

# Lecture 39

... job opportunity ...  
with learning rare ...

"Quality of response" matters

- rare learning

job opportunity



4th positions ✓

~ 24 positions

agenda → winning team

## Vector Stores

- vector store
- cloud storage
- local storage

• earlier data stored in rows & columns  
↳ RDBMS

• vector → Representation

↳ Vector database / vector store

• different from conventional db

• vector store mainly used in recommendations

"vector stores are specialized databases optimized for storing & querying high dimensional vector data"

- efficiently store & retrieve vector



Feature high speed similarity search  
scalability integrator

⇒ each vector database have for 4 main

- ① vector / vector embedding
- ② vector metadata  
(information about data)
- ③ original information/data (→)
- ④ unique id

in each vector

How vector store works

- ① cosine similarity
- ② dot product

in vector  $\vec{u} \cdot \vec{v} = |\vec{u}| |\vec{v}| \cos(\theta)$   
 $\vec{u} = (u_1, u_2, \dots, u_n)$   $\vec{v} = (v_1, v_2, \dots, v_n)$

especially in recommender system

for example 1 vector store 1 million embeddings  
 many algorithms

business point of view

- further research
- horror story (know the vocabulary)  
 • why? why are you doing.



# Lang Chain & Vector Store

provide seamless integration

- cloud cost
  - ① fully managed service / cloud (third party) (api)
  - ② locally managed database (manage itself all by yourself)

Benefits → efficient data storage → resource cost ↓  
 fast retrieval  
 improved semantic understanding

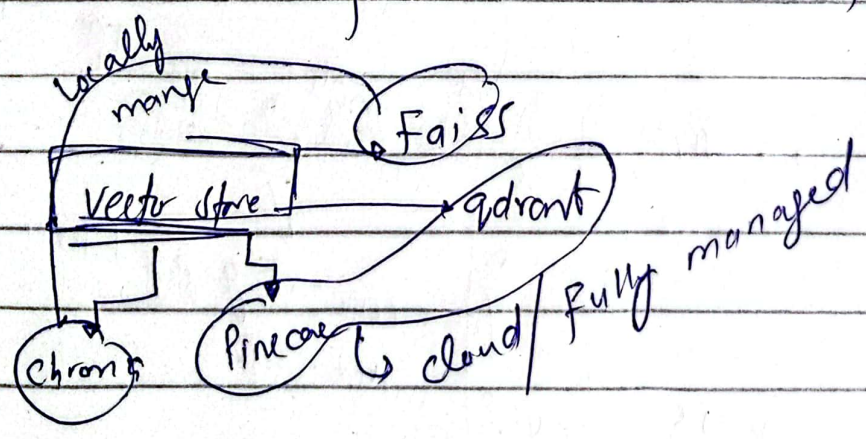
why vector.

Textual data  
 (computer can't understand)

embeddly model → vector form (numbers)  
 (computer can understand by comparig vector value)

• processing can't understand textual data  
 • meaningful information

facebook ai semantic search



## FAISS

↳ Library for effi.

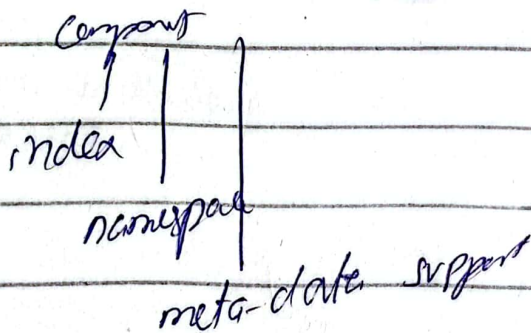


## use cases

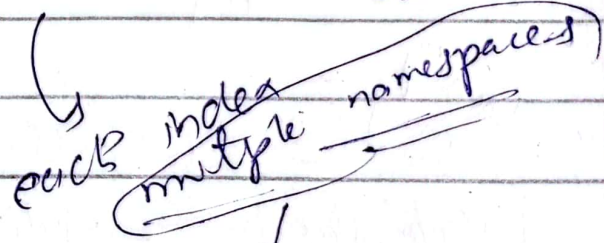
- ① in RAG (semantic search)
- ② → recommender system  
(preferences, queries,
- ③ → anomaly detection

can be multiple use cases depend on critical thinking

## Line core architecture



index ↔ vector store



for concept of partition

the  $\cup$  were no index  $\cup$   $\cap$   
partitions  $\cap$  (namespace)  $\cup$



## Qdrant

cluster — multiple name space  
(vector stores)

- pinecone, qdrant → fully managed.

user → project → vector store

(different name on different platform)

pinecone → index

qdrant → cluster

## Practical

— review of previous 3 steps.

- embeddy model (∴ BAAI/lge-small-en-v1.5)

- ✓ load the model (PyPDF loader)

- ✓ splitting data into chunks

- ✓ meta data pre-processing (edit meta data)

(splitting page into chunks)

• updated  
meta data  
(you define of  
our chunk)



## Manged Service

- Qdrant credential and key collector

automatically manages embeddings.

(directly create)

- Do not spend redundant time

$k \rightarrow$  number of chunks

↓ important question

qdrant similarity-search

query  $\rightarrow$  vector form  $\rightarrow$  vector store

cosine similarity

$\approx$  vector  $\perp$  vs vector  $\uparrow$

$\approx$  (distance) up is  $\downarrow$

- Pinecone

same method

vector similarity-search

Pinecone

serverless

(data store + retrieve) cost

Pools

(store + space + retrieve)

↓  
cost

$\rightarrow$  update your knowledge latest

$\rightarrow$  necessary & critical step