

OSA CON 25



AI and Analytical Databases

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AI and Databases is a Broad Question

This is a pretty broad topic!

AI is Moving so fast

**There are so many perspectives
based on what your job is**

AI and Data and Databases



Storage of the data for AI Training



In Database Processing, Embedding, Vector Search, ML



Natural Language Interfacing (MCP)



AI for Operations, “AI DBA”



AI for Development



Analyses and Visualizations

Storage of Data for AI Training

If I need to store lots of data for AI Training which Database should I use ?

Thought for 8s

Short answer: don't put raw training data "in a database." Use a **data lake / lakehouse** with cheap object storage for the bytes, and add a catalog + table format for structure. Keep a small OLTP/OLAP store for metadata and experiment tracking, and (optionally) a vector DB for embeddings.

What works best (by scale)

1) Tens of TB → low PB (most teams)

- **Storage (ground truth bytes):** S3/GCS/Azure Blob (or on-prem S3-compatible like **MinIO** or **Ceph**).
- **File/table format:** **Parquet** (tabular), **Arrow** (in-memory), **Apache Iceberg** (table format for versioning, schema evolution, partitioning). (Delta Lake/Hudi are fine too; I prefer Iceberg for engine neutrality.)
- **Catalog/Query:** **Trino** or **Spark** for big scans; **DuckDB** for local/dev; **Polars** for fast single-node work.
- **Compute:** your training stack reads directly from object storage (via fsspec/s3fs, smart caching).
- **Why:** cheapest per GB, scalable, easy to share across Spark/Trino/Presto/DuckDB/Polars/PyTorch.

2) On-prem HPC

- **Storage:** Lustre or BeeGFS for ultra-high throughput; or **Ceph/MinIO** if you want S3 semantics.
- **Same table layer:** Iceberg + Parquet.
- **Cache layer (nice-to-have):** Alluxio to hide object-store latency.

Type of Data Matters

Unstructured Data

Metadata

Structured Data

Specific Formats (i.e., Time Series)

Scale and Type of Training

We are excited to hear about extreme scale but often work with “small” scale

“Small” scale – what you have often works best

Few do full blown hyperscale model training, more do fine-tuning/ domain specific tuning with relatively little data



Training Model From Scratch

From Scratch - Massive, general corpus
– GPT-5 Base Model



Fine Tuning for behavior

Pretrained - Task of Instruction Dataset
- ChatGPT



Domain specific Training

Pretrained – Domain Specific corpus to
specialize in a field – LegalGPT,
CodeLLAMA

Difference

In Database Processing

Vector Datatype and Indexing

In Database Embedding Generation

Semantic Search and Hybrid Search

RAG “Inside Database”

External Model Interfacing

AI Supported Functions (Translate, Summarize etc)

Machine Learning, Anomaly Detection etc



Supporting AI Agents (MCP Servers)



Database Interface can be broad part of your interface to your infrastructure



Management and Maintenance



Natural Language Querying

Natural Language Interfacing

**Inside the
Database
Kernel**

**External “AI
DBA”**

AI for Operations

Inside Database Kernel

Learned Indexes and Cardinality Estimation

AI Based Optimizer

Auto Indexing

Adaptive Algorithms

Autotuning

Contention Prediction

Security

External Solutions

Configuration Tuning

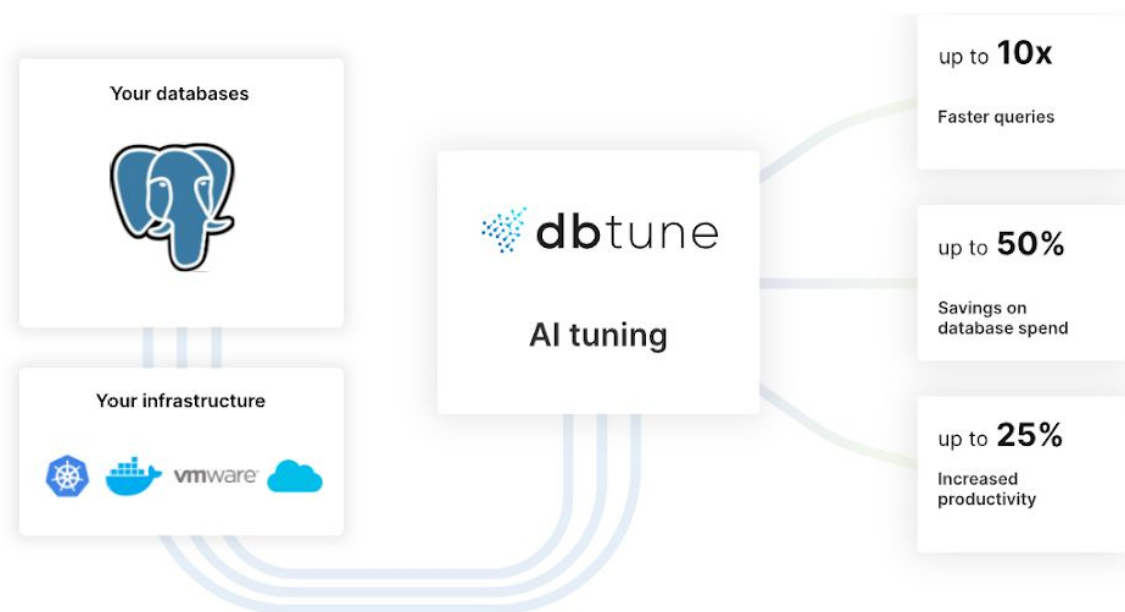
Query Optimization

Autonomous Database (kind of)

Perform better Spend less

DBtune's AI-powered optimizer tunes your PostgreSQL server parameters for top performance, regardless of workload, use case, or machine size.

Focus on strategic tasks, while DBtune saves you time and money.

[Get started](#)[Book a demo](#)

AI powered

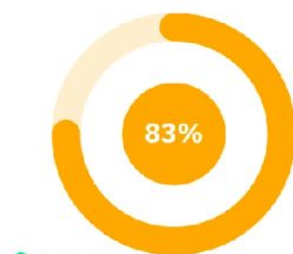
MySQL Performance Monitoring & Tuning

Automate, Optimize, Win

Releem **automatically detects** MySQL performance issues, **tunes configuration** and **optimizes SQL queries**

Start MySQL Optimization Now

Releem Score



Best
Average
Poor

[About Releem Score](#)

Health Status

	System	5/5
	MyISAM / InnoDB	4/5
	Memory	5/5
	Queries / Logs	1/4

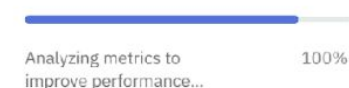
[Health Checks](#)

Security Status

Authentication & Access	3/3
System & Network Security	3/3
Data Integrity & Operations	2/2
 1  2  1	

[Security Checks](#)

Recommended Configuration



10 Unapplied recommendations

[Configuration](#)

Apply



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Tools > Free SQL Optimizer

Free SQL Optimizer for PostgreSQL and MySQL

Optimize SQL queries online for free. Submit your SQL and receive indexing and SQL rewrites recommendations to speed up your PostgreSQL and MySQL queries.

Powered by [Aiven AI Database Optimizer](#).

Database type

PostgreSQL ▾

Version

18 ▾

Please submit the query you would like to optimize:

```
1  -- Demo query
2  SELECT
3    p.id,
4    count(DISTINCT c.id)
5  FROM
6    posts AS p
7  LEFT JOIN
8    comments AS c
9    ON c.PostId = p.id
10 WHERE
11    p.AnswerCount > 3
12    AND p.title LIKE '%optimized%'
```

Next

AI that finds the root cause of your incidents — instantly

Coroot uses eBPF to connect in minutes. No code changes needed. It shows you exactly what broke, why it happened, and how to fix it.

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Root Cause Analysis

The screenshot displays the Coroot 'Incidents' page for incident 'qihh307'. The incident title is 'High CPU Usage from analytics-updater job on node3 caused resource contention affecting catalog and cb-main services'. It started on Jul 28, 15:43:45 (20h ago) and was resolved on Jul 23, 16:32:31, with a duration of 48m. The application is 'front-end' and the root cause analysis is 'done'.

Service Level Objective (SLO)	Objective	Compliance	Error budget burn rate
Availability	99% of requests should not fail	100%	1h: 0.0m 0 threshold: 14
Latency	99.9% of requests should be served faster than 500ms	58.3%	1h: 17.5m 112 threshold: 14

Navigation tabs: OVERVIEW, TRACES

Root Cause

The 'analytics-updater' CronJob running on 'node3' consumed excessive CPU resources, causing CPU delays for both 'catalog' and 'db-main' services. This resource contention led to database connection timeouts, TCP retransmissions, and cascading latency increases throughout the dependency chain from 'db-main' → 'catalog' → 'front-end'.

[Show more details](#)

Immediate Fixes

Monitor and limit resource usage of the 'analytics-updater' CronJob.

```
# Add resource limits to analytics-updater CronJob
resources:
  limits:
    cpu: "500m"
    memory: "512Mi"
  requests:
    cpu: "100m"
    memory: "128Mi"
```

Consider scheduling the job during off-peak hours or on dedicated nodes to prevent resource contention with critical services.

Root cause identified
Recommended fix found

latency, seconds

errors, per second

! Catalog service deployment caused database overload leading to front-end failures and latency spikes

Started: Oct 02, 07:32:41 (11d ago) Resolved: Oct 02, 07:56:41 Duration: 24m Application: front-end Root Cause Analysis: Done ↻ ⓘ

Service Level Objective (SLO)	Objective	Compliance	Error budget burn rate ⓘ
Availability	99.9% of requests should not fail ✎	100%	1h: 0 5m: 0 threshold: 14
Latency	99.9% of requests should be served faster than 500ms ✎	98.94%	6h: 11 15m: 11 threshold: 6

☰ OVERVIEW

📄 TRACES

🔥 Root Cause

The `catalog` service deployment (version `5c66bc476b: catalog:0.50`) introduced performance issues causing high CPU usage on `db-main` database. This led to database connection failures, TCP retransmissions, and context cancellations. The database overload cascaded to `front-end` service causing failed requests (502 errors) and increased latency across all percentiles.

Show more details

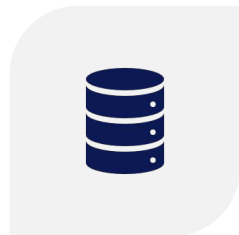
🔧 Immediate Fixes

Rollback the `catalog` service to the previous version:

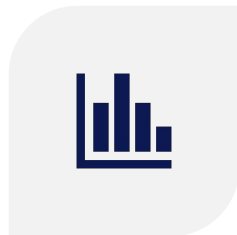
```
kubectl rollout undo deployment/catalog
```



SCHEMA DESIGN



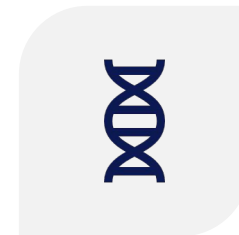
TEXT TO SQL



**QUERY IMPROVEMENT
AND INDEX
RECOMMENDATION**



EXPLAIN EXPLAIN



SYNTHETIC DATA

AI for Database Development

**Text to SQL**

Convert your natural language queries into SQL commands e...

**Explain SQL**

Understand your SQL queries better for clear insights.

**Optimize SQL**

Enhance your SQL query performance.

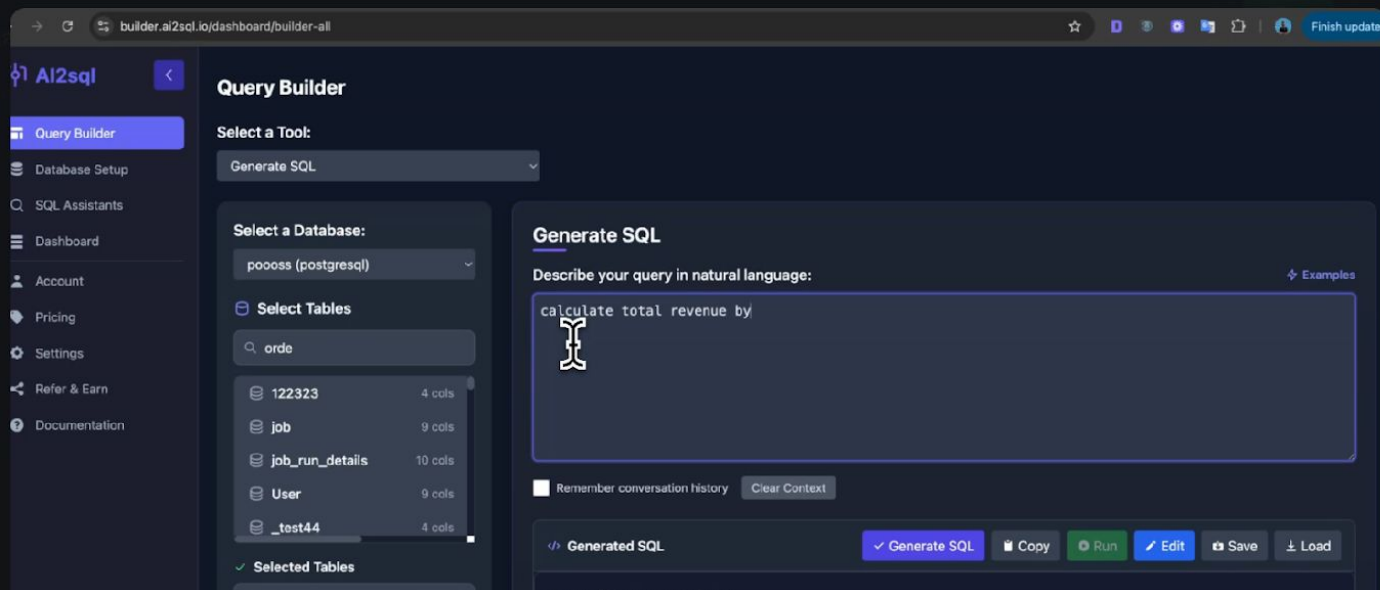
Generate Complex SQL in 10 Seconds — No Coding Required

Instant SQL generation from natural language descriptions



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Data Preparation and
Cleanup



Ask questions in
human language



Visualization, Creating
Dashboards



Explaining Meaning of
Data and Potential
Causes

Analyses and Visualization

DIKW

Data – 100,102,98,105

Information – “Temperatures recorded Hourly in Raleigh, NC Starting at Noon 12th of July 2025”

Knowledge – Temperatures Rose due to Passing Front

Insights – Warm fronts like this often precede storms.
Schedule maintenance early

Wisdom – We will adjust our seasonal maintenance to anticipate similar patterns



**The high pace AI world is “clashing”
with slow moving world of databases**



**A lot of uncertainty how industry will
evolve and how quickly**



**No matter what your job is, track how
AI capabilities are evolving to do it
faster and better**

Final Thoughts

Thank you, Let's Connect!

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