

YDB: How To Implement Streaming RAG In A Distributed Database

Alexander Zevaykin, **PhD, YDB Team Leader**

Elena Kalinina, **YDB Technical Manager**



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Streaming RAG

Retrieval Augmented Generation



Enterprises Are Not Developing Their Own Generative Al Models



Millions of dollars

A lot of data acquisition

→



Custom model building

- Compute infrastructure

- Scarce GPU GPU
- Expertise
- Data privacy

- Powerful foundation models (with general-purpose reasoning)
- Domain-specific enterprise data via RAG

What can I see in Paris?



LLM

I don't know...



LLM

I don't know...



LLM

I don't know...

Vector

database



Trend #2: Moving Toward Streaming RAG



Al without up-to-date data is frustrating and its value is limited





Generative AI in the enterprise is more about streaming, not batch

020	2022	2023
PT-3	GPT-3.5	GPT-4

Use Cases Of Streaming RAG

Real-Time Financial Advisory Platform

Dynamic Healthcare Monitoring and Assistance

Live News Analysis





What Is YDB?

Distributed SQL RDMS for operational, analytical and streaming workloads



github.com/ydb-platform/ydb



ydb.tech

- Horizontal scaling
- ACID transactions in multiple AZ
- Operability and automatic recovery in case of failures
- Scaling by millions of transactions per second and **petabytes** of data
- Production installations of tens of thousands of servers
- Open-Source under Apache 2.0 license

YDB As A Platform



YDB Intrinsic Advantages



Sharding

Replication

Spiky workload

Cross Datacenter





Herculean tasks



YDB: Real-time Streaming RAG



Streaming Processing

Elena Kalinina,

Technical Project Manager, YDB



YDB Topics is an implementation of persistent queues within YDB

Main features

- Reliability
- High throughput

Based on YDB platform

- Change Data Capture (CDC)
- Transactions with topics and tables

API

YDB Topic API

C++ SDK, Java SDK, Python SDK, Go SDK All YDB Topics features are supported

Apache Kafka API



Change Data Capture



Change feed (topic)

able 1		
Key	Data	
2	abcd	
i		

Changefeeds

for capture any table changes

- Exactly once delivery
- Change records are sharded
- Order of changes

Transfer Data From Topic To Table

CREATE TABLE TargetTable (<Some Columns>);

CREATE TOPIC SourceTopic;

\$transform = {
 < Some Complex Transformation Logic >
};

CREATE TRANSFER ExampleTransfer
FROM SourceTopic TO TargetTable USING \$transform;

Transactions With Tables And Topics

- ACID transactions involving tables and topics
- Within one database



Distributed Transaction Example: Enrich Events



BEGIN TRANSACTION Tx1;

READ 1 EVENT FROM Topic1; $\mathbf{A} =$

READ Data **FROM** Table1 **WHERE** UserID = GetUserID(A); = B

EnrichEvent (A,B); **C** =

WRITE INTO Topic2: EVENT C;

COMMIT Tx1;

- **Table 1** user profiles

Transactions With Tables And Topics

We add ACID guarantees to topic-table operations

CPU usage and system throughput are the same

It simplifies user code

Minimal impact on latency

Topics Autopartitioning

- Topic is divided into the partitions for scalability
- Partitions count can be increased automatically
- Guarantees:
 - Exactly once for writing
 - Reading order



Streaming For RAG in YDB

- Deliver table changes with changefeeds
- Transfer data from topic to table
- Any topic-table data transformations within classic ACID transactions
- Topics autopartitioning

- High throughput
- Reliable
- Kafka API compatible

YDB: Real-time Streaming RAG



Vector Index



Vector Index Requirements

- The index is global
- The index is synchronous and consistent
- Table size = billions
- Search latency = tens of ms

- Creation time = O (table size) •
- Occupied space = O (table size)



Why Don't Existing Algorithms Suit Us?

Distributed system

Automatic scaling

Consistent transactional insertion and searching

SQL Commands

CREATE TABLE table (id Uint32, embedding String, **INDEX** idx vector GLOBAL USING vector kmeans tree ON (embedding) WITH similarity=inner product, vector type=float, vector dimension=1024), PRIMARY KEY (id)

SELECT * FROM table VIEW idx_vector ORDER BY Knn::CosineDistance(embedding, \$target) LIMIT \$k

Vector Index As An Inverted List



Centroids

Search Space Pruning Algorithms

R-tree



Hierarchy Of YDB Vector Index Clusters



level table (parent, id), centroid

posting table (parent, PK), covered indexed table (PK), embedding, covered

Filterable Vector Index

- **SELECT * FROM** table **VIEW** idx vector WHERE user id = \$target user id
- **ORDER BY** Knn::CosineDistance(
- embedding,\$target embedding) LIMIT \$k;

Can't filter

Before vector index

- can't use a single index
- need for full scan

After vector index

need for a repeat request

Need to filter Inside index

Filtrable Vector Index



Vector index 1



Vector index N



- Just one centroid of one cluster would be changed
- The change needs to be pushed up through the levels

Let's Stay In Touch

- How to try YDB?
- Why does it scale so well?
- Why is it so robust?
- What client utilities/ languages are supported?







Conclusion

- YDB is a distributed database
- YDB as a platform offers sophisticated streaming and vector index
- There is a trend for RAG, especially streaming
- We are combining Big Data and AI



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