

Navigating the Qdrant Ecosystem from Local to Cloud

Presented to you by Mohamed Arbi Nsibi



Mohamed Arbi Nsibi

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Qdrant is ...

Fully Open-source

Self-hosting : run on your own infra

Super fast: Latency ~0.024s (~24ms)

Hybrid Search

UI support

Free Tier ~1M(vectors) 768-dim

Vector Search in Production

- Written in **Rust** and offers **great performance**
- Allows to interact by **HTTP or gRPC protocols**.
- Runs both in **single and multiple node** setup.
- Incorporates **category, geo-coordinates** and **full-text filters**
- Supports **hybrid, multimodal, multivector** and **multi-staged** search
- Official **Python, Javascript/Typescript, Rust and Go** SDKs.
- Makes vector search **affordable**.

For Managed Cloud solutions, check out
Cloud Embeddings Inference.



Creating a collection/table:

```
PUT /collections/rentals
{
  "vectors": {
    "size": 300,
    "distance": "Cosine"
  }
}
```

SQL equivalent needs a schema

```
CREATE TABLE rentals (
  id INTEGER PRIMARY KEY, vector FLOAT[], city TEXT,
  sqft INTEGER, img_url TEXT, tags TEXT[], description TEXT
);
```

Field indexing:

```
// Vector indexing happens by default
// Each payload index adds more links to keep the graph connected for effective filtering
// Repeat for 'sqft' field with 'integer' type
PUT /collections/rentals/index
{
  "field_name": "city",
  "field_schema": {
    "type": "keyword",
  }
}
```

```
CREATE INDEX rentals_city_sqft_idx ON rentals (city, sqft);
```

Search/Read:

```
POST /collections/rentals/points/search
{
  "query": [0.2, 0.3, ..., 0.4], // generated from user query (text) using same model
  "filter": { "must": [{ "key": "city", "match": {"value": "Bangalore"}}, {"key": "sqft", "range": { "gte": 1000 }}]}, "limit": 10
}
// Response:
[
  {"id": 4, "score": 0.56, "payload": {...}},
  {"id": 2, "score": 0.40, "payload": {...}},
  {"id": 5, "score": 0.23, "payload": {...}}, ]
```

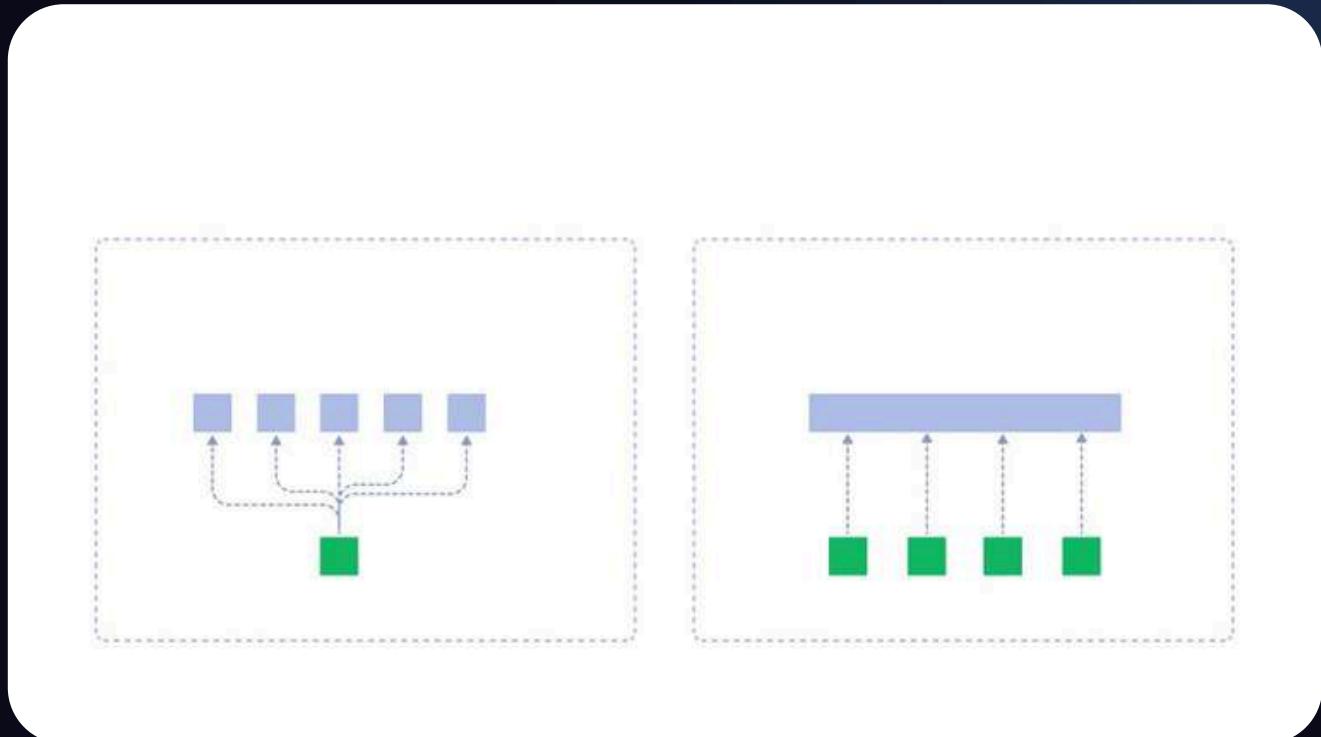
Postgres equivalent (just filtering, no vector search):

```
SELECT * FROM rentals WHERE city = 'Bangalore' AND sqft > 1000 LIMIT 10
```

Latency and Throughput

- Latency: Time taken for a single request
- Throughput: Number of requests handled / second
- Min. latency via

```
num_segments == num_cpu
```
- Max. RPS via fewer but larger segments
But longer indexing time



Vector Search Basics

Dense embedding:
e.g. from *BERT*



Gemini

jina



aws



TwelveLabs

NOMIC

How Qdrant Achieves Search

Core Capabilities

Q Vector Search

Scalable similarity and discovery search (billions of vectors)

Filtering

Numeric, categorical, geo, temporal filters out-of-the-box

Hybrid Search

Combine dense + sparse embeddings, filters, and metadata

Distributed & Resilient

Replication, sharding, multi-tenancy

I Re-ranking

Maximum Marginal Relevance (MMR), score boosting

L Quantization

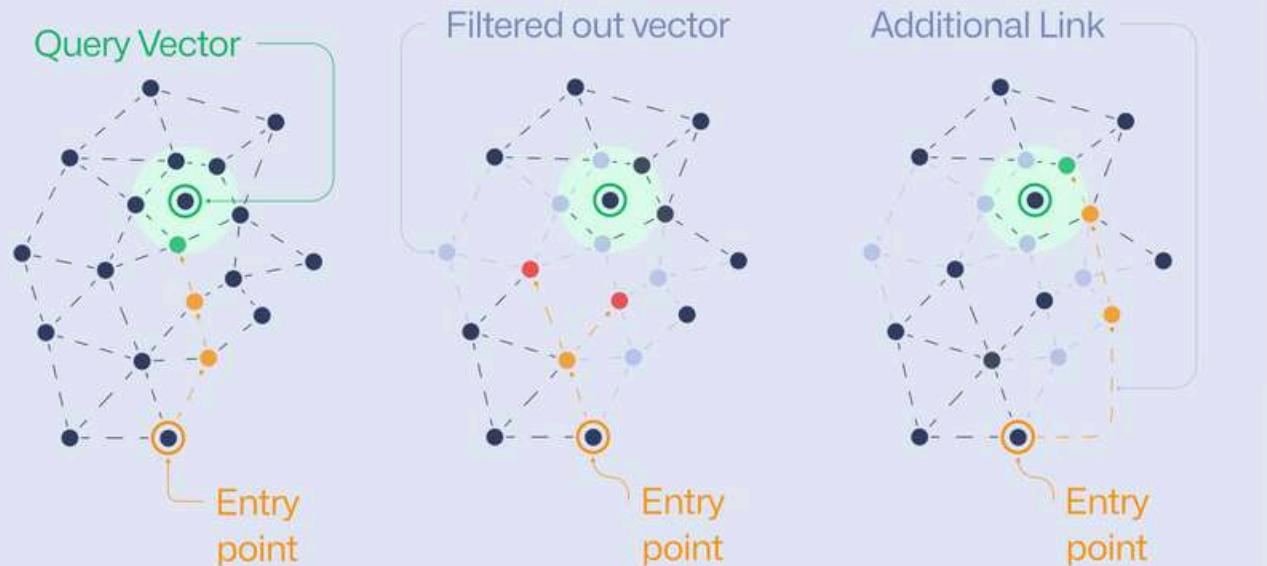
Binary, scalar & product; lower cost without major recall loss

↗ Multi-vectors: Late interaction for retrieval models (e.g. ColBERT)

Performance Optimizations

HNSW tuning, payload indexing, prefetching

Filterable HNSW

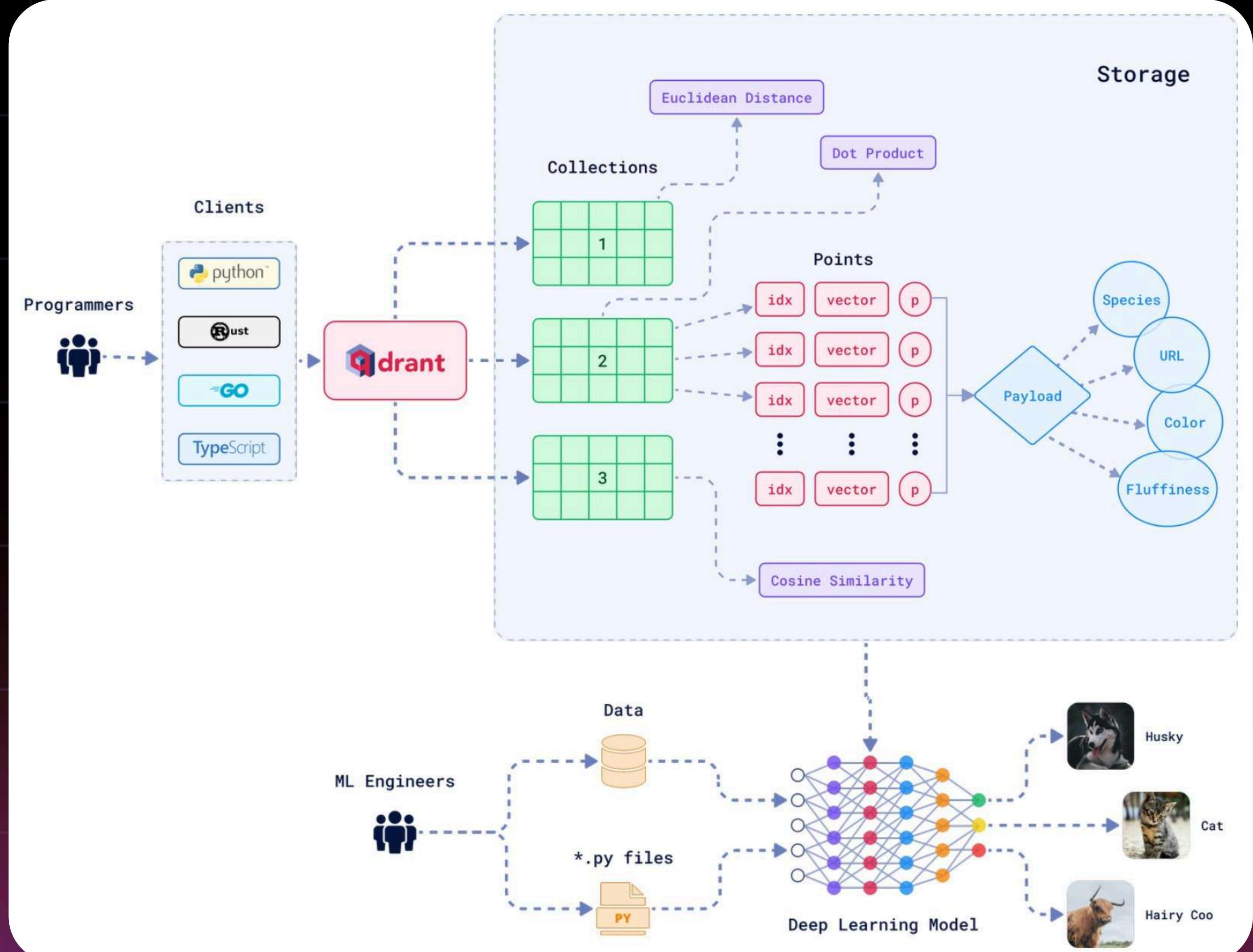


Similarity Search



Similarity Search with MMR





DEMO



Qdrant Innovations

to make development easier

FastEmbed

Generate high-quality embeddings fast. A small Python library for embedding generation, built in and integrated with Qdrant.

- Works out of the box in Qdrant.
- Few dependencies: runs on **CPU**; skips multi-GB **PyTorch** downloads.
- Made for speed: uses **ONNX** Runtime and data parallelism.

Key features

- Use Qdrant models (**miniCOIL**, **BM42**).
- Support for late-interaction (**ColPali**, **ColBERT**) and sparse-neural methods (**SPLADE**, **BM42**, **miniCOIL**), **MUVERA** embeddings and more.
- Run inference and upsert/search in one call.

Import:

```
from qdrant_client.models import  
Document, Image
```

MCP Servers

Build custom retrieval-based AI apps fast. Start from these servers and add tools/commands for your data and workflows.

- **mcp-server-qdrant**: official MCP server for storing and retrieving data in Qdrant.
- **mcp-for-docs**: open-source API reference for AI coding assistants using semantic code retrieval.

Key features

- Automate codebase documentation.
- Personalize your coding assistant to your project's **conventions** and **rules**.
- Do **inline RAG**.
- Speaks **stdio**, **sse**, and **streamable-http** protocols.

Run:

```
docker run mcp-server-qdrant
```

Qdrant Edge

Bring vector search to the edge: an embeddable, high-performance engine that runs directly on mobile and other edge devices.

- Run on **low-CPU** devices.
- Use one API to manage and synchronize data **on-device** and in your **cloud cluster**.
- Fit common on-device cases: **phones** and **laptops**, smart-home/**IoT**, **robotics**.

Key features

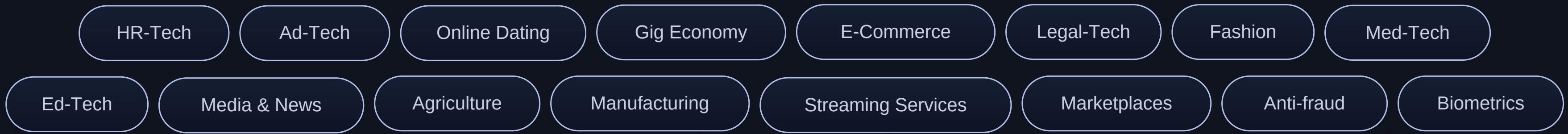
- Use **local storage** to avoid network latency.
- Support **multi-tenant** setups; treat each device as its own tenant.
- Embed as a library; runs in-process with **no** background **daemons**.

Use:

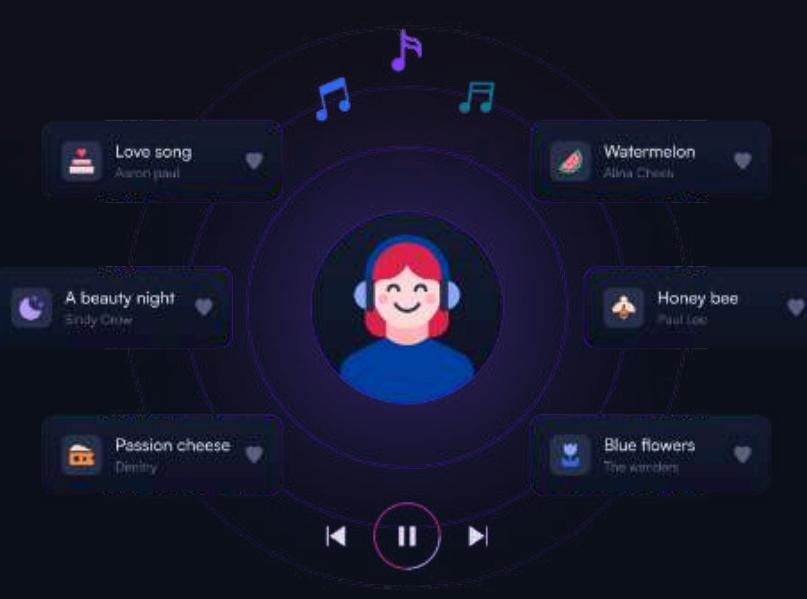
```
client =  
QdrantClient(path="qdrant_edge.db")
```

Vector Search

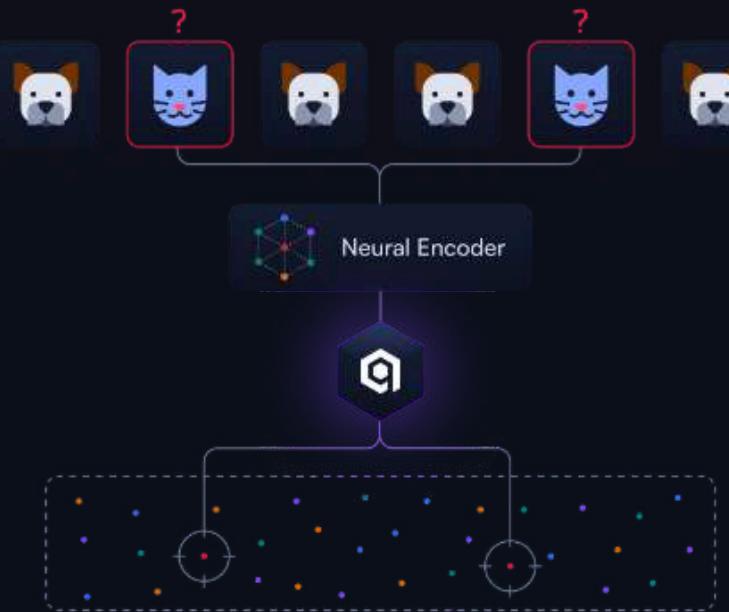
An essential part of the AI Transformation



Search Systems



Recommendations



Anomaly Detection



RAG / Information Assistants

LLMs providers you can start with :

Gemini

deepseek

OpenAI

Hugging Face

cohere

MISTRAL
AI_

Getting Started with Qdrant

Ecosystem

Qdrant Open Source

Usually deployed with Docker containers. Lightweight, offers all the functionalities of Qdrant.

Qdrant Hybrid Cloud

All the benefits of cloud deployment, but keeping the data on your premises. Requires a Kubernetes cluster and might be managed from Qdrant Cloud UI, but no data leaves your environment.

Python SDK Local Mode

Suitable mostly for quick experiments, but not intended to be running in production.



and more...

Trends in DB industry

Vector Search is a very hot paradigm right now.

- Serverless: User shouldn't know/care about the number of machines required. Generally paired with multi-tenancy.
- Neon got acquired.
- Decoupling storage and compute: Put stuff in S3, load only whatever is required.
- Often serverless. Turbopuffer, Tantivy.

Jobs in DB industry

Most DBs are open source. Pick a niche and start contributing

Internships are easier. Best way to enter via GSoC/LFX as a student for good mentorship.

- Some hot DB startups:

Vector DBs: Qdrant, LanceDB, Turbopuffer, Weaviate ,e6data,

Couchbase, Parsable, Yugabyte, Databricks, PureStorage,

Tigerbeetle, Turso, Grafana, Questdb, Supabase

60K



Community Members

26K+

Github Stars

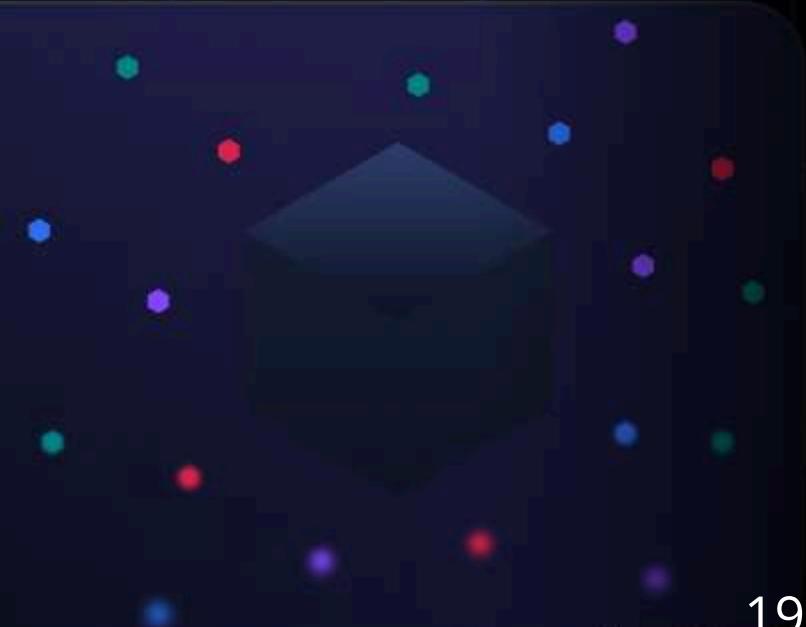


250M+

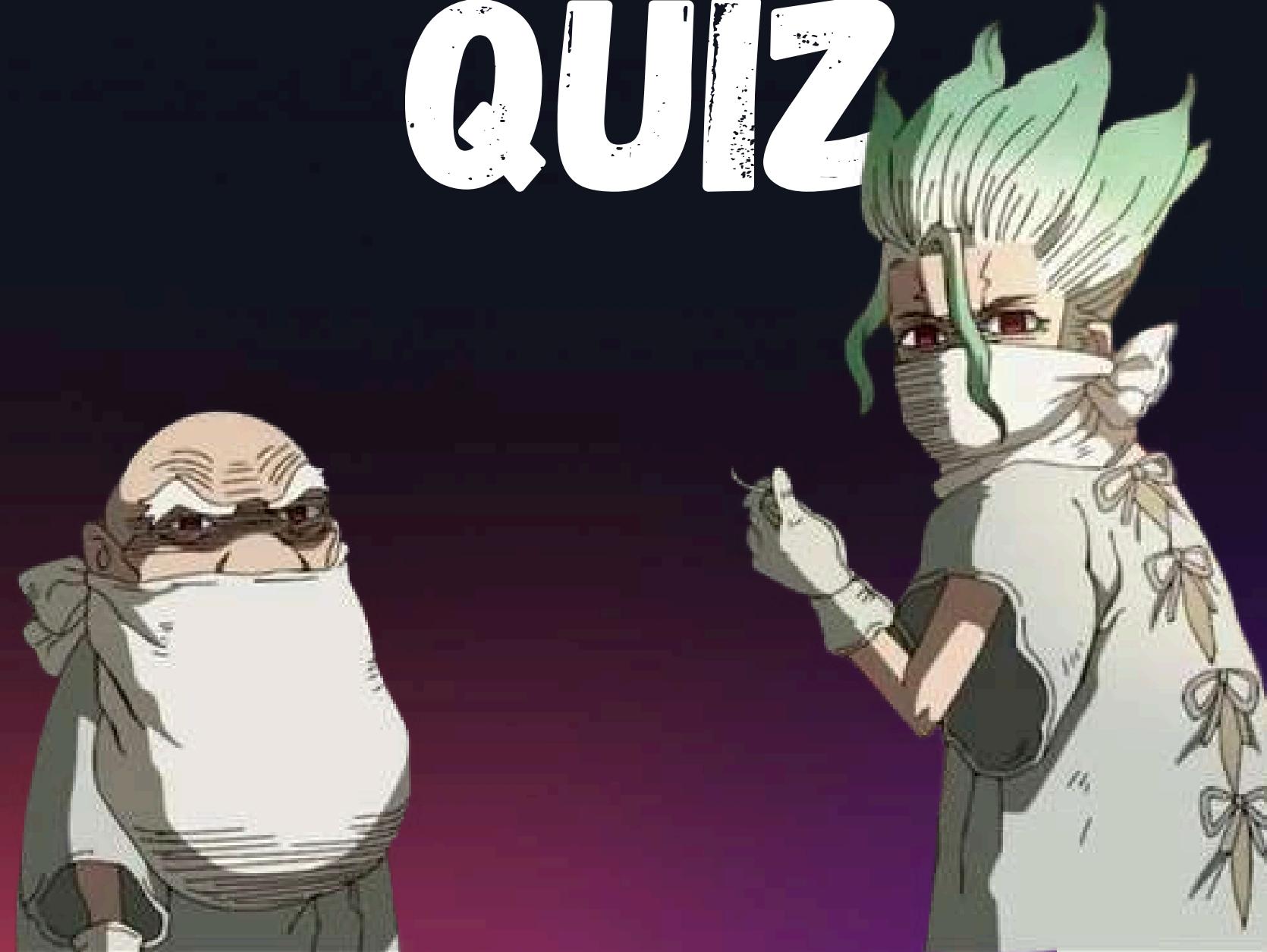
OSS Downloads

>140

Contributors



QUIZ



Resources

- [Qdrant Newsletter subscription](#)
- [Qdrant Homepage](#)
- [Qdrant Documentations](#)
- [Qdrant Cloud signup](#)
- [Just-RAG github repo](#)
- [Qdrant-resources Github Repo](#)
- [Previous Talk materials](#)
- [Qdrant On edge](#)
- [Metric learning for anomaly detection](#)
- [Music recommendation system](#)
- [Filtering with qdrant](#)

Apply for the Private Research Beta

Qdrant Edge is currently in private beta. Due to the highly targeted nature of this release, we will be selecting a limited number of partners who are actively building AI systems for embedded or real-time environments.

If you're working on robotics, edge inference, autonomous systems, or device-native assistants, we encourage you to apply.

[Apply to Join the Qdrant Edge Beta](#)

THANK YOU FOR YOUR ATTENTION!!



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