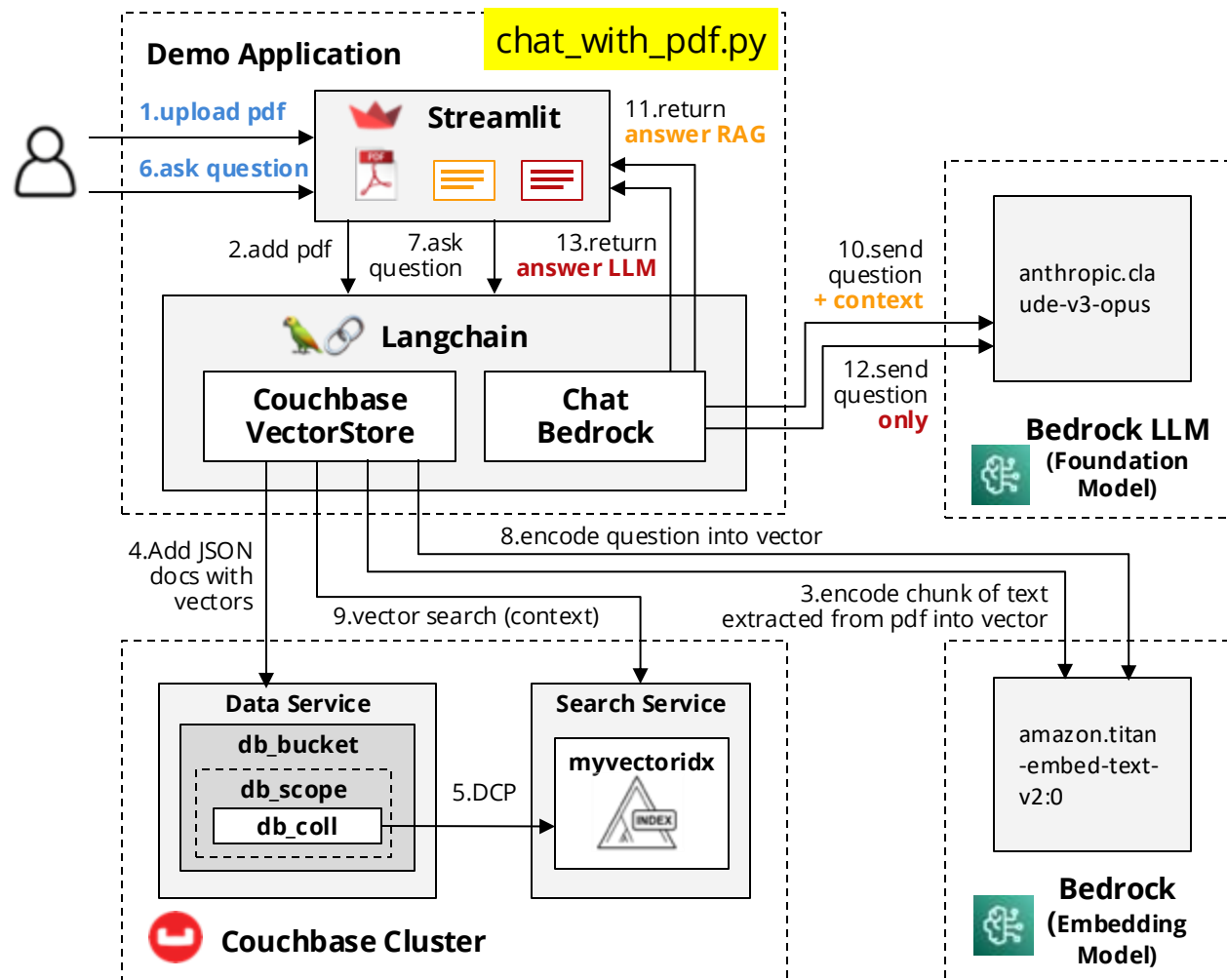
## ➤ RAG Application with PDF (Couchbase + LangChain + Bedrock)

Retrieval-Augmented-Generation Demo  
using Couchbase, Streamlit, Langchain, and OpenAI



- A full web based RAG demo that imports PDFs as your vector corpus.

<https://github.com/couchbase-examples/rag-streamlit-aws-bedrock>





# Retrieval-Augmented Generation (RAG)

<https://github.com/couchbase-examples/rag-streamlit-aws-bedrock>

```
- Create and activate a virtual environment in a new empty demo directory
$ mkdir MYDEMO
$ cd MYDEMO
$ python3 -m venv .venv
$ source .venv/bin/activate
```

```
$ git clone https://github.com/jon-strabala/easy-webrag-langchain-demo.git
$ cd easy-webrag-langchain-demo
```

```
- Install dependencies
$ pip install -r requirements.txt
```

```
- Copy the template environment template
$ cp _setup.tmpl _setup
```

```
$ vi _setup
export CB_BUCKET=vector demos
export CB_SCOPE=langchain
export CB_COLLECTION=webrag
export CB_SEARCHINDEX=webrag_index
```

```
$ source _setup
$ ./check_couchbase.sh
$ ./check_openai.py
$ ./setup.py
```

```
### Run the application this will start a webserver
$ streamlit run chat_with_pdf.py
```

Upload your PDF

Choose a PDF.

Drag and drop file here  
Limit 200MB per file • PDF

Browse files

230602 [보도자료] ...  
1.1MB

Upload & Vectorize

How does it work?

- ☒ Use pure LLM (ChatGPT)
- ☒ Use RAG (vector query against Couchbase)

For RAG, we are using [Langchain](#), [Couchbase Vector Search](#) & [OpenAI](#). We fetch parts of the PDF relevant to the question using Vector search & add it as the context to the LLM. The LLM is instructed to answer based on the context from the Vector Store.

## Chat with PDF

Below you can enter questions and we will process your question twice to showcase running without and with Retrieval Augmented Generation (RAG) while you chat with your PDF(s).

Answers with are generated by pure LLM (ChatGPT) while are generated using RAG (vector query against Couchbase) and then passing that context with the question to the LLM

국가 보조금이란

국가 보조금이란 정부가 경제의 특정 분야나 그룹을 지원하기 위해 제공하는 재정적 지원입니다. 이러한 보조금은 개인, 기업, 기관, 지방자치단체 등에게 다양한 형태로 제공될 수 있으며, 일반적으로 경제 발전을 촉진하고, 공공의 이익을 증진하고, 사회적 불균형을 해소하며, 특정 산업이나 활동을 장려하고자 할 때 사용됩니다.

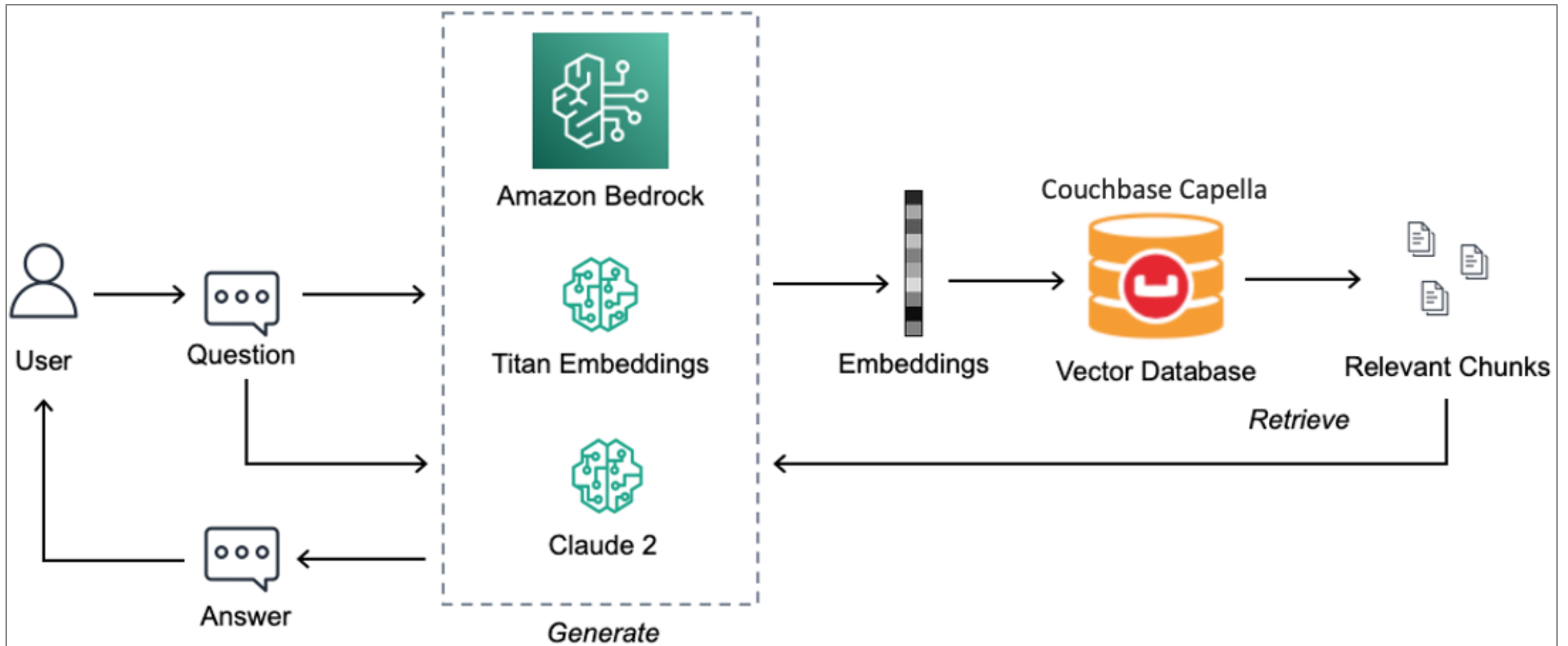
보조금은 직접적인 현금 지급, 세금 감면, 저리의 대출, 보증, 물품이나 서비스의 제공 등 다양한 방식으로 이루어질 수 있습니다. 예를 들어, 재생 가능 에너지 분야에 투자하는 기업에 대한 세금 혜택, 농민들에게 주어지는 농업 보조금, 저소득층 주택 지원을 위한 보조금, 교육 장려를 위한 장학금 등이 있습니다.

국가 보조금은 해당 국가의 경제 정책과 사회복지 정책의 일환으로 운영되며, 정부의 재정 상태, 경제적 우선 순위, 정치적 결정에 따라 크기와 형태가 달라질 수 있습니다. 보조금은 도움이 필요한 분야에 자금을 투입함으로써 경제적 혜택을 제공하고 사회적 안정을 도모할 수 있는 중요한 수단입니다. 그러나 잘못 관리될 경우 비효율성, 경제적 왜곡, 예산 낭비 등의 부정적인 결과를 초래할 수도 있습니다. 따라서 보조금 정책은 신중

Ask a question based on the PDF(s)

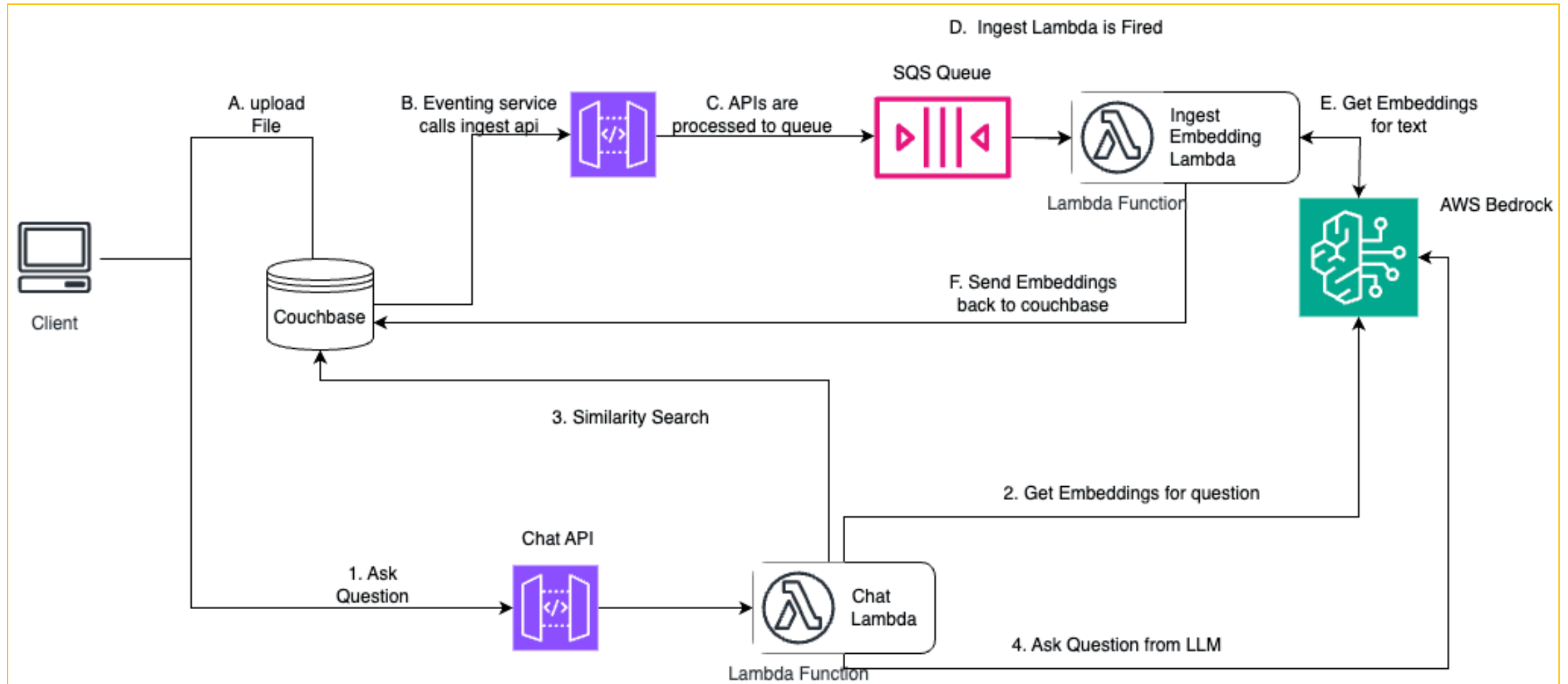
# Travel-sample

[https://github.com/couchbaselabs/RAG\\_Amazon\\_Bedrock\\_example](https://github.com/couchbaselabs/RAG_Amazon_Bedrock_example)



# GenAI(LLM & RAG) on AWS Demo.

<https://developer.couchbase.com/tutorial-bedrock-serverless-pdf-chat>



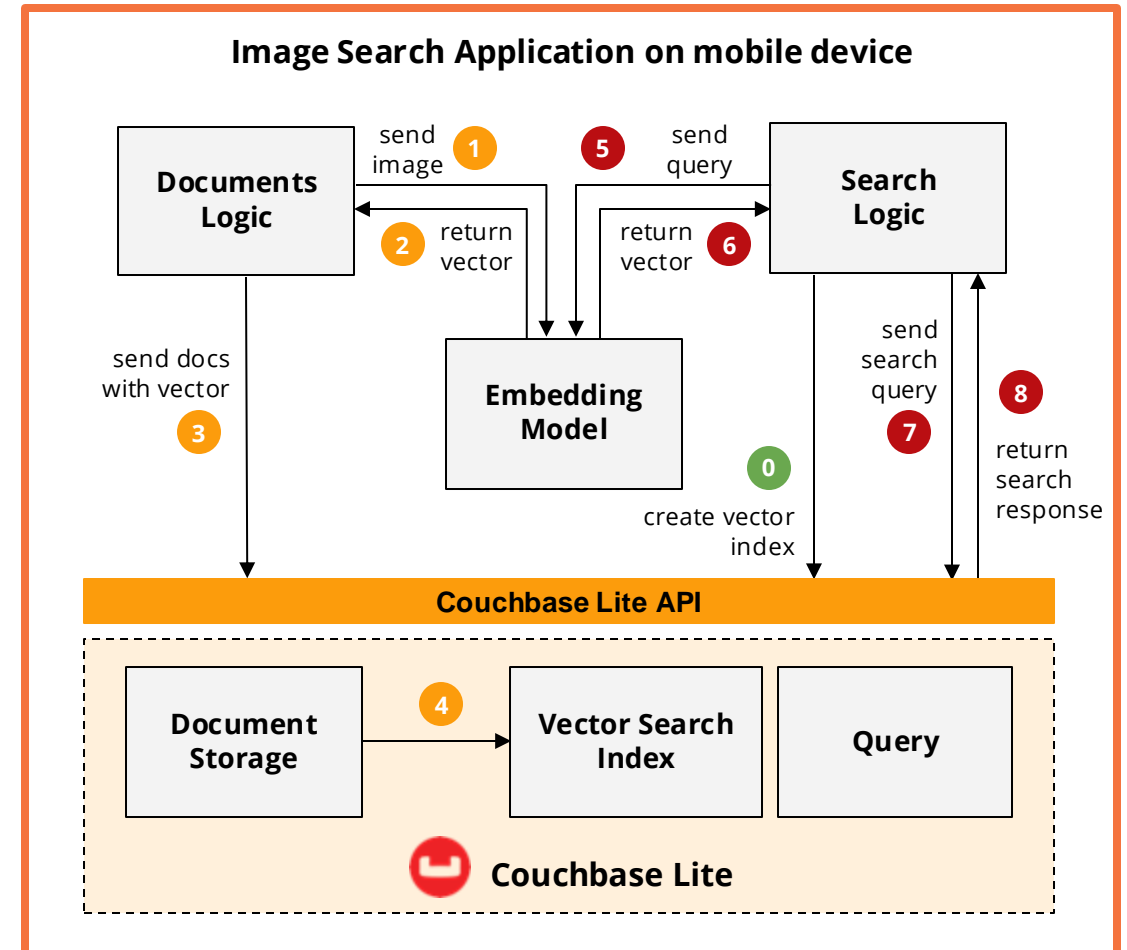
# Appendix. 참고자료



# Couchbase Mobile Vector Search Demo.

## ➤ Image Search with Couchbase Lite

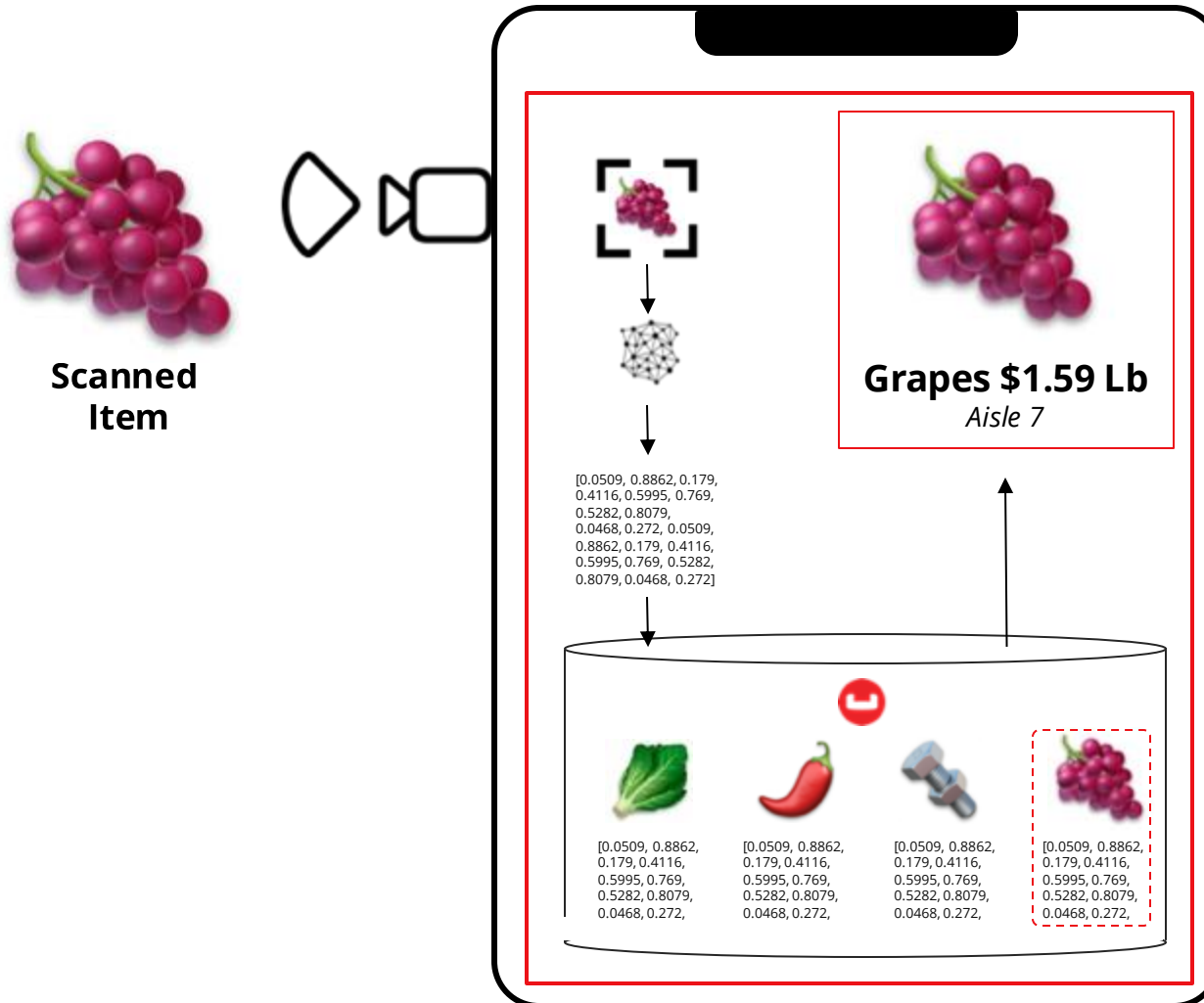
- 0 Application creates vector index
- 1 Application sends documents to embedding model
- 2 Embedding Model returns embedding vector and application adds vectors to JSON documents
- 3 Application creates/updates JSON documents (image with vector) to Couchbase Lite
- 4 This triggers update to vector search index on Couchbase Lite that updates index with document
- 5 Application sends captured image to Embedding Model
- 6 Embedding Model returns corresponding embedding vector
- 7 Application calls Couchbase Lite with a pure vector search, or hybrid mixed vector search and FTS, or hybrid vector and SQL++ query
- 8 Couchbase Lite returns top k-NN results of query with requested fields or objects - the vector search returns similar images





# Couchbase Mobile Vector Search Demo.

## ➤ Vector Search For Image Lookup



### Tablet-based Point-of-Sale app

- Item is scanned with device camera
- Vectorized with local embedding model
- Couchbase Lite vector store is searched
- Nearest vector match found
- App displays matching item and data

## 4. Couchbase Mobile Vector Search Demo.

### ➤ Simple Intelligence App



<https://apps.apple.com/us/app/simple-intelligence/id6504311724>



# 수고하셨습니다.



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