

# Observability for critical applications at Mercado Libre

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Portuguese Localization

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# AGENDA

01

**Observability**

02

The Problem at Scale

03

Our Solution & The New Data Challenge

04

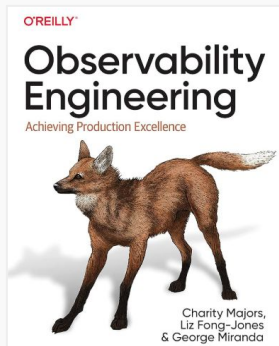
Our ClickHouse Implementation

05

Impact & What's Next



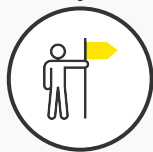
# Observability



Understand the **inner workings** of your application.



Understand **any system state** your application may have gotten itself into, even new ones you have never seen before and couldn't have predicted.



Understand the internal state **without shipping any new custom code to handle it.**



# Telemetry Signals

## Log



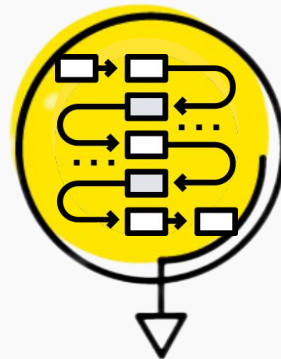
Timestamped **text record**,  
either structured or not.

## Metric



A **measurement** of a service  
captured at runtime.

## Traces



It is a big picture of the **path**  
taken by a **request**.



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# The MELI Ecosystem

To understand our problem, you first need to understand our scale.



**+ 35.000**

MICROSERVICES



**+ 30.000**

DEPLOYS / DAY



**+ 15 MM**

REQUESTS / SEC.



**+ 7.8 B**

SPANS / MIN.





# The "Million-Dollar Question"

Scenario:

- A payment fails.
- A shipping order is lost.

How do you find **one failed request**  
among **billions**?





# The Observability Gap

Our standard tools couldn't answer this.

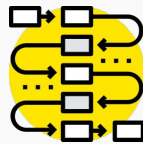
- **Metrics:** Too expensive for high-cardinality debugging.
- **Logs:** Hard to correlate across dozens of services.
- **Traces:** Standard tracing is **sampled**. We can't rely on 'coincidence' to debug a critical failure.



METRICS



LOGS



TRACES



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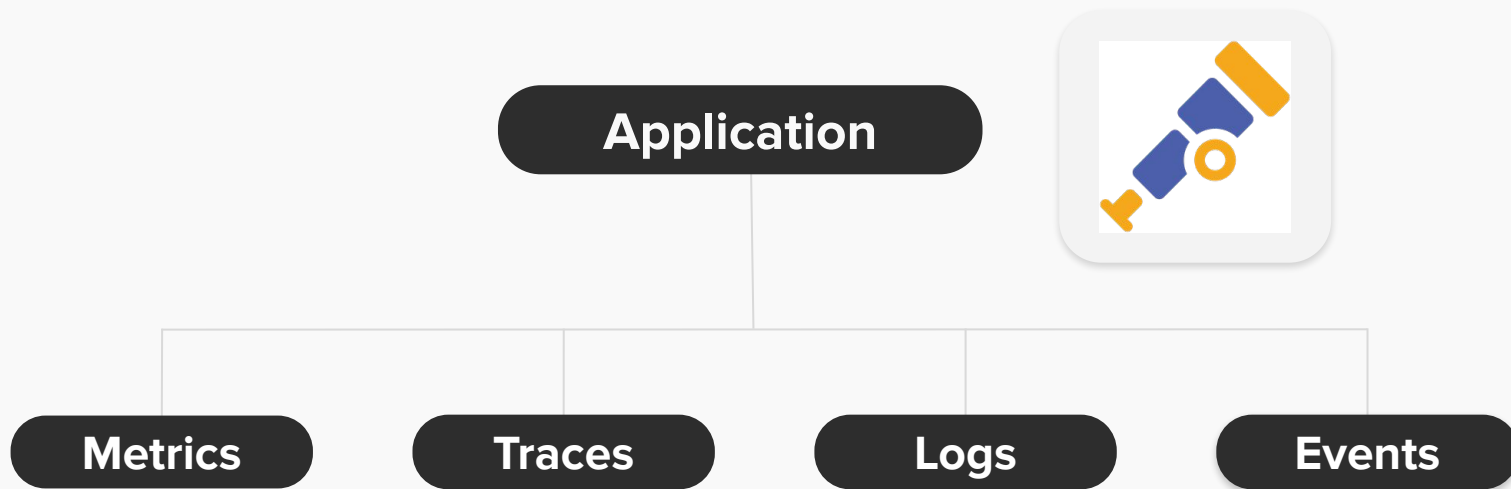
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# Our Solution: Observability Events

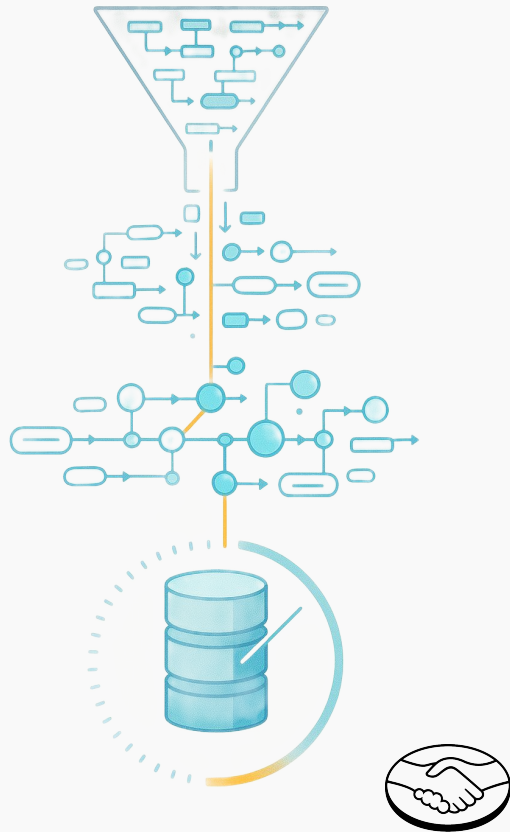
The **Observability Events** arrives as a tool to complement Mercado Libre's observability stack.



# Our Solution: Observability Events

We enable critical business applications to capture 100% of traces.

- ➔ **Full Granularity:** 100% sampling is required for deep, reliable troubleshooting. We can't rely on samples.
- ➔ **Business Context:** We allow traces to be enriched with high-cardinality data, such as **payment.id** or **user.id**.
- ➔ **Long-Term Retention:** Longer period, allowing troubleshooting for older cases and bugs.

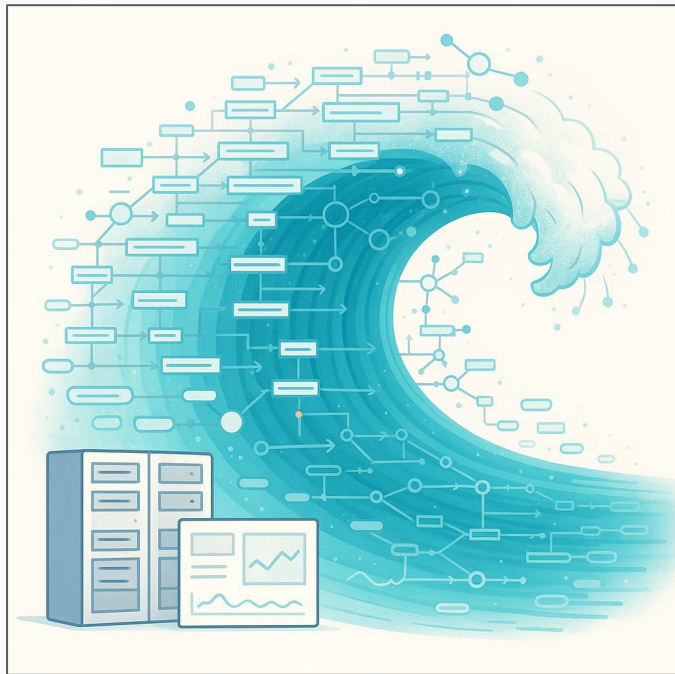


# The "Data Tsunami"

Enabling 100% sampling creates a new, massive problem.

- Critical applications *combined* can generate a peak of **3 billion** spans per minute.
- A *single* large application can generate up to **300 million** spans per minute by itself.

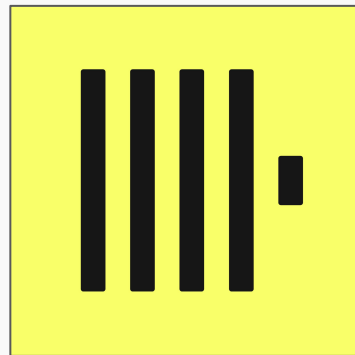
How do you store this affordably and query it instantly?



# Why ClickHouse?

We evaluated several vendors for this extreme data challenge. ClickHouse stood out for its exceptional combination of advantages:

- **Performance:** Delivers outstanding speed for both high-volume data ingestion and large-scale analytics.
- **Cost Efficiency:** Advanced compression and storage optimization significantly reduce overall costs.
- **Open Source & Maturity:** A proven, open-source solution backed by strong documentation and an active, experienced community.



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**Our ClickHouse Implementation**

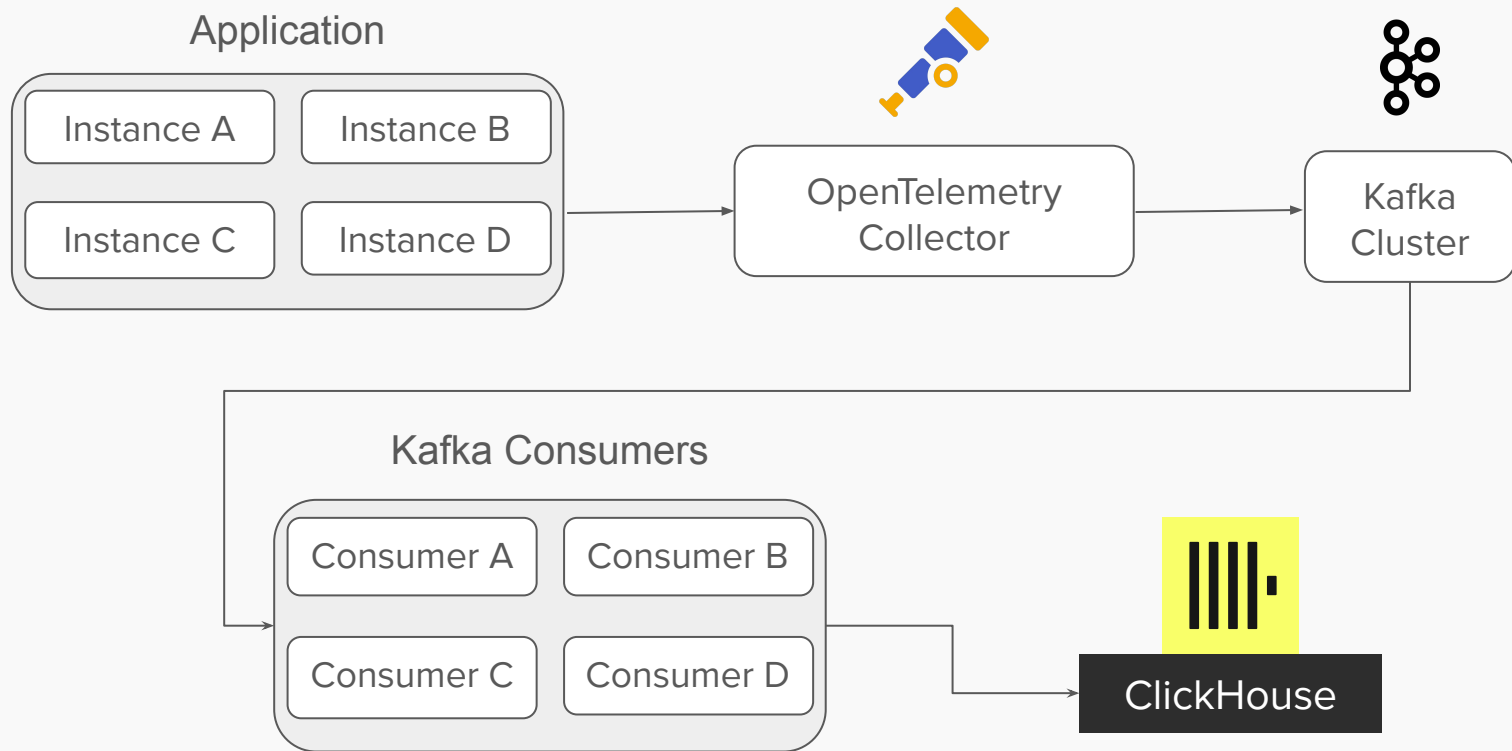
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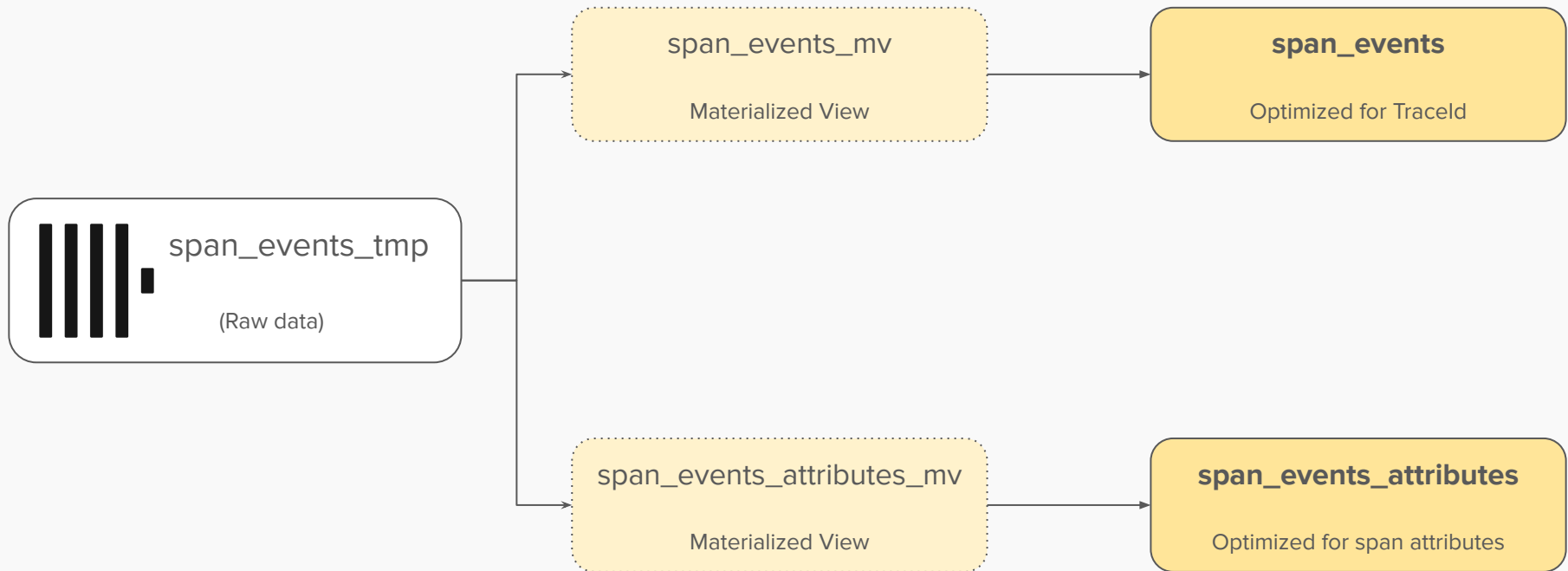




# High-Level Architecture



# ClickHouse Ingestion & Data Flow



# Schema Deep Dive: The "span\_events" Table

Optimized for whole-trace lookups.

- **Engine:** SharedMergeTree
- **Partition Key:**  
toDate(TimestampSeconds)
- **TTL:** 30 days
- **Key Optimization (ORDER BY):**  
TraceId, TimestampSeconds

```
1 CREATE TABLE default.span_events
2 (
3     `Timestamp` DateTime64(9) CODEC(Delta(8), ZSTD(1)),
4     `TimestampSeconds` DateTime CODEC(Delta(4), ZSTD(1)),
5     `TraceId` String CODEC(ZSTD(1)),
6     `SpanId` String CODEC(ZSTD(1)),
7     `SpanName` String CODEC(ZSTD(1)),
8
9     -- [...] OpenTelemetry schema
10
11     `ResourceAttributes` Map(LowCardinality(String), String) CODEC(ZSTD(1)),
12     `SpanAttributes` Map(LowCardinality(String), String) CODEC(ZSTD(1))
13 )
14 ENGINE = SharedMergeTree()
15 PARTITION BY toDate(TimestampSeconds)
16 ORDER BY (TraceId, TimestampSeconds)
17 TTL TimestampSeconds + toIntervalDay(30)
18 SETTINGS index_granularity = 8192,
19         ttl_only_drop_parts = 1,
20         parts_to_delay_insert = 1200,
21         parts_to_throw_insert = 5000;
```

# Schema Deep Dive: The "span\_events" Table

Optimized for whole-trace lookups.

## → Key Optimization (ORDER BY):

- ◆ TraceId, TimestampSeconds
- ◆ This is our most important optimization for traces. By ordering by **TraceId**, all spans for a single trace are physically stored together on disk.
- ◆ This makes lookups for a full trace **extremely fast**.



```
1 ENGINE = SharedMergeTree()  
2 PARTITION BY toDate(TimestampSeconds)  
3 ORDER BY (TraceId, TimestampSeconds)  
4 TTL TimestampSeconds + toIntervalDay(30)  
5 SETTINGS index_granularity = 8192,  
6         ttl_only_drop_parts = 1,  
7         parts_to_delay_insert = 1200,  
8         parts_to_throw_insert = 5000;
```



# Schema Deep Dive: The "span\_events\_attributes" Table

Our "Inverted Index" for fast attribute search.

```
1 CREATE TABLE default.span_events_attributes
2 (
3     `TimestampHours` DateTime CODEC(Delta(4), ZSTD(1)),
4     `TimestampMinutes` DateTime CODEC(Delta(4), ZSTD(1)),
5     `Timestamp` DateTime64(9) CODEC(Delta(8), ZSTD(1)),
6     `TraceId` String CODEC(ZSTD(1)),
7     `AttributeKey` LowCardinality(String) CODEC(ZSTD(1)),
8     `AttributeValue` String CODEC(ZSTD(1)),
9     INDEX idx_attr_value AttributeValue TYPE bloom_filter(0.001) GRANULARITY 1
10 )
11 ENGINE = SharedReplacingMergeTree()
12 PARTITION BY toDate(TimestampHours)
13 ORDER BY (AttributeKey, TimestampHours, TimestampMinutes, AttributeValue, TraceId)
14 TTL TimestampMinutes + toIntervalDay(30)
15 SETTINGS index_granularity = 8192,
16     ttl_only_drop_parts = 1,
17     parts_to_delay_insert = 1200,
18     parts_to_throw_insert = 5000;
```



# Schema Deep Dive: The "span\_events\_attributes" Table

Our "Inverted Index" for fast attribute search.

## Why a new table?

The **OpenTelemetry** schema wouldn't allow us to run high-speed searches over high-cardinality attributes inside the span attributes column (e.g., finding a specific payment.id).

## The solution!

We use a Materialized View to **ARRAY JOIN** and explode the **Map** attributes into simple key-value rows.

```
1 CREATE TABLE default.span_events_attributes
2 (
3     -- [ ... ] Table schema
4
5     INDEX idx_attr_value AttributeValue
6     TYPE bloom_filter(0.001)
7     GRANULARITY 1
8 )
9 ENGINE = SharedReplacingMergeTree()
10 PARTITION BY toDate(TimestampHours)
11 ORDER BY (
12     AttributeKey,
13     TimestampHours,
14     TimestampMinutes,
15     AttributeValue,
16     TraceId
17 );
18 TTL TimestampMinutes + toIntervalDay(30)
19 SETTINGS index_granularity = 8192,
20     ttl_only_drop_parts = 1,
21     parts_to_delay_insert = 1200,
22     parts_to_throw_insert = 5000;
```

# Schema Deep Dive: The "span\_events\_attributes" Table

Our "Inverted Index" for fast attribute search.

→ **Engine:** ReplacingMergeTree

→ **Index:** AttributeValue, TYPE

bloom\_filter(0.001)

The bloom filter quickly skips data blocks where a value doesn't exist.

```
1 CREATE TABLE default.span_events_attributes
2 (
3     -- [ ... ] Table schema
4
5     INDEX idx_attr_value AttributeValue
6     TYPE bloom_filter(0.001)
7     GRANULARITY 1
8 )
9 ENGINE = SharedReplacingMergeTree()
10 PARTITION BY toDate(TimestampHours)
11 ORDER BY (
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```




# Schema Deep Dive: The "span\_events\_attributes" Table

Our "Inverted Index" for fast attribute search.

## Why this specific order?

- 1. AttributeKey:** All our searches start with an attribute key (e.g., payment.id). This is the primary filter and massively prunes the search space.
- 2. TimestampHours, TimestampMinutes:** Allows ClickHouse to skip massive chunks of data (granules) outside the query's time range.
- 3. AttributeValue:** Finally, it filters by the specific value.



```
1  ORDER BY (  
2      AttributeKey,  
3      TimestampHours,  
4      TimestampMinutes,  
5      AttributeValue,  
6      TraceId  
7  );
```

# Schema Deep Dive: Data Types & Codecs

## → LowCardinality

We use this *everywhere* possible: **SpanKind**, **SpanStatus**, **FuryApplication**.

This is critical for storage and query speed.

## → CODEC(Delta, ZSTD)

We use **Delta** for all **Timestamp** columns and **ZSTD(1)** for everything else. For distributed trace data, **ZSTD** is the better choice.



# Schema Deep Dive: Data Types & Codecs

## → Ingestion Tuning (SETTINGS)

### ◆ **ttl\_only\_drop\_parts = 1**

Optimizes our TTL. Instead of deleting row-by-row, ClickHouse drops the entire part once all rows in it are expired.

### ◆ **parts\_to\_delay\_insert = 1200** Helps the cluster handle high-velocity writes by slowing down ingestion slightly if merges fall behind, preventing "Too Many Parts" errors.

### ◆ **parts\_to\_throw\_insert = 5000**

This is the hard limit before the server rejects new inserts. Useful to handle sudden, unexpected traffic spikes.



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# ClickHouse at Scale: Our Ingestion Performance

While still onboarding apps, this schema currently allows us to handle:

- ~ 90 TB of new, uncompressed data per day
- ~ 400B rows written daily to **span\_events\_tmp**
- ~ 1.15T total rows written daily  
*(including all Materialized Views)*



# Demo: Finding the Needle in the Haystack

## Use case

A payment fails. We need to find that trace by the given **user.id**.

1. In our Grafana dashboard, the user selects the filters to **user.id**.
2. In ClickHouse, we use our **span\_events\_attributes** table to find the matching **TraceId**.
3. It then uses the **span\_events** table (ordered by **TraceId**) to retrieve all spans for that trace.



# Demo: Finding the Needle in the Haystack

The query for `span_events`




```
1  SELECT
2    TraceId,
3    toUnixTimestamp(min(StartTime)) as startTime,
4    argMinMerge(RequestId) as requestId,
5    argMinMerge(ServiceName) as serviceName,
6    argMinMerge(Operation) as operation,
7    countMerge(ErrorCount) as errors,
8    countMerge(SpanCount) as total
9  FROM
10   span_events_trace_summary_by_trace
11 WHERE
12   (TraceId, toStartOfMinute(StartTime)) IN (
13     -- {{ span_events_attributes_query }}
14   )
15 GROUP BY
16   TraceId
17 ORDER BY startTime DESC, TraceId
18 LIMIT 25;
```





# Demo: Finding the Needle in the Haystack

The query for `span_events_attributes`



```
1  SELECT
2    TraceId,
3    TimestampMinutes
4  FROM default.span_events_attributes
5  WHERE
6    TimestampHours ≥ toStartOfHour(toDateTime(1762261200))
7    AND TimestampHours ≤ toStartOfHour(toDateTime(1762271999))
8    AND TimestampMinutes ≥ toStartOfMinute(toDateTime(1762261200))
9    AND TimestampMinutes ≤ toStartOfMinute(toDateTime(1762271999))
10   AND (AttributeKey = 'user.id' AND AttributeValue = '156851284')
11  GROUP BY TraceId, TimestampMinutes
12  ORDER BY TimestampMinutes DESC, TraceId
13  LIMIT 25
```



# Demo: Finding the Needle in the Haystack

For this query:

→ **73.723** rows were scanned...

→ ... and ~ **6 MB** of data in ~ **2 seconds**.

Q Search results... Elapsed: 2.096s Read: 73,723 rows (6.03 MB)				
#	TraceId	startTime	serviceName	operation
1	00000000000000005aeb121aff...	1762266117	production-writer-v2.trans...	/v1/transaction-intents/pr...
2	0000000000000000b79b56bfe2...	1762265977	internal-write-kvs.openpla...	POST /payments
3	0000000000000000165d5e20b6...	1762263746	internal-update.openplatfo...	PUT /payments/{paymentId:[...



# Accomplishments & Future Roadmap

Provide a reliable observability ecosystem for core business flows.

## Accomplishments

- **Successfully migrated** our high-volume, 100%-sampling tracing tool to ClickHouse.
- **Achieved** sub-second query latency for high-cardinality trace lookups.
- **Onboarding** our most critical business flows to the platform.



# Accomplishments & Future Roadmap

Provide a reliable observability ecosystem for core business flows.

## Future Roadmap

- ➔ **Advanced Analytics:** Move beyond troubleshooting to perform large-scale analytics, anomaly detection, and business performance monitoring our trace data.
- ➔ **Expand Data-Driven Insights:** Use this rich dataset to build new tools and provide deeper insights into our core business processes.
- ➔ ... Inverted Index? 🙄



# Obrigado!



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