Five Things Every New ClickHouse® User Should Know

Part 1: Applications

Robert Hodges - Altinity CEO 22 July 2025



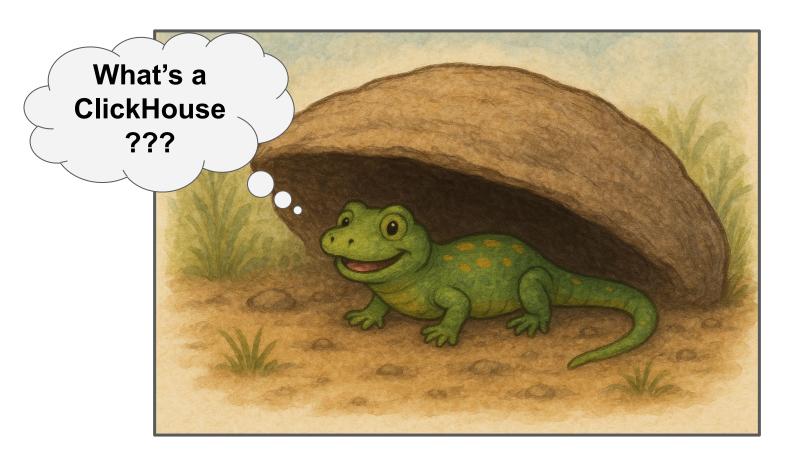


Run Open Source ClickHouse® Better

Altinity.Cloud Enterprise Support

Altinity® is a Registered Trademark of Altinity, Inc.
ClickHouse® is a registered trademark of ClickHouse, Inc.;
Altinity is not affiliated with or associated with ClickHouse, Inc.







ClickHouse® is a real-time analytic database

Understands SQL

Runs on bare metal to cloud

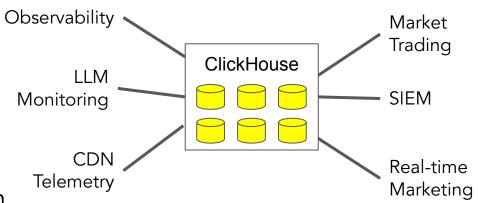
Shared nothing architecture

Stores data in columns

Parallel and vectorized execution

Scales to many petabytes

Is Open source (Apache 2.0)



41.8k GitHub Watchers Can't Be Wrong!



Lesson #1

ClickHouse runs anywhere. Find the one that works for you.



ClickHouse installation is great on Linux! (Including WSL2)

curl https://clickhouse.com | sh



☐ Hard to configure server



For a more conventional install use apt or rpm

Ubuntu example
sudo apt-get install -y clickhouse-server clickhouse-client
sudo systemctl start clickhouse-server

ClickHouse Official Builds

Great for early adopters

Monthly + 2 LTS releases per year

1 year of support

https://clickhouse.com/install

Altinity Stable Builds

Great for enterprises seeking stability
Based on upstream LTS releases
3 years of support, feature backports
https://builds.altinity.cloud



Use Docker on MacOS or for clusters (docker compose)

ClickHouse Official Builds

Tag: clickhouse/clickhouse-server

Altinity Stable Builds

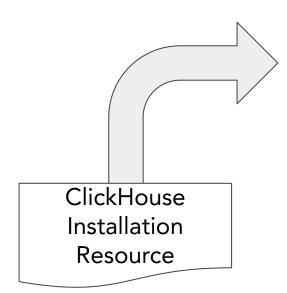
Tag: altinity/clickhouse-server

```
mkdir $HOME/clickhouse-data

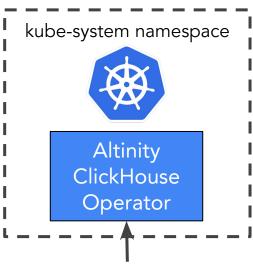
docker run -d --name altinity-stable-build \
    --ulimit nofile=262144:262144 \
    --network=host \
    --volume=$HOME/clickhouse-data:/var/lib/clickhouse \
    altinity/clickhouse-server:24.8.14.10501.altinitystable
```



Run ClickHouse on Kubernetes to build clusters quickly

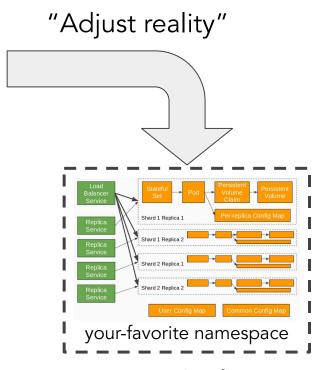


kubectl -f apply demo.yaml



Apache 2.0 source, distributed as Docker image

© 2025 Altinity, Inc.



Best practice deployment .



Or you could skip the setup and just run in a cloud

ClickHouse Cloud

https://clickhouse.com

Altinity.Cloud

https://altinity.com

SaaS version of ClickHouse with Snowflake-like convenience and built-in tools. Cloud platform with SaaS and BYOC models. Runs any version of open source ClickHouse.



Lesson #2

Pay attention to partitioning, sorting, and compression.

They make column storage work.



ClickHouse tables are built for fast query

```
Columns default to
CREATE TABLE default.ontime ref(
                                                      LZ4 compression
     `Year` UInt16,
     `Quarter` UInt8,
     `Month` UInt8,
     `FlightDate` Date,
                                                      Standard engine for
     `Carrier` FixedString(2),
                                                      fast analytics
                                                      How to break table
ENGINE = MergeTree
                                                      into parts
PARTITION BY Year
                                                      How to sort rows
ORDER BY (Carrier, FlightDate)
                                                      within parts
```



Finding airlines with the most cancellations in a year

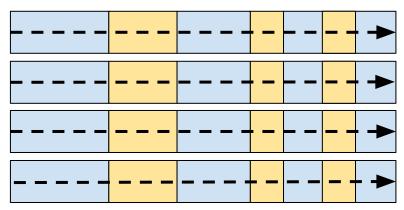
10 rows in set. Elapsed: 0.674 sec. Processed 196.51 million rows, 982.57 MB (291.68 million rows/s., 1.46 GB/s.)10 rows



ClickHouse stores table data in compressed columns

PostgreSQL, MySQL

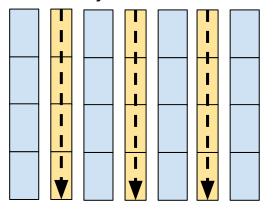
Read all columns in row



Rows minimally or not compressed

ClickHouse

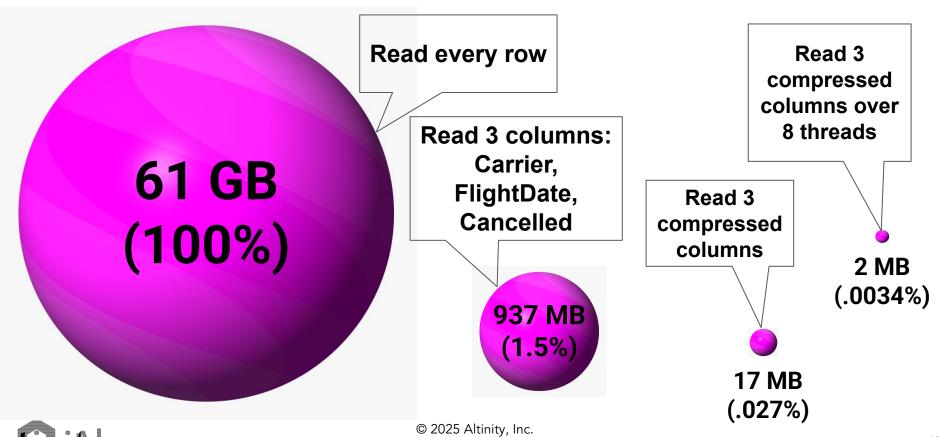
Read only selected columns



Columns highly compressed

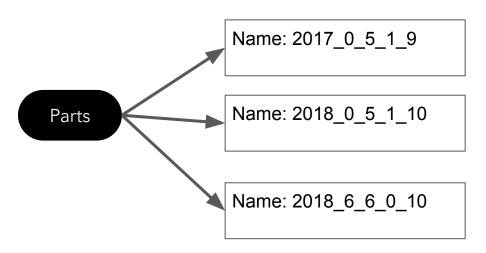


Visualizing effect of columns and compression on I/O



Best practice: partition by time

```
CREATE TABLE default.ontime_ref( . . .)
ENGINE = MergeTree
PARTITION BY Year ORDER BY (Carrier, FlightDate)
```

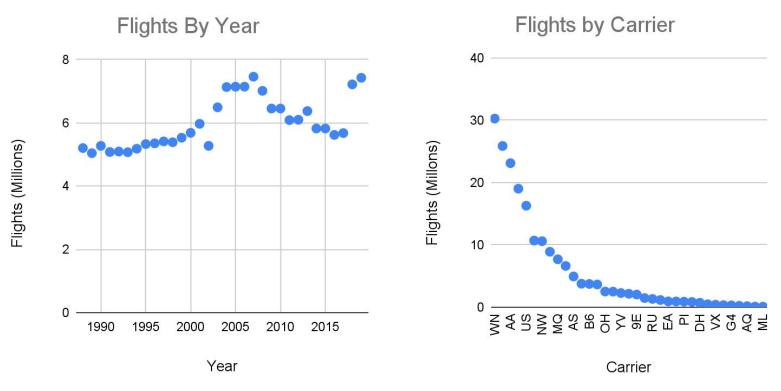


Rule of thumb:

Choose partitions that result in ~1000 parts or less



Why it's better to partition by time

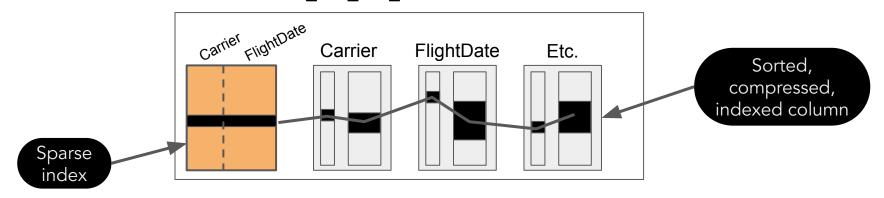




Order by increasing cardinality, with tenant first

```
CREATE TABLE default.ontime_ref( . . .)
ENGINE = MergeTree
PARTITION BY Year ORDER BY (Carrier, FlightDate)
```

Name: 201905_510_815_3





Compress to taste, any time you want to

```
CREATE TABLE default.ontime_ref( . . .)
ENGINE = MergeTree

PARTITION BY Year ORDER BY (Carrier, FlightDate)

TTL FlightDate + INTERVAL 6 MONTH RECOMPRESS CODEC (ZSTD(1)),
    FlightDate + INTERVAL 12 MONTH RECOMPRESS CODEC (ZSTD(10))
```

Automatically increase compression over time



Figure out compression with amazing system tables!

Other great tables: system.parts and system.tables



Lesson #3

ClickHouse likes big parts!

Help out by making inserts as big as possible.

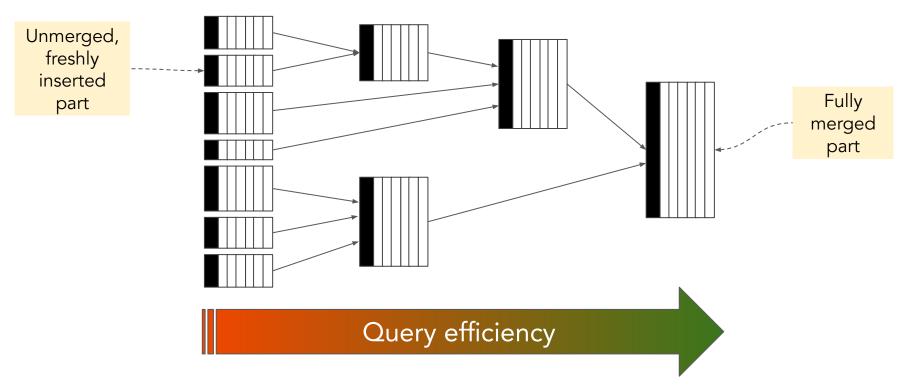


Small inserts can crush your ClickHouse server

ClickHouse Small inserts from many MergeTree clients mytable Storage



Lots of small parts == slow queries and high merge load

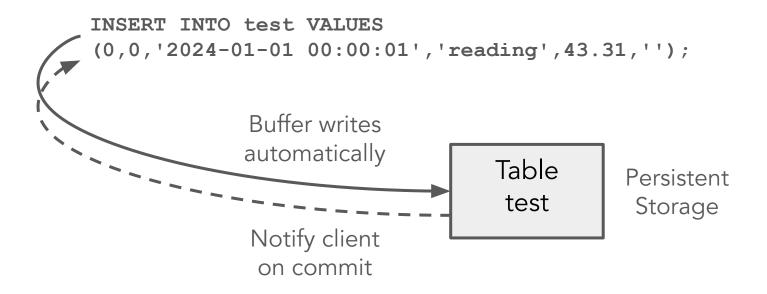




Fix #1: Use big batches in your application



Fix #2: Enable async inserts



https://kb.altinity.com/altinity-kb-queries-and-syntax/async-inserts/

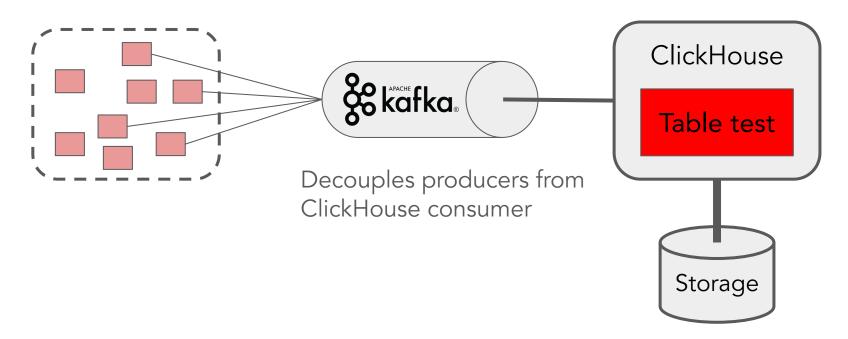


Enable async inserts using property settings

```
CREATE SETTINGS PROFILE IF NOT EXISTS `async profile`
                                                           Use async insert
ON CLUSTER '{cluster}'
                                                          and wait for answer
SETTINGS
  async insert = 1,
                                                            Wait this long
  wait for async insert=1,
  async insert busy timeout ms = 10000, 	←
                                                              Don't let
  async insert use adaptive busy timeout = 0
                                                            ClickHouse set
                                                           automatic values
CREATE USER IF NOT EXISTS async ON CLUSTER '{cluster}'
  IDENTIFIED WITH sha256 password BY 'topsecret' HOST ANY
  SETTINGS PROFILE `async profile`
                                                          User with settings
```

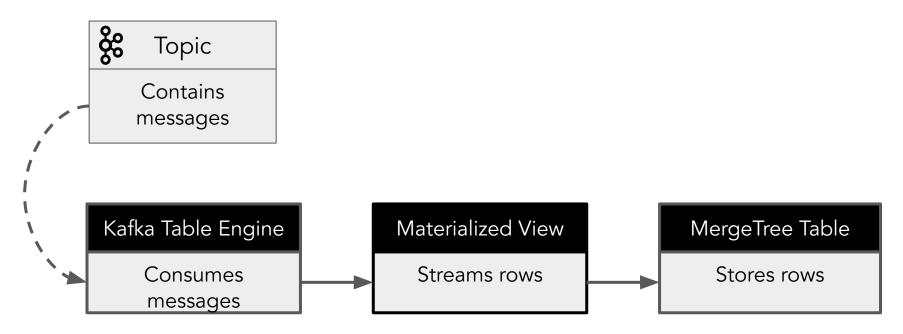


Fix #3: Use Kafka to buffer data from upstream producers





Using the Kafka table engine to read from Kafka



https://kb.altinity.com/altinity-kb-integrations/altinity-kb-kafka/



Lesson #4

Joins are different in ClickHouse. Learn to use them properly.



ClickHouse can do joins (of course!)

SELECT Dest, Name as AirportName, count(*) Flights
FROM default.ontime_ref o
LEFT JOIN default.dot_airports a ON (a.AirportID = o.DestAirportID)
GROUP BY Dest, AirportName ORDER BY Flights DESC LIMIT 10

—Dest—	—AirportName—————————————————	Flights
DEN	Denver International	12103062
ATL	Hartsfield Jackson Atlanta International Airport	10605117
LAS	McCarran International Airport	4361486

10 rows in set. Elapsed: 2.581 sec. Processed 196.52 million rows, 982.84 MB (76.13 million rows/s., 380.75 MB/s.)



ClickHouse has a rich set of algorithms and join types

Join Algorithms

hash
partial_merge
parallel_hash
grace_hash
full_sorting_merge
direct

Caveat: BI-style large table joins are "difficult"

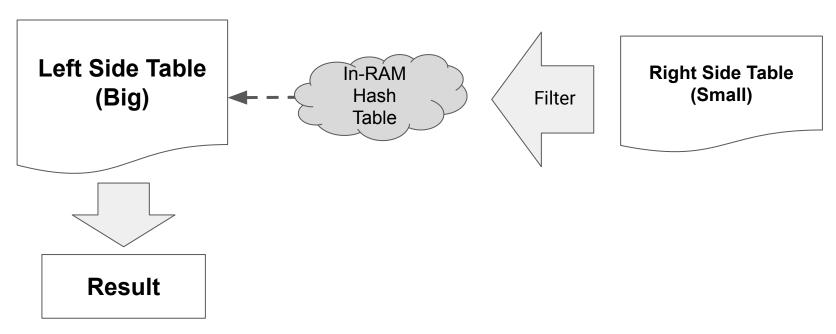
Join Types

inner join left [outer] join full [outer] join right [outer] join cross join semi join anti join asof join paste join



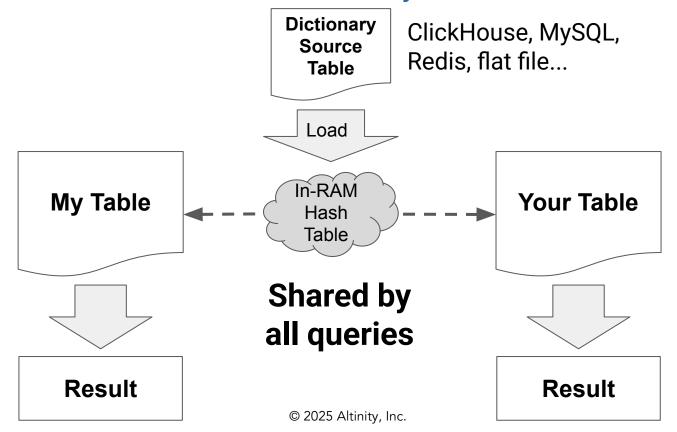
How ClickHouse does joins between tables

Hash Join Algorithm





Dictionaries are an alternative to joins





Creating a dictionary on a table

https://clickhouse.com/docs/en/sql-reference/statements/create/dictionary/



Restructure joins to reduce data scanning

SELECT Dest, Name, count(*) c, avg(ArrDelayMinutes) ad
 FROM default.ontime_ref
LEFT JOIN default.dot_airports ON DestAirportID = AirportID
GROUP BY Dest, Name HAVING c > 100000
ORDER BY ad DESC LIMIT 10



3.472 sec.

15.820 sec.

SELECT Dest, Name, c AS flights, ad

FROM (SELECT DestAirportID, any(Dest) as Dest,
count(*) c, avg(ArrDelayMinutes) ad
FROM default.ontime_ref
GROUP BY DestAirportID HAVING c > 100000
ORDER BY ad DESC LIMIT 10) a

LEFT JOIN default.dot airports ON DestAirportID = AirportID

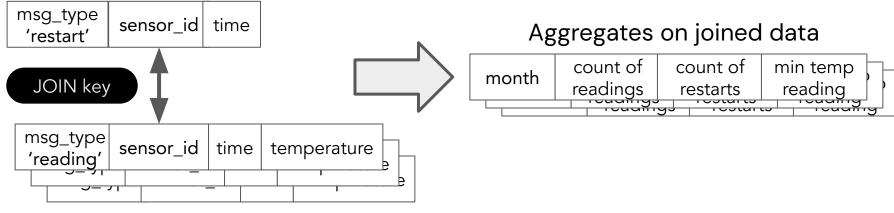
Smaller base query



How can we handle table joins on very large tables??

Use case: join sensor restart with temperature reading data

Restart times



Temperature readings

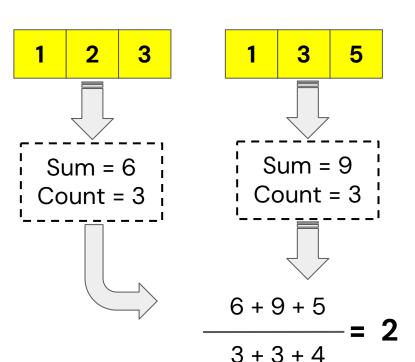


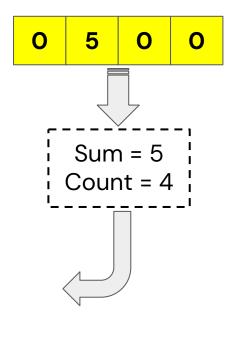
Hint: Aggregation runs in a single pass

No need to move data

Parallelizes!

Intermediate results are reusable







Conditional aggregation on different entities in one table

```
SELECT toYYYYMM(time) AS month,
    countIf(msg_type = 'reading') AS readings,
    countIf(msg_type = 'restart') AS restarts,
    minIf(temperature, msg_type = 'reading') AS min,
    round(avgIf(temperature, msg_type = 'reading')) AS avg,
    maxIf(temperature, msg_type = 'reading') AS max
FROM test.readings_multi WHERE sensor_id = 3
GROUP BY month ORDER BY month ASC
```

month—	extstyle ext	extstyle ext	min—	—avg—	max—
201901	44640	1	0	75	118.33
201902	40320	0	68.09	81	93.98
201903	15840	0	73.19	84	95.3



Lesson #5

ClickHouse prioritizes speed and scalability over anything else.

Lean on the trade-offs.



Size



Immutable data

Speed





Eventual consistency



ReplacingMergeTree deduplicates rows in ORDER BY

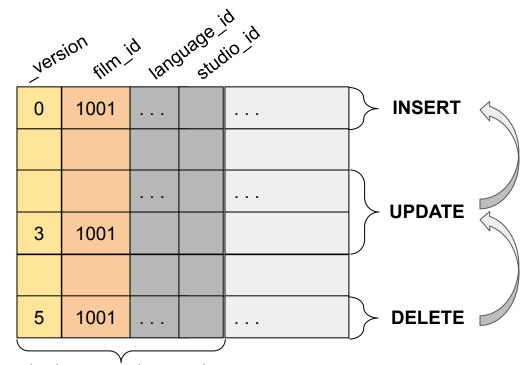
Other cols go to left

Row key goes on right (if you have one)



How ReplacingMergeTree works

Eventually consistent replacement of rows



De-duplicate on these columns



Updating a row in the RMT table

```
INSERT INTO sakila.film VALUES
(1001, 'Blade Runner - Director''s Cut', 'Best. Sci-fi. Film.
Ever.',...,3)
SELECT title, release year
FROM film WHERE film id = 1001
                                                          Unmerged
                                                          rowsl
                                   -release_year-
  title-
  Blade Runner - Director's Cut
                                    1982
                  -release year—
  Blade Runner
                  1982
```



Rows are replaced when merges occur

Part Merged Part 1001 Part 1001 1001 Pro tip: never assume rows will merge fully



FINAL keyword merges data dynamically

```
SELECT film_id, title
FROM sakila.film FINAL
WHERE film id = 1001

Adds initial scan to merge rows
```

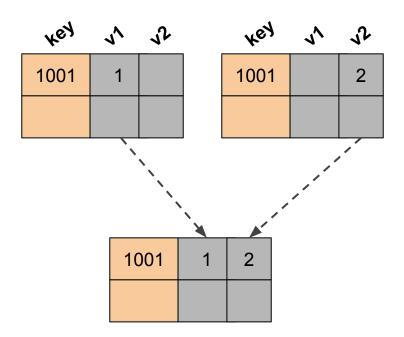
```
Title release_year Blade Runner - Director's Cut | 1982
```

https://altinity.com/blog/clickhouse-replacingmergetree-explained-the-good-the-bad-and-the-ugly



ClickHouse has a raft of tables built on similar principles

- ReplacingMergeTree
- SummingMergeTree
- AggregatingMergeTree
- CollapsingMergeTree
- VersionedCollapsingMergeTree
- CoalescingMergeTree





Wrap-up and Questions



Summary of 5 things every beginner should know

- Pick the right install for your work. There are lots of them!
- Pay close attention to partitioning, ordering, and compression in tables.
- Make your inserts as big as possible.
- Joins are different in ClickHouse. You need to think about the mechanisms.
- Immutable data and eventual consistency are the foundation of big systems. Lean in on features that use them.





Join our **Open Source Analytics Festival**

Where:

IBM Silicon Valley Lab 555 Bailey Ave San Jose, CA 95141 When: Thursday 31 July 2025

Thank you! Questions?

Contact us to learn more about Altinity.Cloud and Enterprise Support

https://altinity.com

https://altinity.com/slack

