

Vector Databases and Large Language Models

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Sam Partee – Principal Engineer, Applied AI



Large Language Models

What are they and what do they do?



redis

- LLMs are massive, general purpose neural networks, pre-trained on large amounts of text.
- Specifically focused on language understanding and generation (GPT, BERT, LLAMA).
- Commonly utilize **vector similarity search** to retrieve information from external databases
- Use Cases:
 - Translation
 - Sentiment Analysis
 - Content Generation/Summarization
 - Question Answering

Vector Embeddings

What are vector embeddings and how are they created?

Vectors

- Commonly represent unstructured data
 - Audio, text, images, etc
- Usually of high-dimension in the form of a **dense** embedding.
- Packed with information
- Easy to use API to create



Vector Embedding Creation

- Simple creation APIs
- Example with HuggingFace Sentence Transformer

•••

```
1 from sentence_transformers import SentenceTransformer
2 model = SentenceTransformer('sentence-transformers/all-MiniLM-L6-v2')
3
4 sentences = [
5 "That is a very happy Person",
6 "That is a Happy Dog",
7 "Today is a sunny day"
8 ]
9 embeddings = model.encode(sentences)
```



Vector Similarity Search

How are vector embeddings used for similarity search?

- 3 semantic vectors = **Search Space**
 - "today is a sunny day"
 - "that is a very happy person"
 - "that is a very happy dog"
- 1 Semantic vector = **Query**
 - "That is a happy person"





https://mlops.community/vector-similarity-search-from-basics-to-production/

Vector Database

How are vector embeddings used in production?



Redis – Vector Database



redis

- **Redis**: Low-latency, scalable, in-memory database
- Indexing methods
 - HSNW (ANN)
 - Flat (KNN)
- Distance metrics
 - L2, Cosine, internal product
- Support for hybrid queries
 - Vector search + filtering by text, geo, etc.
- Store vectors in JSON (new in 2.6)

Design Patterns

For using Large Language Models with Vector Databases



Vector Database + LLM

Design Patterns



Context Retrieval

- Search for relevant sources of text from the "knowledge base"
- Provide as "context" to LLM



LLM "Memory"

- Persist embedded
 conversation history
- Search for relevant conversation pieces as context for LLM



LLM Cache

- Search for semantically similar LLM prompts (inputs)
- Return cached responses



Context Retrieval

Finding relevant information for LLM queries



Document QnA Example: https://github.com/RedisVentures/redis-openai-qna **Chatbot Example w/ Langchain:** https://github.com/RedisVentures/redis-langchain-chatbot

• Description

- Vector database is used as an external knowledge base for the large language model.
- Queries are used to detect similar information (context) within the knowledge base

Benefits

- Cheaper and faster than fine-tuning
- **Real-time updates** to knowledge base
- Sensitive data doesn't need to be used in model training or fine tuning
- Use Cases
 - Document discovery and analysis
 - Chatbots

Long-Term Memory

Increasing available context to LLMs



Repository: https://github.com/continuum-llms/chatgpt-memory

• Description

- Theoretically infinite, contextual memory that encompasses multiple simultaneous sessions
- Retrieves only last K messages relevant to the current message in the entire history.

Benefits

- Provides solution to context length
 limitations of large language models
- Capable of addressing topic changes in conversation without context overflow

• Use Cases

- Chatbots
- Information retrieval

LLM Caching

Reducing cost and improving QPS of LLMs



Description

- Vector database used to cache similar queries and answers
- Queries embedded and used as a cache lookup prior to LLM invocation

Benefits

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- Saves on computational and monetary cost of calling LLM models.
- Can **speed up applications** (LLMs are slow)
- Use Cases
 - Every single use case we've talked about that uses an LLM.

Architecture

Considerations for LLM + Vector database designs



Providers

Embedding, integration, model providers





- Open Source vs Closed Source
- Platform vs roll your own
- Enterprise readiness
- Cost

- Scale and performance
- Integration providers
 - Ex. Langchain vs Relevance
- Model Providers
 - Ex. Huggingface vs OpenAl or Cohere
 - Infra cost vs model cost? Dev time?
- Embeddings Providers
 - Model/embedding manipulation? Fine-tuning?

Vector Database Performance

Queries per Second (QPS) to Recall



- Performance for ANN benchmarks commonly calculated by QPS/recall
 - Importance of this varies by use case
 - Rec-Sys? Important

- Chatbots? Less Important
- Higher in performance usually leads to having less features
 - PQ for scale, but then maintain cookbook
 - Faiss re-training
 - Hybrid search
 - Database featuress



Example Use cases

For joint Vector Database + LLM architectures





Document Intelligence + retrieval

Discovering insights in documents



How to get started?

Where do I go next?

- Getting Started Resources
 - OpenAl Cookbook: https://github.com/openai/openai-cookbook/tree/main/examples/vector_databases/redis
 - Redis Ventures: https://github.com/redisventures
- Examples
 - LLM Document Chat: https://github.com/RedisVentures/LLM-Document-Chat
 - **OpenAl QnA:** https://github.com/RedisVentures/redis-openai-qna
- More information
 - Follow me @sampartee or add me on linkedin
 - Come to the LLM in Prod part 2 talk June 16th

