

# Five Ways to Make your ClickHouse® Slow (and How to Avoid Them)

Robert Hodges - Altinity CEO

Mikhail Filimonov - ClickHouse Architect

18 February 2026



# ALTINITY®

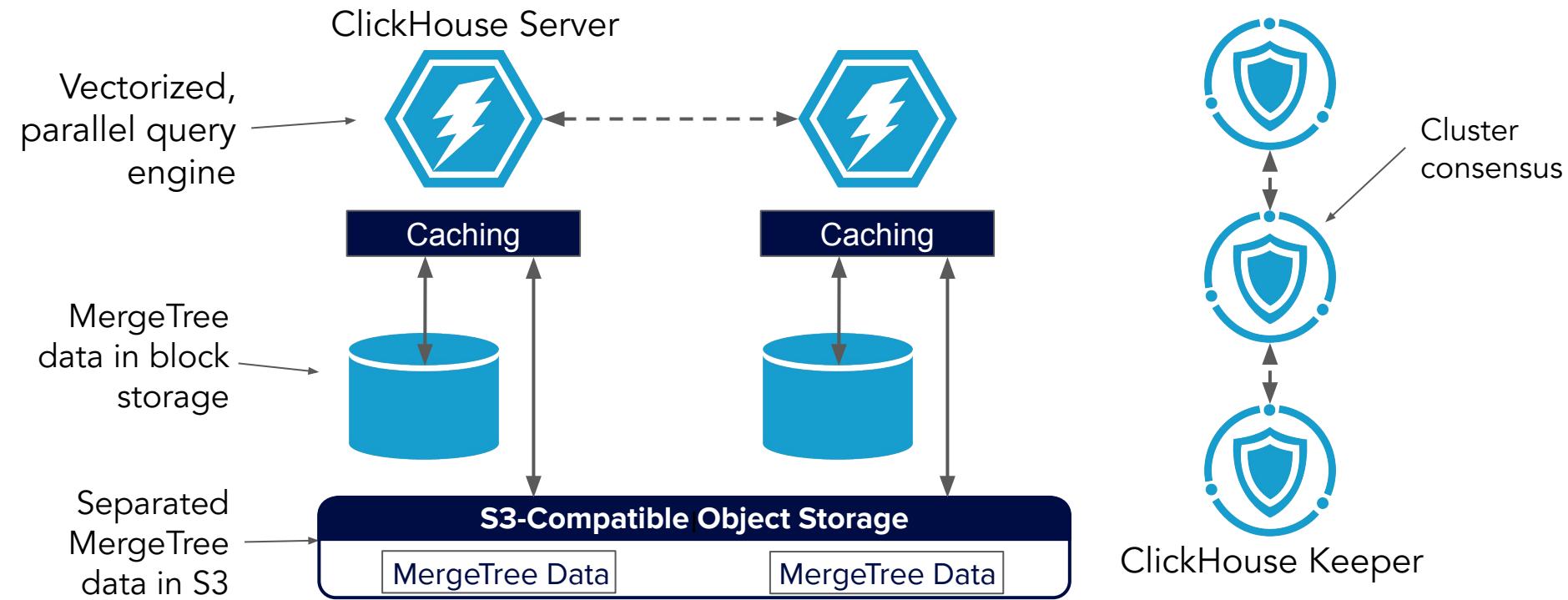
Run Open Source ClickHouse® Better

## Altinity.Cloud Enterprise Support

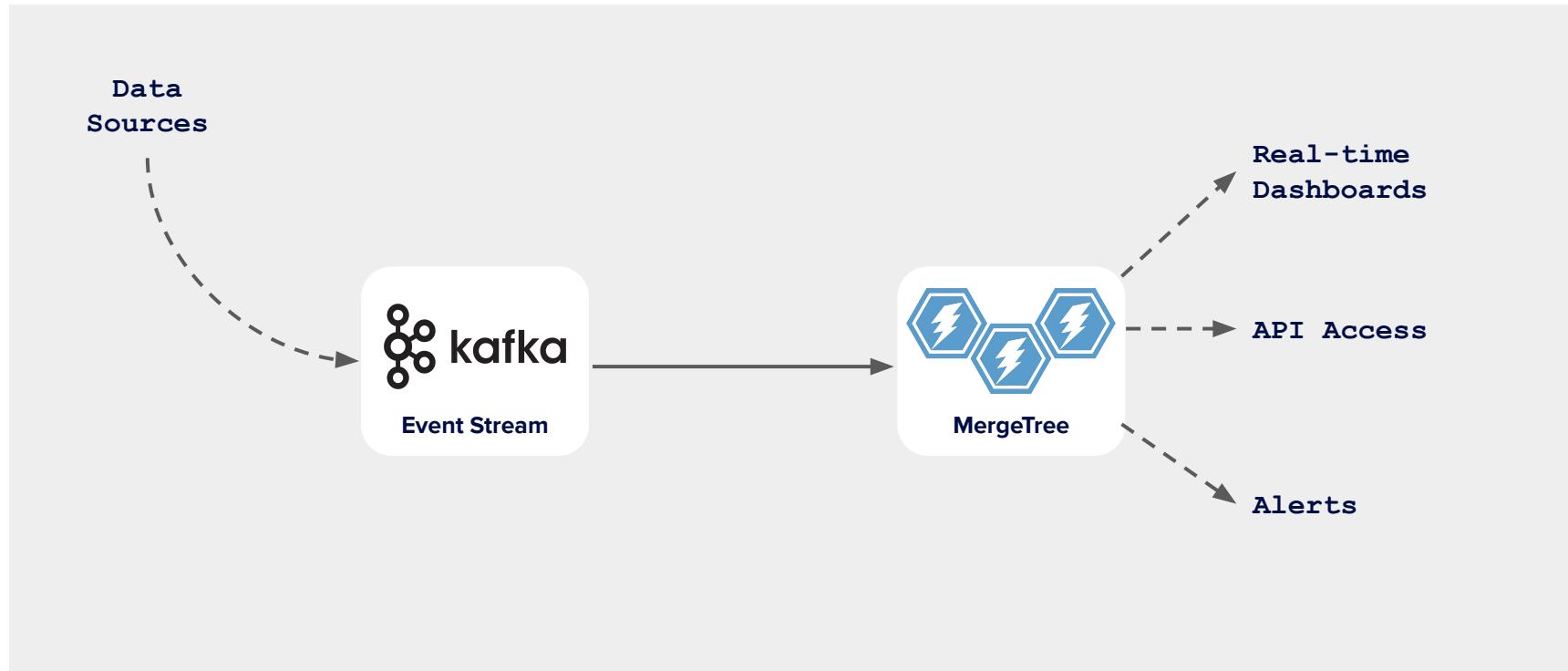
Altinity® is a Registered Trademark of Altinity, Inc.

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

# ClickHouse shared nothing architecture



# Traditional real-time event pipeline with ClickHouse



# ClickHouse works great for almost any real-time analytics

```
SELECT Carrier, toYear(FlightDate) AS Year,  
       (sum(Cancelled) / count(*)) * 100. AS cancelled_pct  
FROM default.ontime_ref  
GROUP BY Carrier, Year HAVING cancelled_pct > 1.  
ORDER BY cancelled_pct DESC LIMIT 10
```

	Carrier	Year	cancelled_pct
1.	G4	2020	16.733186040434276
2.	EA	1989	10.321500966388536
3.	WN	2020	9.284307653599388
...			

10 rows in set. Elapsed: 0.825 sec. Processed 196.51 million rows, 982.57 MB (756.93 million rows/s., 1.19 GB/s.)

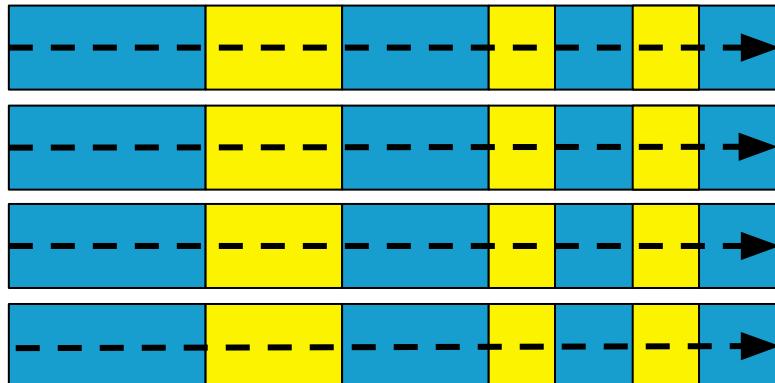
# But...we did everything possible to make it slow!!

- Underpowered AWS m7g.xlarge Graviton with 4 vCPUs & 14GB RAM
- Slowest EBS storage speed: 125 MiB/sec
- Force cold reads with **SETTINGS min\_bytes\_to\_use\_direct\_io = 1**

# The secret of ClickHouse's success: compressed columns

PostgreSQL, MySQL

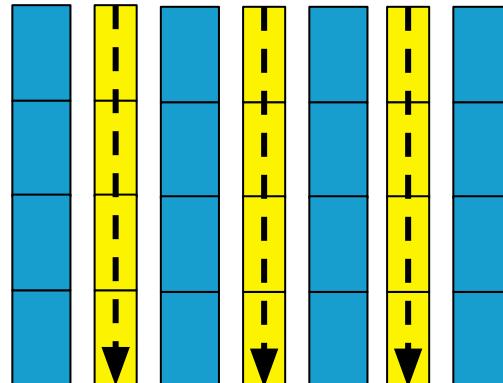
Read all columns in row



Rows minimally or not compressed

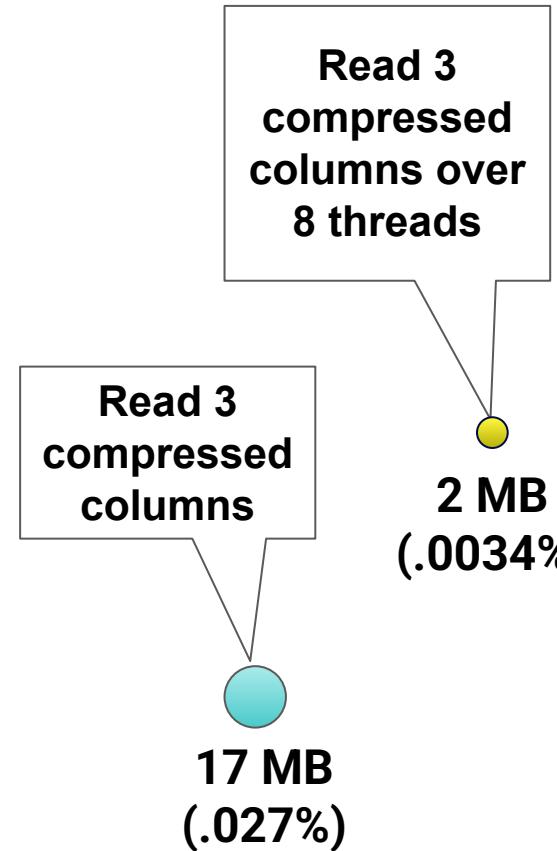
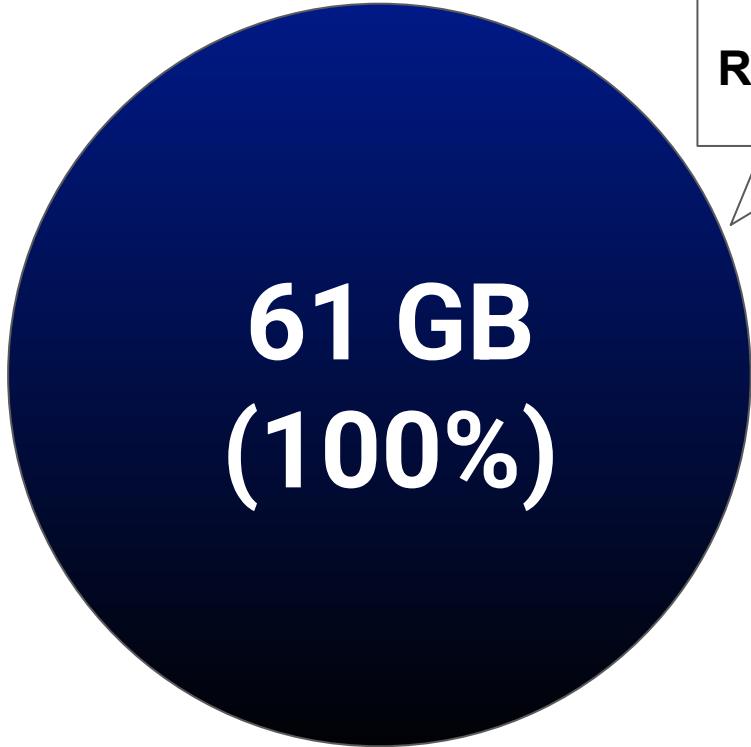
ClickHouse

Read only selected columns



Columns highly compressed

# Visualizing effect of columns and compression





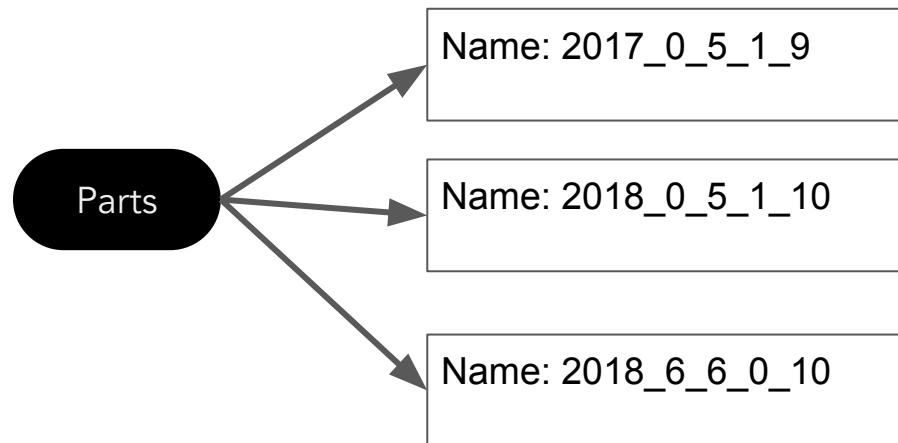
# So...What could possibly go wrong?

## Problem #1

# Bad table design

# Best practice: partition by time

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



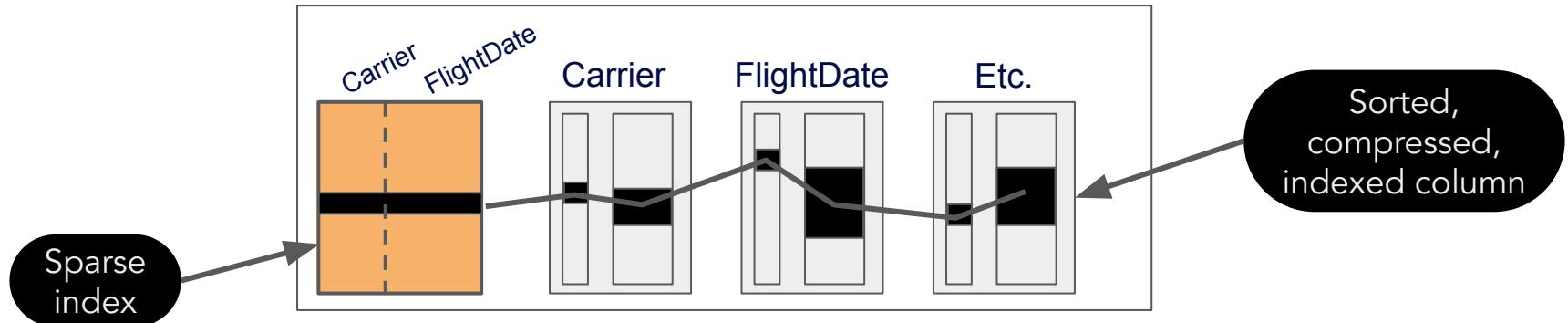
Rule of thumb:

Choose partitions that result in ~1000 parts or less

# 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



# But what if we made a different choice of schema?

```
CREATE TABLE test.ontime_bad_partitioning
AS default.ontime_ref
ENGINE = MergeTree
PARTITION BY (Carrier, toYYYYMM(FlightDate))
ORDER BY (Carrier, FlightDate)

INSERT INTO test.ontime_bad_partitioning
SELECT *
FROM default.ontime_ref
SETTINGS max_threads = 1, max_insert_threads = 1
```

Pro tip: Reduce threads to avoid running out of memory

# We can now make ClickHouse really slow!

```
SELECT Carrier, toYear(FlightDate) AS Year,  
       (sum(Cancelled) / count(*)) * 100. AS cancelled_pct  
FROM test.ontime_bad_partitioning  
GROUP BY Carrier, Year HAVING cancelled_pct > 1.  
ORDER BY cancelled_pct DESC LIMIT 10  
[SETTINGS min_bytes_to_use_direct_io = 1]
```

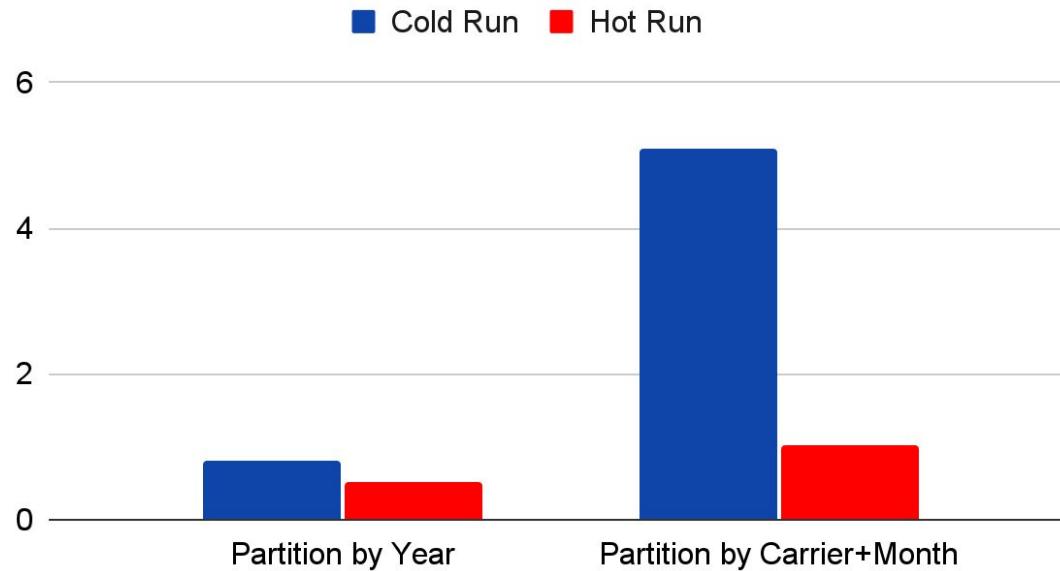
	Carrier	Year	cancelled_pct
1.	G4	2020	16.733186040434276

10 rows in set. Elapsed: 5.092 sec. Processed 196.51 million  
rows, 982.57 MB (38.59 million rows/s., 160.74 MB/s.)

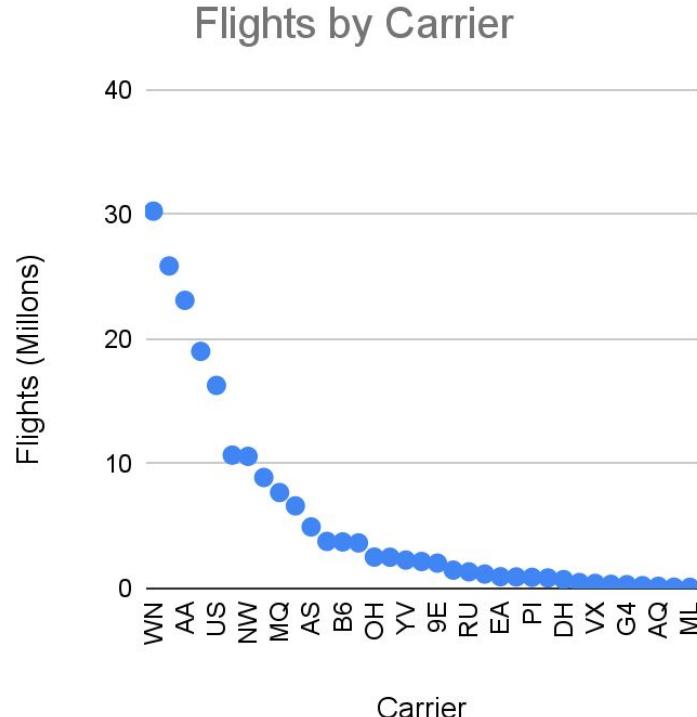
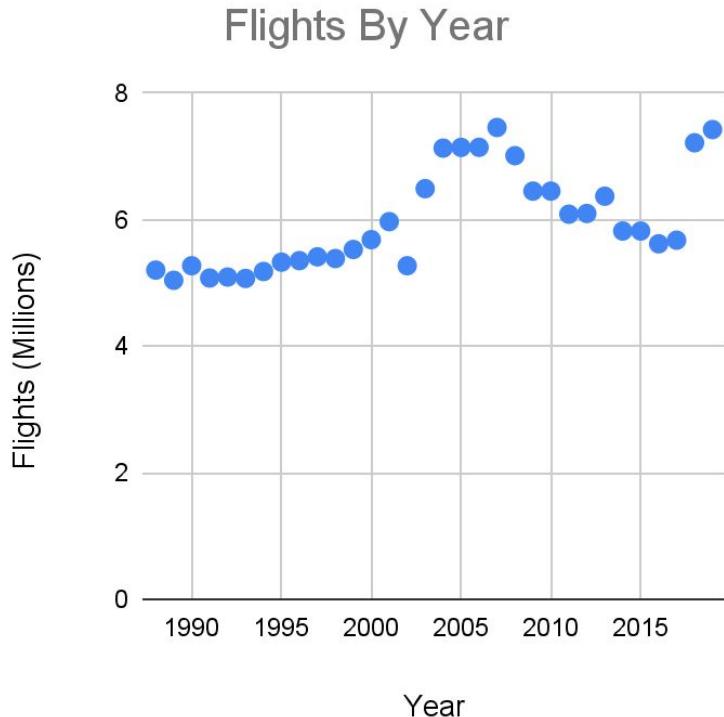
Force  
direct I/O

# Bad partitioning == bad performance!!

## Effects of partitioning choices



# Why it's better to partition by year?



# Cheat sheet for schema design

Hard to change!

1. Time-based column in **PARTITION BY**
2. Put tenants using **ORDER BY**, then add columns in order of cardinality
3. Use appropriate datatypes (e.g., Int32, not String)
4. Use codecs like Delta or LowCardinality
5. Use ZSTD compression instead of default LZ4 to really squeeze space
  - a. Use it *\*if\** you hit I/O limits but have free CPU capacity

# Measure compression with amazing system tables!

```
SELECT
    count() ,
    formatReadableSize(sum(data_compressed_bytes) ,
    formatReadableSize(sum(data_uncompressed_bytes)
FROM system.columns
WHERE (database = 'default') AND (`table` = 'ontime_ref')
AND (name IN ('Carrier', 'FlightDate', 'Cancelled'))
```

Other great tables: system.parts and system.tables



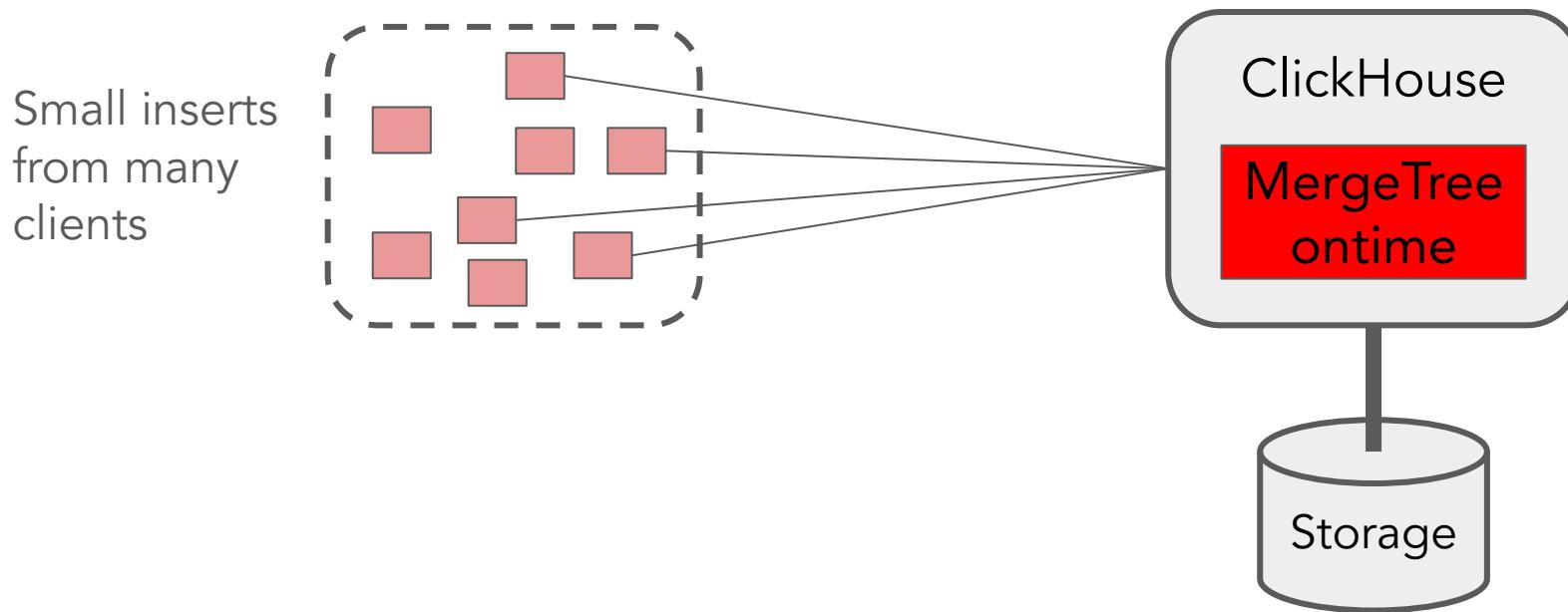
## Problem #2

# Too many tiny inserts

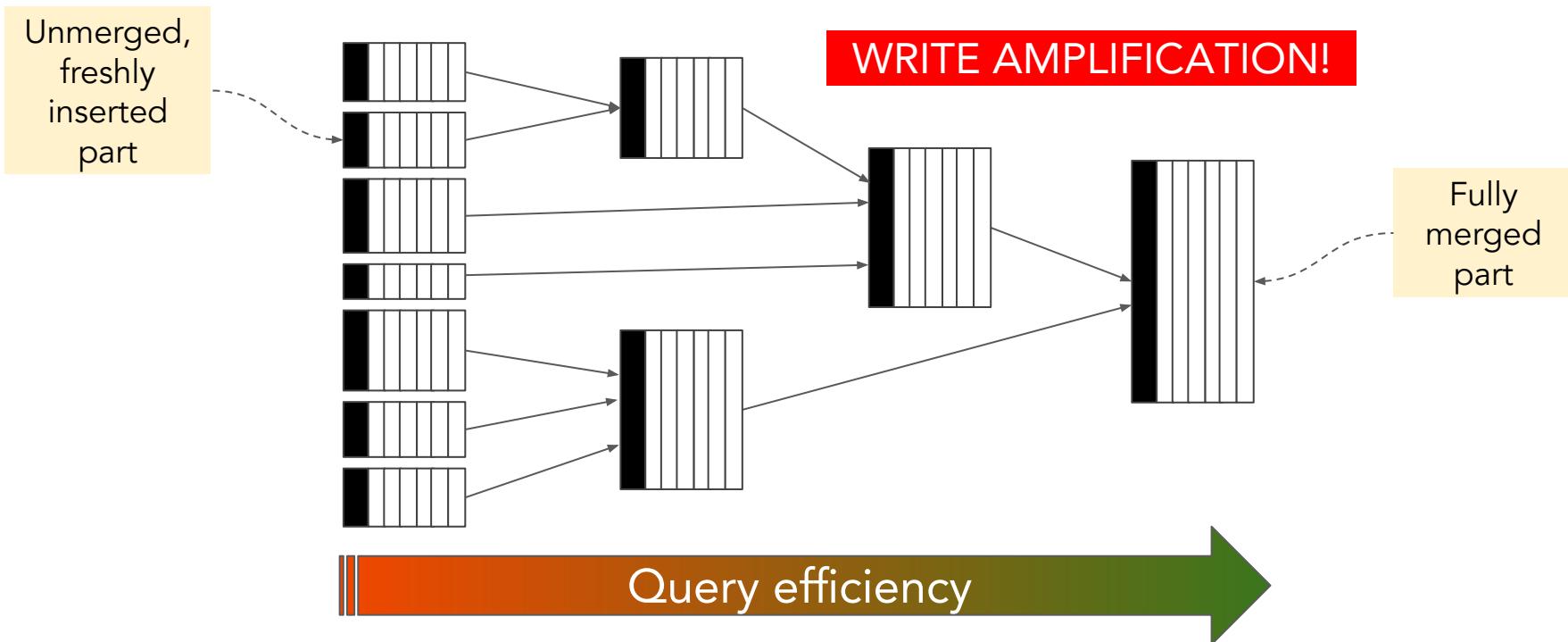
# If your inserts look like this, you are doing it wrong!

```
INSERT INTO default.ontime_ref VALUES
(2017,4,12,12,2,'2017-12-12','UA\0\0\0\0\0',
19977,'UA',...),
(2017,4,12,12,2,'2017-12-12','UA\0\0\0\0\0',
19977,'UA',...)
```

# Small inserts can crush your ClickHouse server



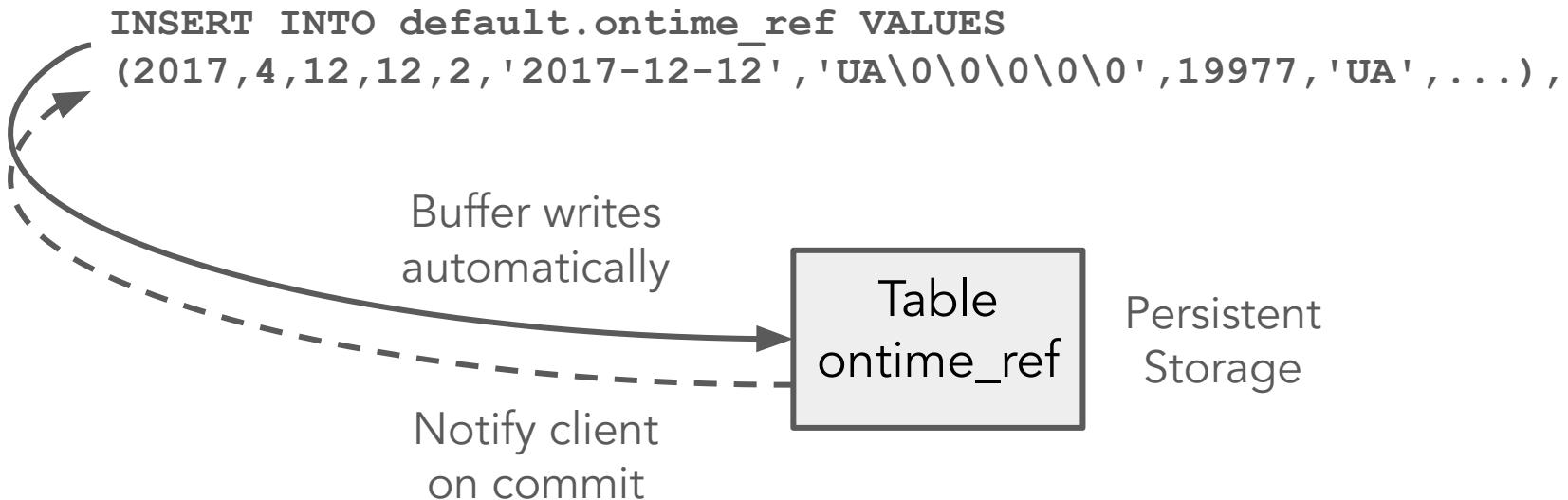
# Lots of small parts == slow queries and high merge load



# Fix #1: Use big batches in your application

```
#!/bin/bash
INSERT='INSERT+INTO+ontime+Format+CSVWithNames'
cat test.csv | curl -X POST --data-binary @- \
"http://localhost:8123/?query=${INSERT}"
```

## 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`  
ON CLUSTER '{cluster}'  
SETTINGS  
    async_insert = 1,           ←----- Use async insert  
    wait_for_async_insert=1,    ←----- and wait for answer  
    async_insert_busy_timeout ms = 10000 ←----- Wait this long  
    async_insert_use_adaptive_busy_timeout = 0 ←----- Don't let  
;                                         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`  
;
```



## Problem #3

# Bad queries

# Small differences in queries make big differences in response

```
0.64 sec  
114 KB RAM
```

**4.25x slower**  
**21,891x more RAM used**

```
2.72 sec  
2.38 GB RAM
```

```
SELECT Carrier,  
       avg(DepDelay)AS Delay  
  FROM ontime_ref  
 GROUP BY Carrier  
 ORDER BY Delay DESC  
LIMIT 50
```

```
SELECT Carrier, FlightDate,  
       avg(DepDelay) AS Delay,  
       uniqExact(TailNum) AS Aircraft  
  FROM ontime_ref  
 GROUP BY Carrier, FlightDate  
 ORDER BY Delay DESC  
LIMIT 50
```

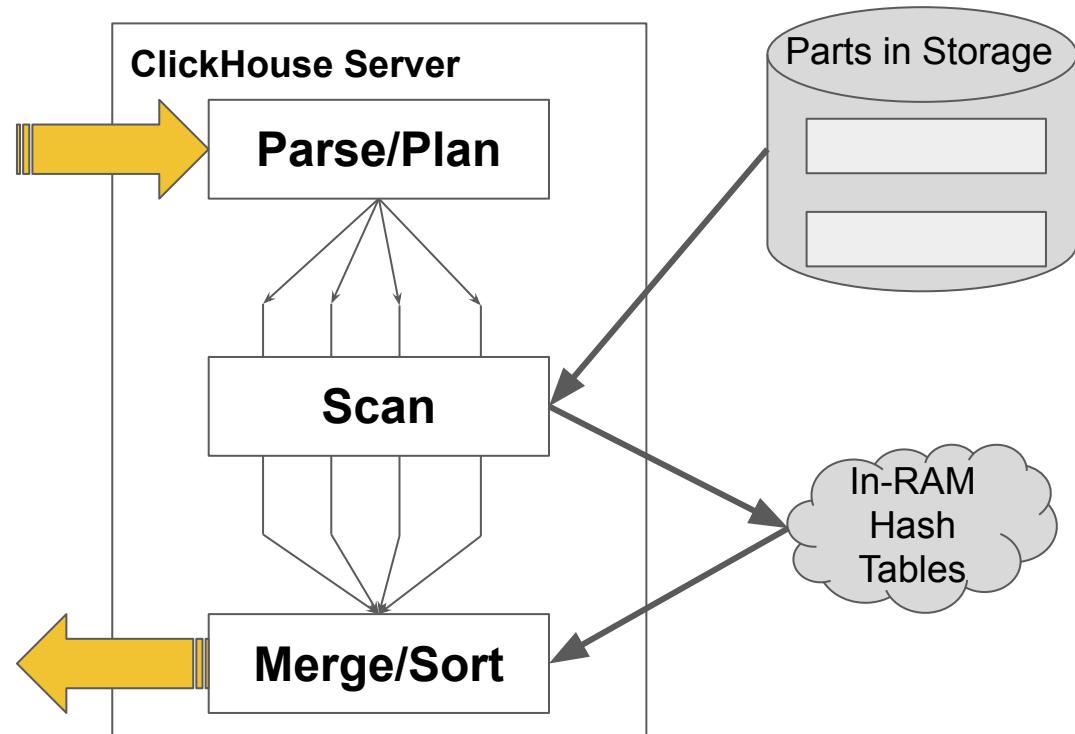
Simple aggregate, short  
GROUP BY key with few values

More complex aggregates, longer  
GROUP BY with more values

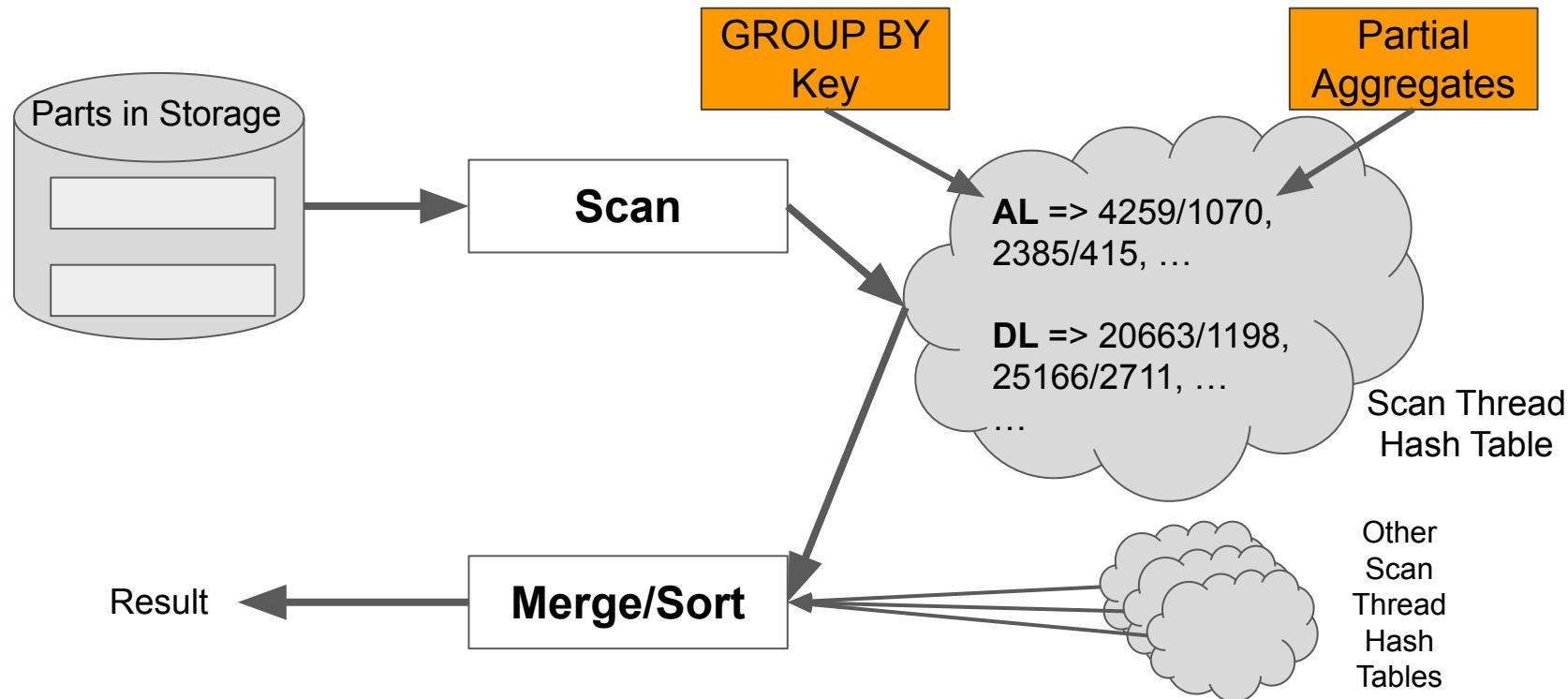
# How does ClickHouse process a query with aggregates?

```
SELECT Carrier,  
       avg(DepDelay) AS Delay  
  FROM ontime  
 GROUP BY Carrier  
 ORDER BY Delay DESC
```

Carrier	Delay
B6	12.058290698785067
EV	12.035012037703922
NK	10.437692933474269
...	



# How does a ClickHouse thread do aggregation?



# We can optimize queries by choosing better aggregates

```
2.72 sec  
2.38 GB RAM
```

15% faster

5.4x less RAM used

```
2.30 sec  
450 GB RAM
```

```
SELECT Carrier, FlightDate,  
       avg(DepDelay) AS Delay,  
       uniqExact(TailNum) AS Aircraft  
  FROM ontime  
 GROUP BY Carrier, FlightDate  
 ORDER BY Delay DESC  
 LIMIT 50
```

uniqExact stores each unique value in a hash table that grows

```
SELECT Carrier, FlightDate,  
       avg(DepDelay) AS Delay,  
       uniqHLL12(TailNum) AS Aircraft  
  FROM ontime  
 GROUP BY Carrier, FlightDate  
 ORDER BY Delay DESC  
 LIMIT 50
```

uniqHLL12 uses fixed size HyperLogLog structure

# Here's some magic with optimizing joins

```
SELECT o.Dest, any(a.Name) AS AirportName,  
       count(Dest) AS Flights  
  FROM ontime_ref o  
 JOIN default.airports a ON a.IATA = o.Dest  
 GROUP BY Dest ORDER BY Flights  
 DESC LIMIT 10
```

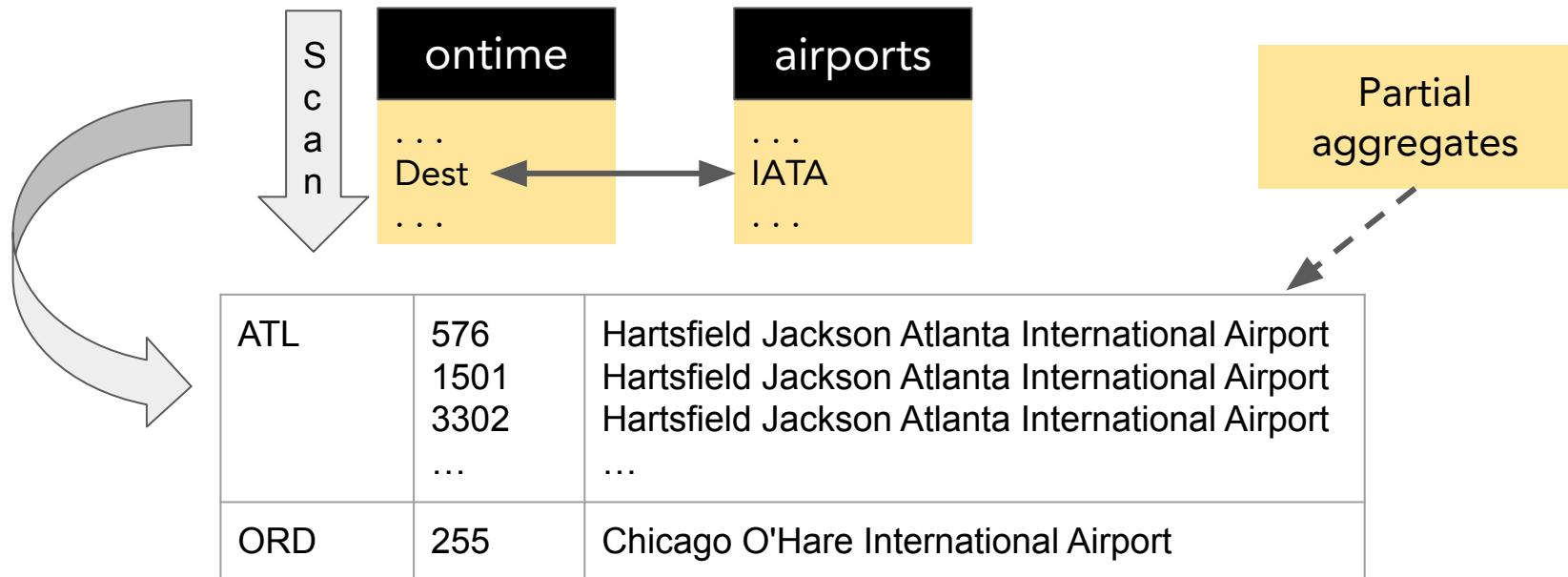
2.685 sec  
39.18 MB RAM

```
SELECT o.Dest, a.Name AS AirportName, o.Flights  
  FROM (  
    SELECT Dest, count(Dest) AS Flights  
      FROM ontime_ref GROUP BY Dest ) AS o  
 JOIN default.airports a ON a.IATA = o.Dest  
 ORDER BY Flights DESC LIMIT 10
```

0.524 sec  
1.02 MB RAM

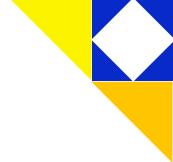
# Let's look more deeply at what's happening in the scan

```
SELECT . . . FROM ontime o JOIN airports a ON a.IATA = o.Dest
```



# Where did those awesome query stats come from?

```
SELECT
    event_time,
    type,
    is_initial_query,
    query_duration_ms / 1000 AS duration,
    read_rows,
    read_bytes,
    result_rows,
    formatReadableSize(memory_usage) AS memory,
    query
FROM system.query_log
WHERE (user = 'default') AND (type = 'QueryFinish')
ORDER BY event_time DESC
LIMIT 50
```



# Fixing queries efficiently

#1: Run against real data (and plenty of it)

#2: Isolate slow queries and optimize them

#3: Rinse and repeat



## Problem #4

# Insufficient resources

# Is your query still **too slow?** Throw money at it!

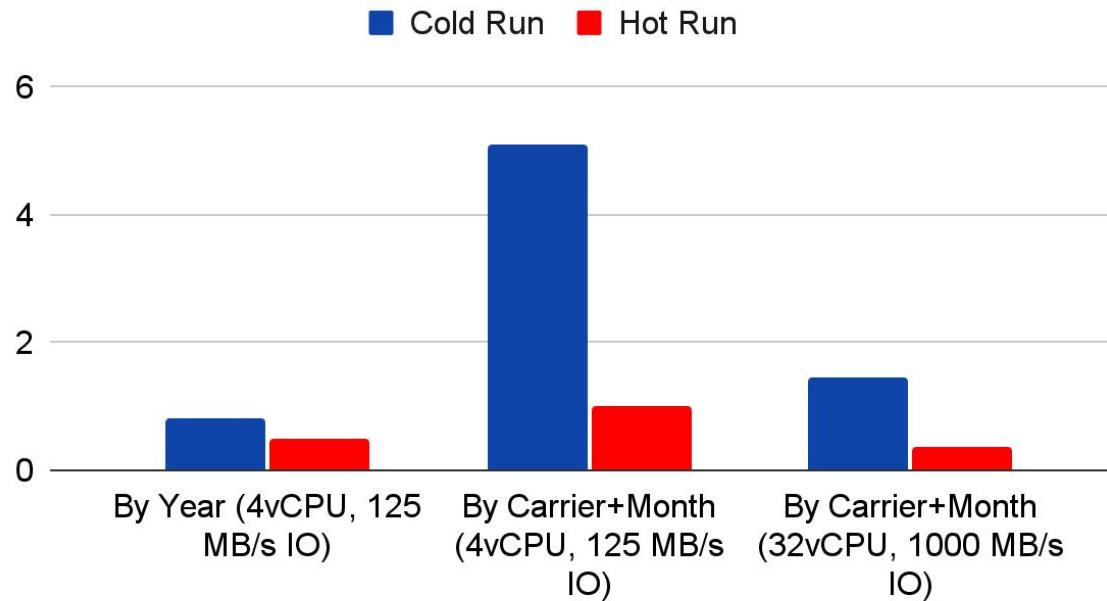
```
SELECT Carrier, toYear(FlightDate) AS Year,  
       (sum(Cancelled) / count(*)) * 100. AS cancelled_pct  
FROM test.ontime_bad_partitioning  
GROUP BY Carrier, Year HAVING cancelled_pct > 1.  
ORDER BY cancelled_pct DESC LIMIT 10  
[SETTINGS min_bytes_to_use_direct_io = 1]
```

	Carrier	Year	cancelled_pct
1.	G4	2020	16.733186040434276

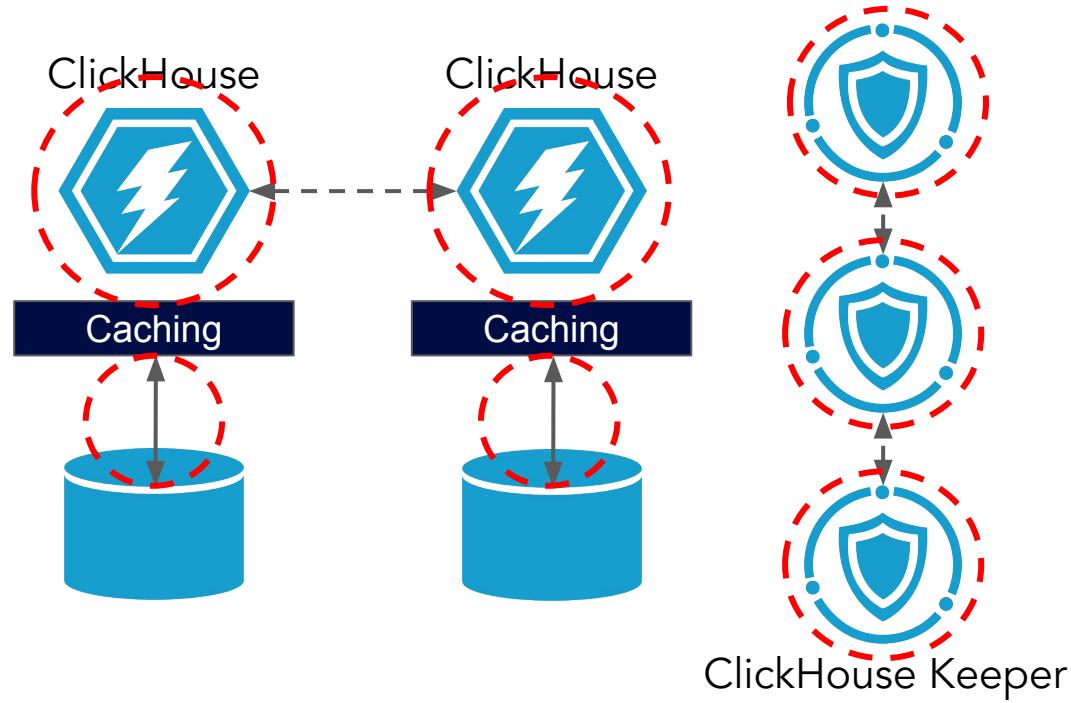
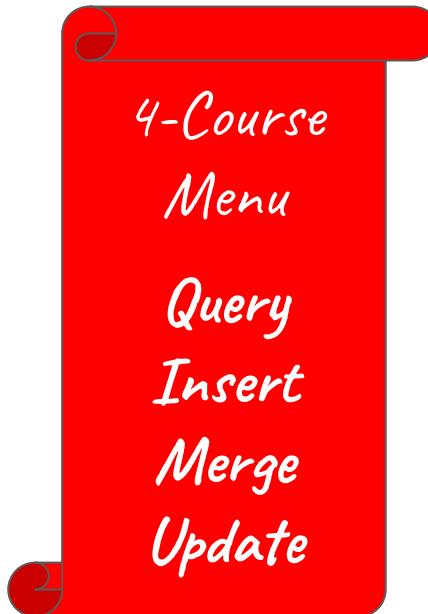
10 rows in set. Elapsed: 5.092 sec. Processed 196.51 million  
rows, 982.57 MB (38.59 million rows/s., 160.74 MB/s.)

# Better hardware can make slow queries fast...

Effects of partitioning choices with more resources



# But ClickHouse servers don't just run queries...





## Common issues with resource management

#1: Testing real workloads (**large & concurrent**)

#2: Detecting trouble: CPU, IOWait, RAM, Network

#3: Scaling quickly when trouble hits

#4: Fixing apps that overuse resources

## Problem #5

Migrations from  
non-compatible  
databases to ClickHouse

# Some migrations to ClickHouse just work

PostgreSQL, MySQL

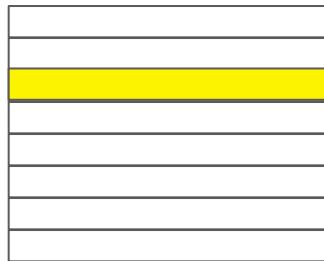
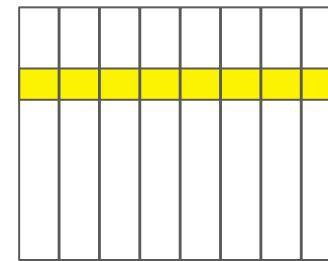


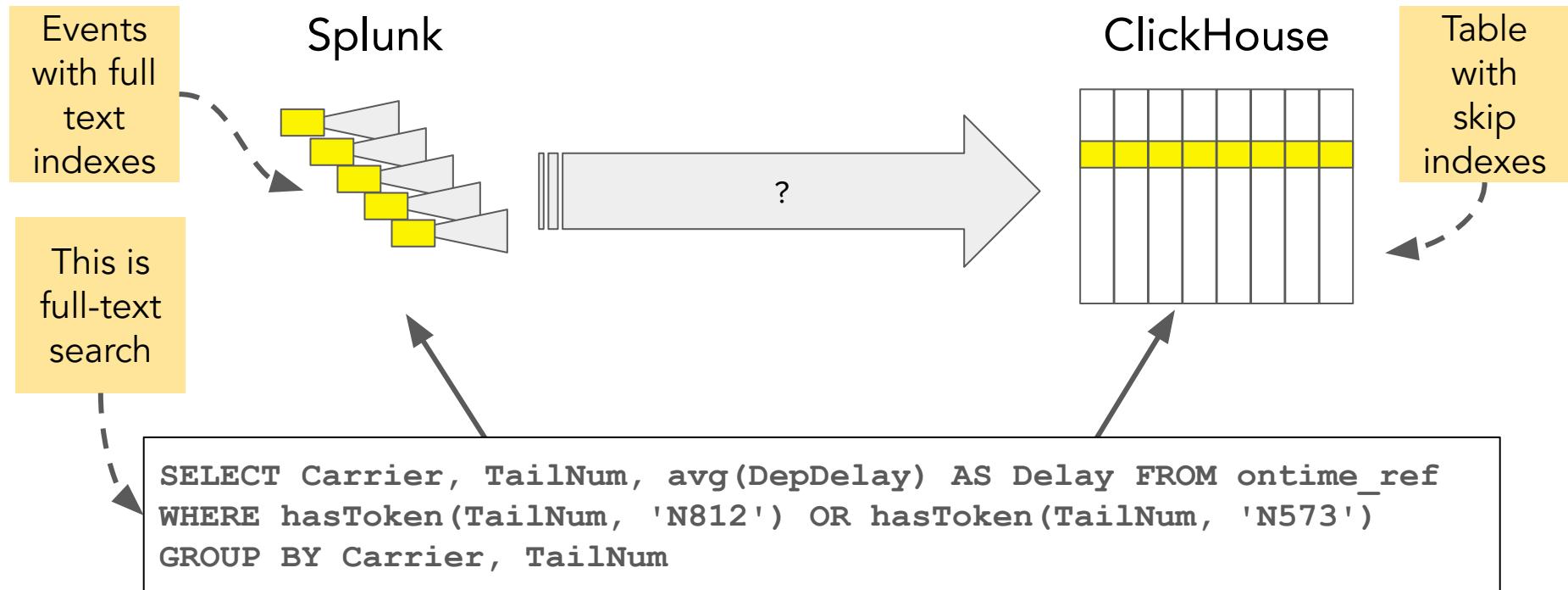
Table-to-Table Migration

ClickHouse



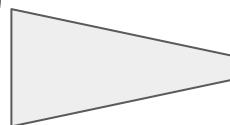
```
SELECT Carrier, avg(DepDelay)AS Delay FROM ontime_ref  
WHERE TailNum = 'N812AW' AND Year = 2016 GROUP BY Carrier
```

# Others are more “challenging”

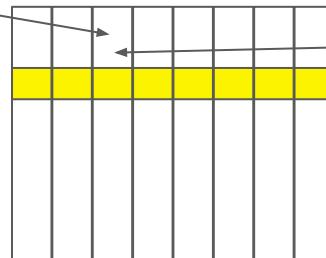


# We need to decide how to implement full text search

1



ClickHouse



Full text index – In beta,  
potentially very large

2

Bloom filter index  
(ngram\_bf) – Small but  
needs careful tuning

```
SELECT Carrier, TailNum, avg(DepDelay) AS Delay FROM ontime_ref
WHERE TailNum LIKE 'N812%' OR TailNum LIKE 'N573%'
GROUP BY Carrier, TailNum
```

3

LIKE operator – It's fast and ngram\_bf index makes it faster



# Migrating different database types to ClickHouse

#1: Test queries under realistic load

#2: Rethink slow query patterns

#3: It takes time to tune indexes and queries

# Server log messages are your friend

```
SELECT Carrier, TailNum, avg(DepDelay) AS Delay FROM
rhodes_7273b.ontime_bloom_filter
WHERE TailNum LIKE '%N812%' OR TailNum LIKE '%N128%'
GROUP BY Carrier, TailNum ORDER BY Delay DESC LIMIT 10
SETTINGS send logs level='debug'
```

```
... 2026.02.18 07:00:54.615183 [ 31 ]
{caf33ae2-a6ee-424d-b4f0-0ac022edab32} <Debug> executeQuery: (from
[::ffff:10.129.59.185]:60032, user: admin) (query 1, line 1) SELECT ...
... 2026.02.18 07:00:54.630439 [ 31 ]
{caf33ae2-a6ee-424d-b4f0-0ac022edab32} <Debug>
rhodes_7273b.ontime_bloom_filter (SelectExecutor): Index
`TailNum Ngrambf` has dropped 14033/24089 granules, it took 21ms across 4
threads
```



# Avoiding the Five Performance Problems

- Tune your schema to reduce I/O
- Make inserts as big as possible
- Test queries on real data and fix the slow ones
- Test hardware on realistic workloads and increase it before you hit problems
- Design and test migrations from other databases carefully!

Don't assume ClickHouse will be fast. Prove it!!!

Check out our TTL Guide!



# Thank you! Questions?

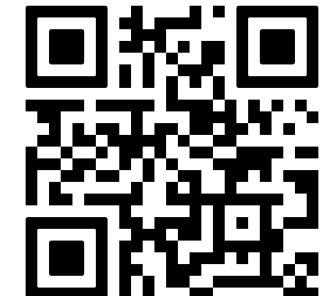
Robert Hodges  
CEO Altinity

<https://altinity.com>

We're hiring!



My LinkedIn

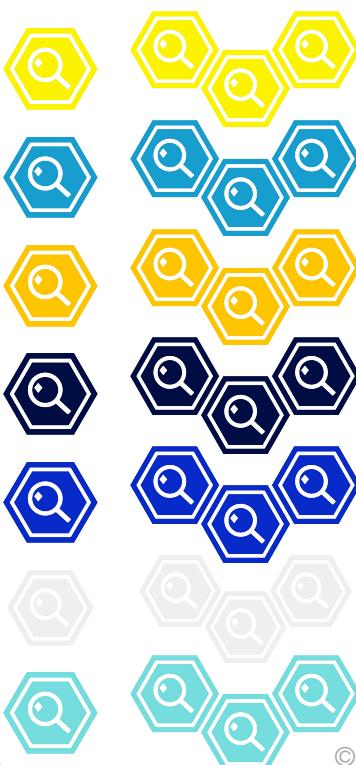


# Icons- Transparent

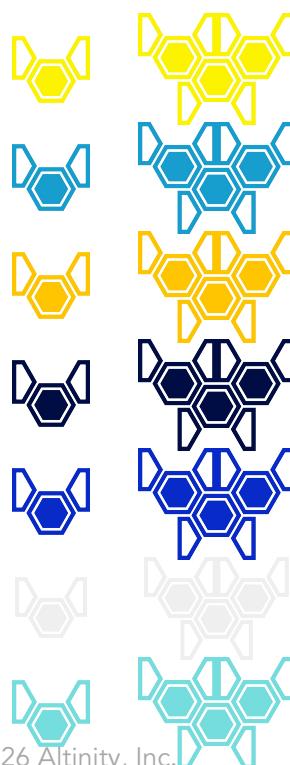
Clickhouse (Native) Cluster



Director Cluster



Swarm Cluster



Keeper



Other



# Icons-Stackable on White Backgrounds

Clickhouse (Native) Cluster



Director Cluster



Swarm Cluster



Keeper



Other



# Icons-Stackable on Dark Backgrounds

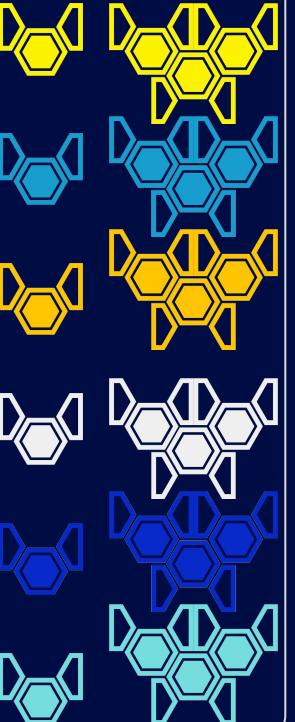
Clickhouse (Native) Cluster



Director Cluster



Swarm Cluster



Keeper



Other

