

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

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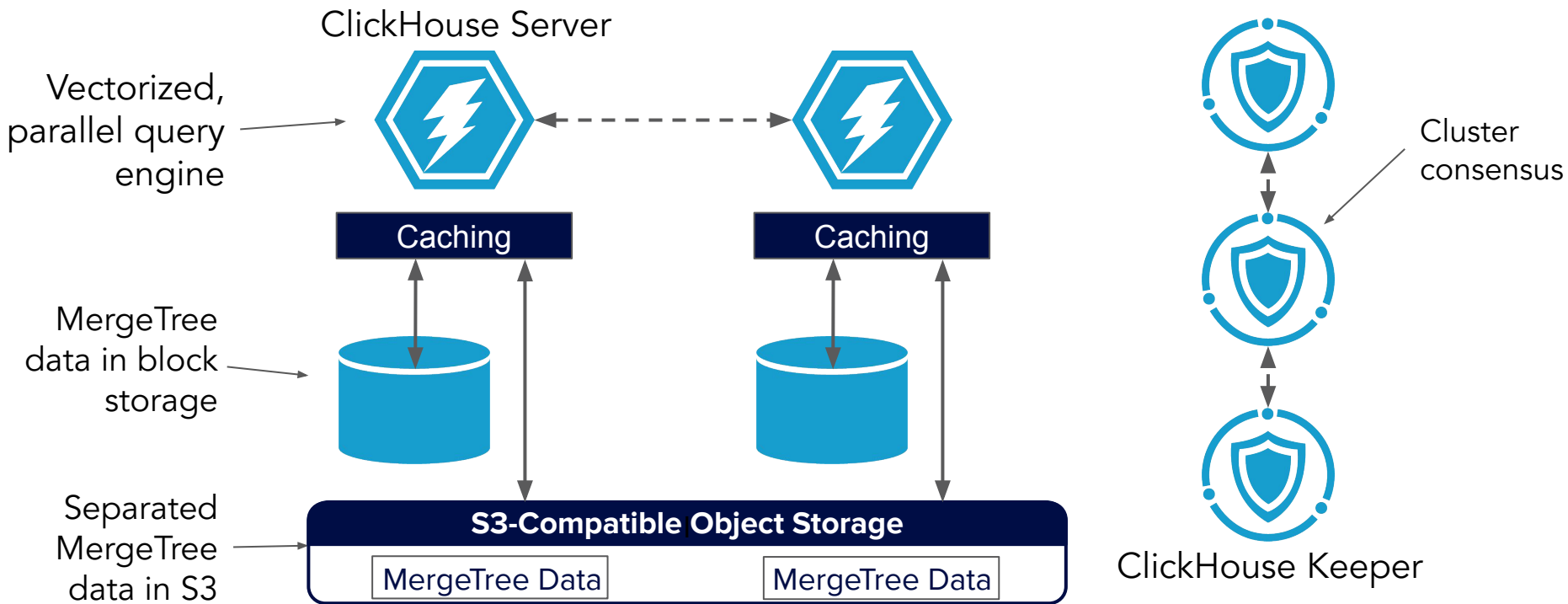
ALTINITY®

Run Open Source ClickHouse® Better

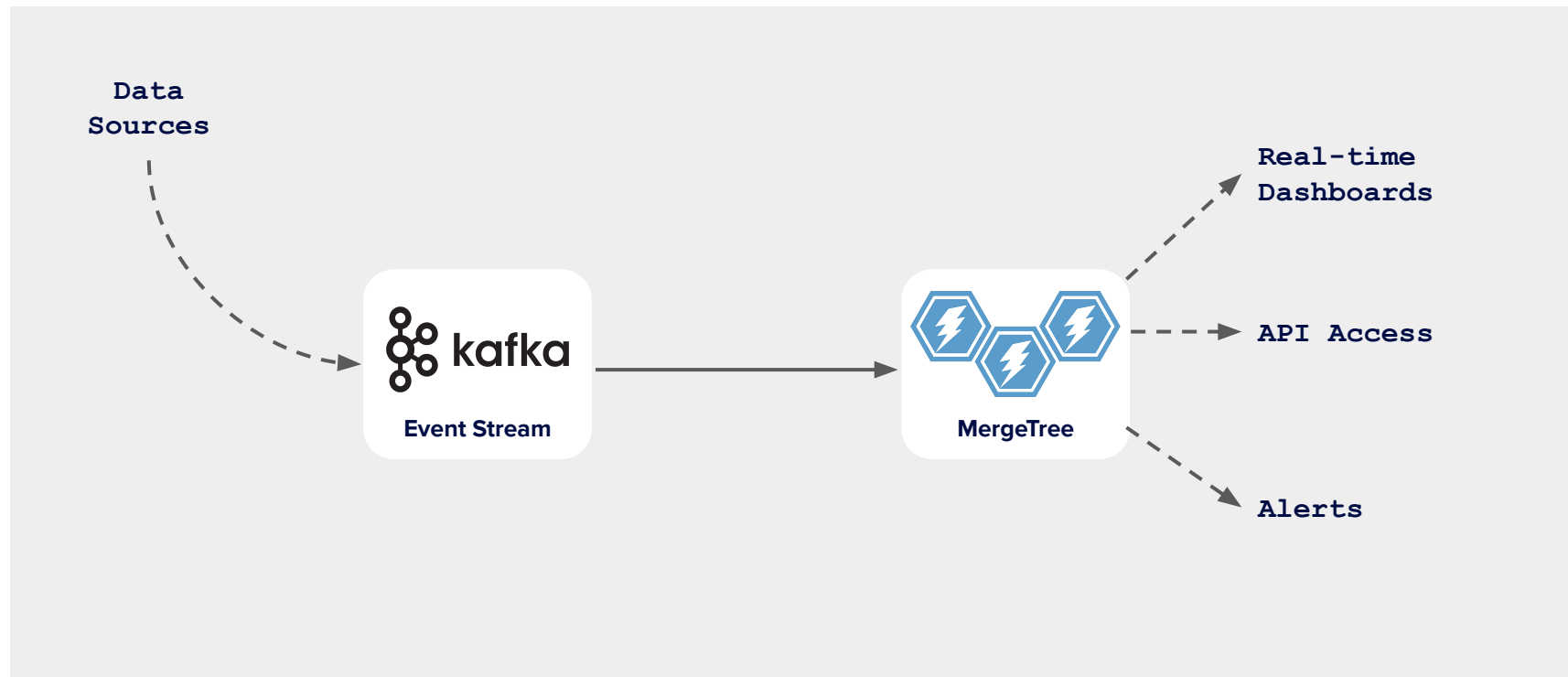
Altinity.Cloud Enterprise Support

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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.)

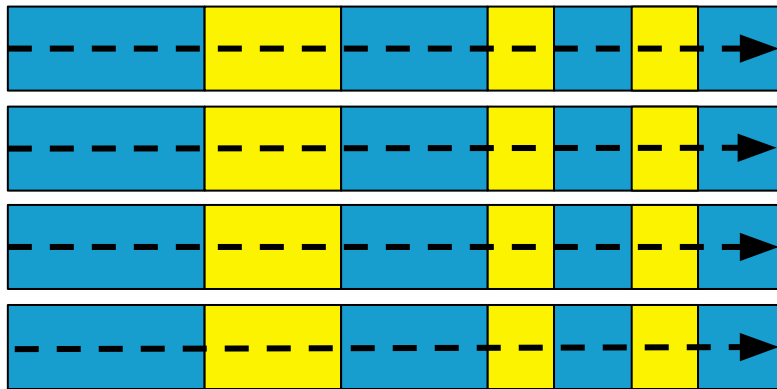
But...we did everything possible to make it slow!!

- 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

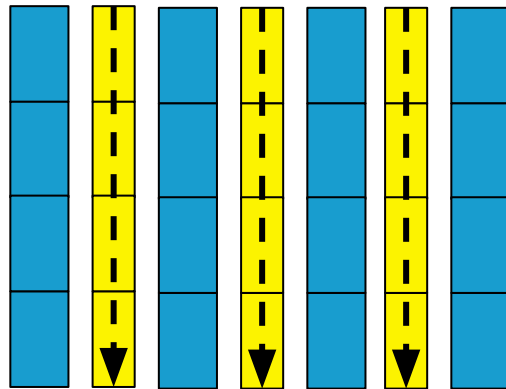
Read all columns in row



Rows minimally or not compressed

ClickHouse

Read only selected columns



Columns highly compressed

Visualizing effect of columns and compression

**61 GB
(100%)**

Read every row

**Read 3 columns:
Carrier,
FlightDate,
Cancelled**


**937 MB
(1.5%)**

**Read 3
compressed
columns**

**17 MB
(.027%)**

**Read 3
compressed
columns over
8 threads**

**2 MB
(.0034%)**



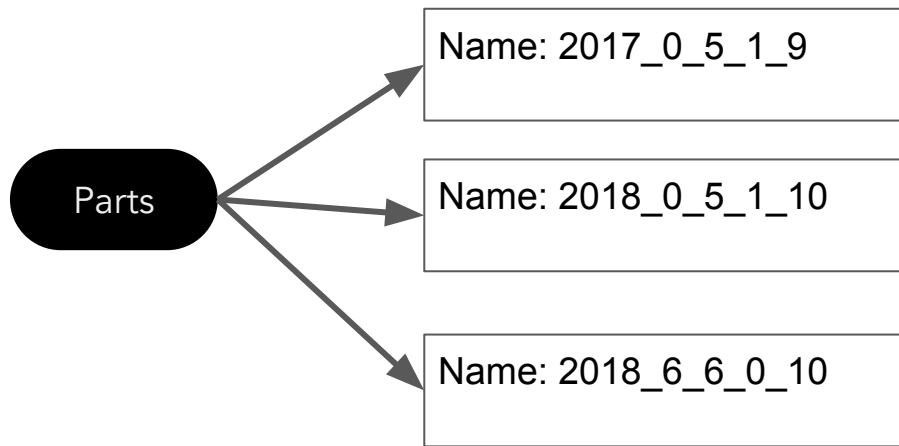
So...What could
possibly go wrong?

Problem #1

Bad table design

Best practice: partition by time

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



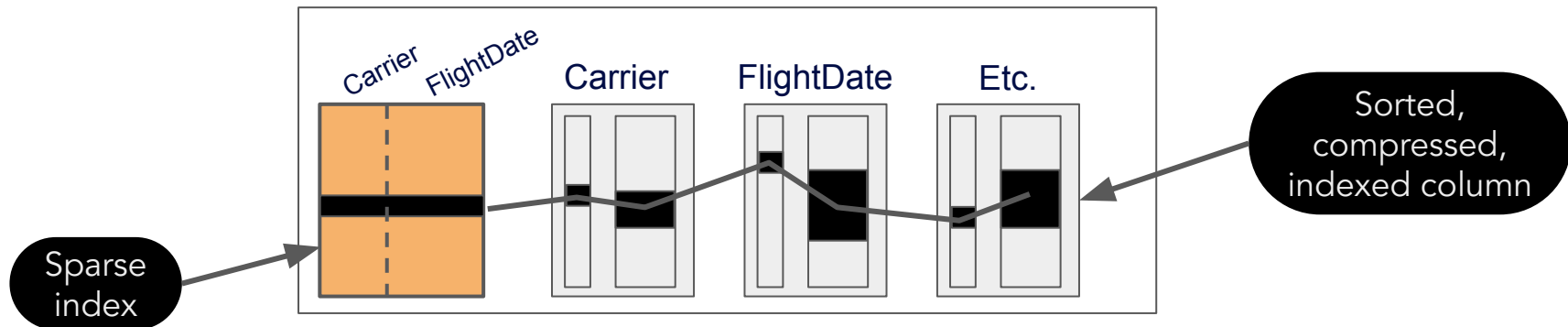
Rule of thumb:

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]
```

Force
direct I/O

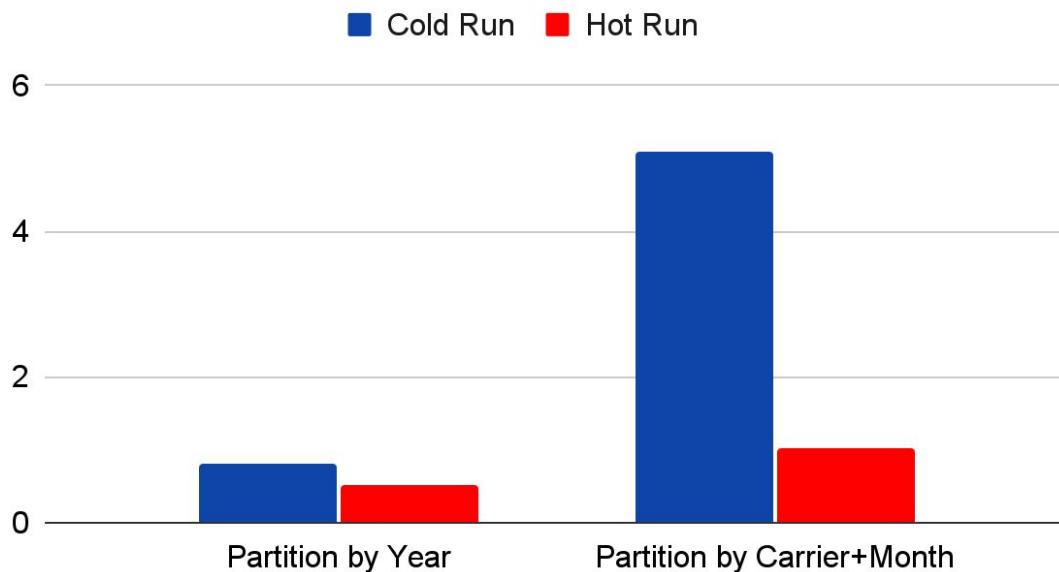
	Carrier	Year	cancelled_pct
1.	G4	2020	16.733186040434276

. . .

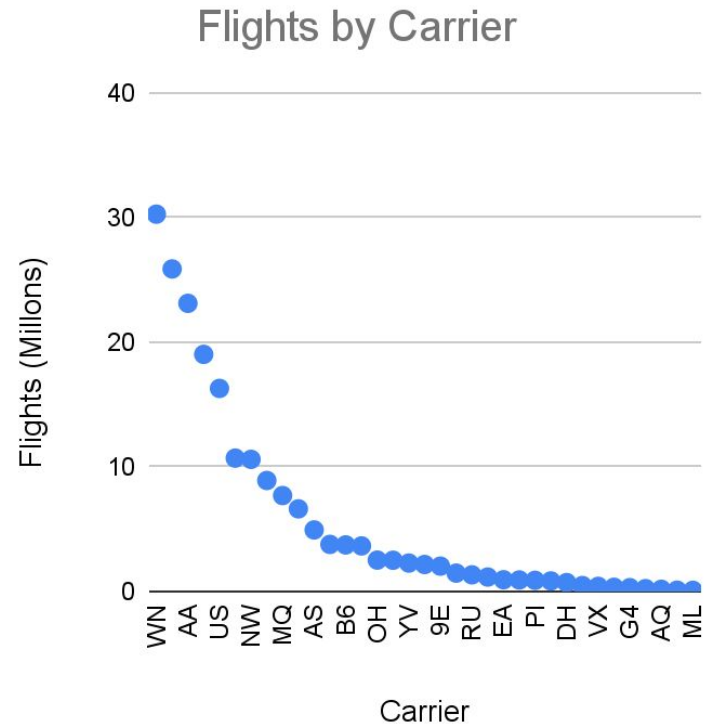
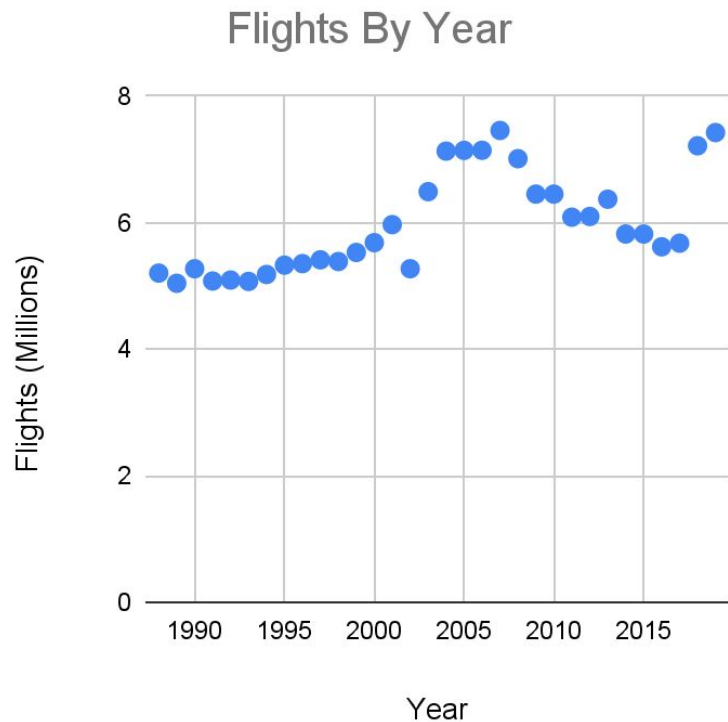
10 rows in set. Elapsed: **5.092 sec.** Processed 196.51 million
rows, 982.57 MB (**38.59 million rows/s., 160.74 MB/s.**)

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

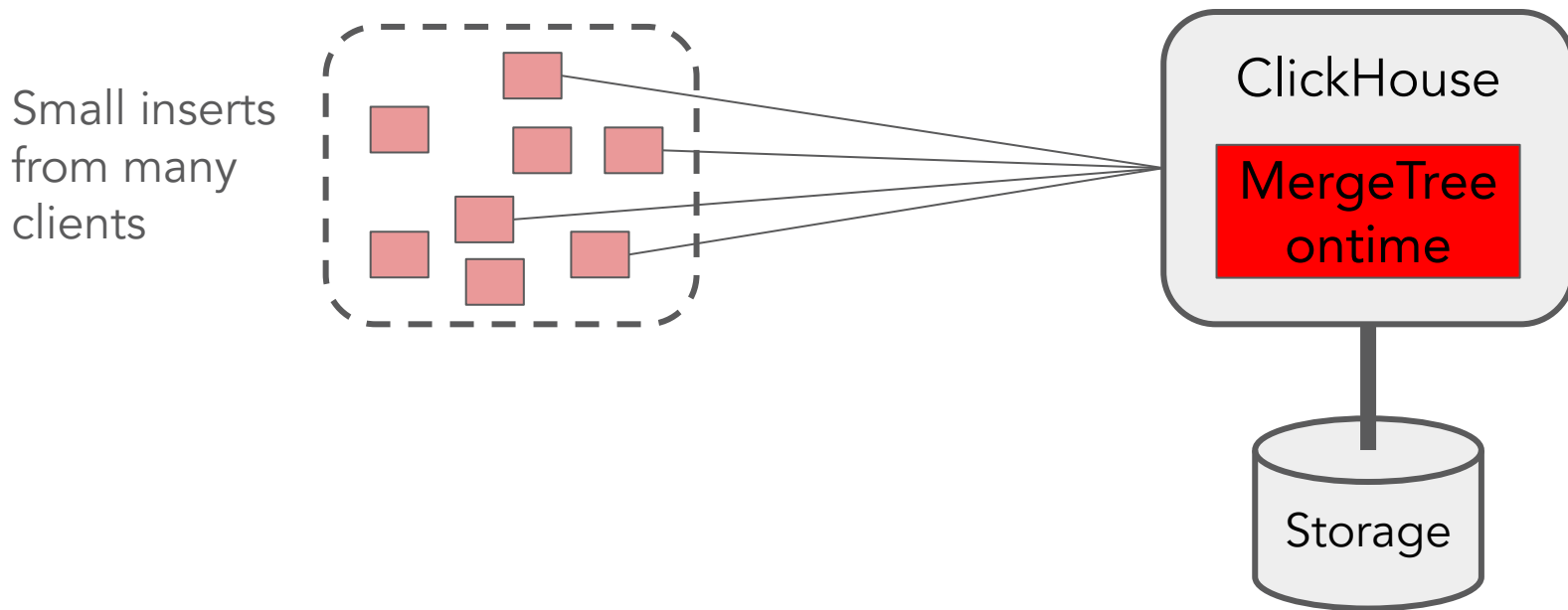
Too many tiny inserts



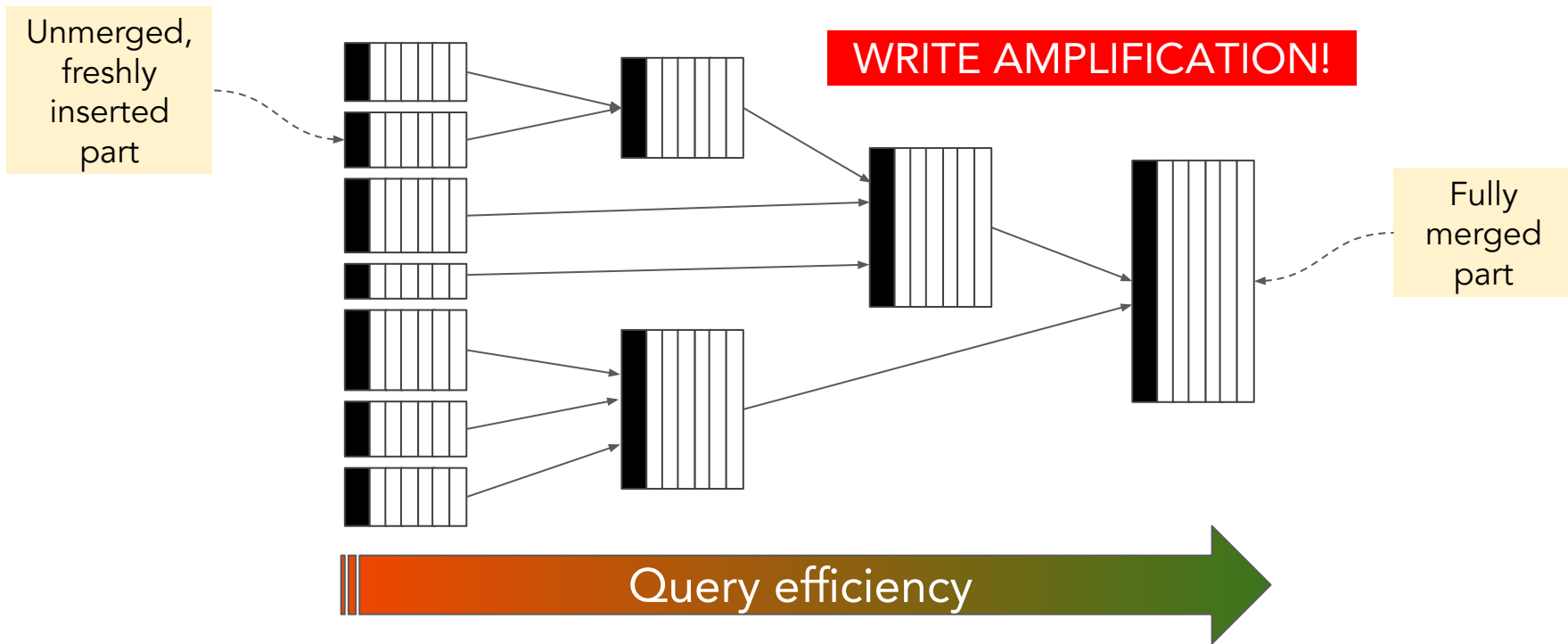
If your inserts look like this, you are doing it wrong!

```
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



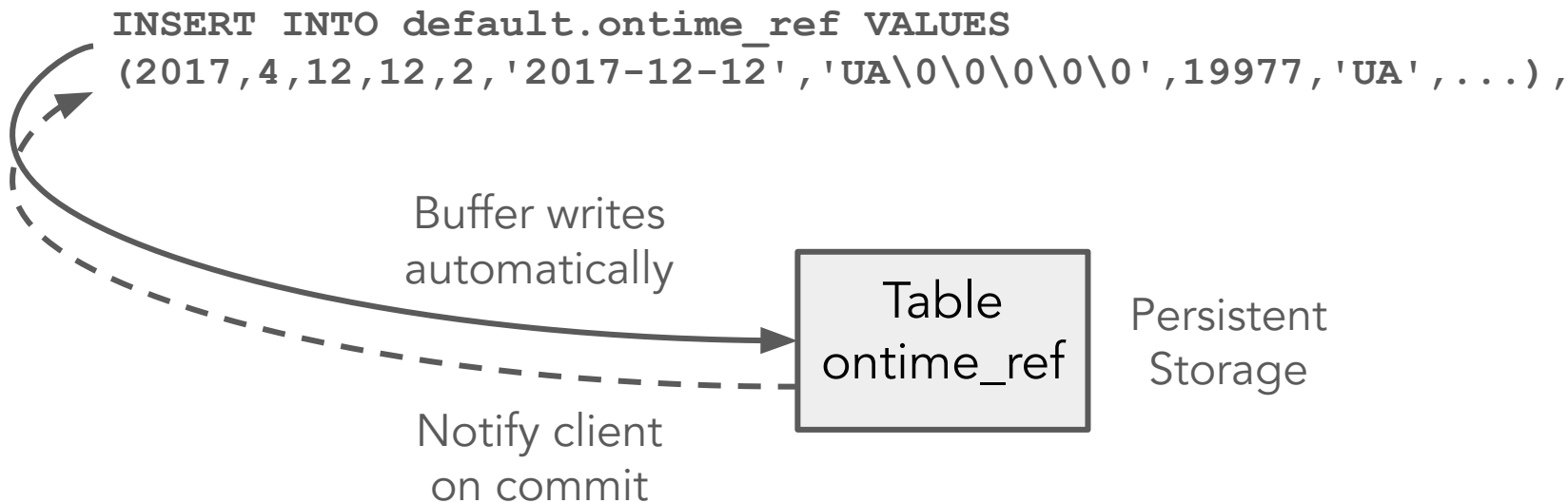
Lots of small parts == slow queries and high merge load



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,  
    wait_for_async_insert=1,  
    async_insert_busy_timeout ms = 10000,  
    async_insert_use_adaptive_busy_timeout = 0
```

```
;
```

```
CREATE USER IF NOT EXISTS async ON CLUSTER '{cluster}'  
    IDENTIFIED WITH sha256_password BY 'topsecret' HOST ANY  
    SETTINGS PROFILE `async_profile`  
;
```

Use async insert
and wait for answer

Wait this long

Don't let
ClickHouse set
automatic values

User with settings

Problem #3

Bad queries

Small differences in queries make big differences in response



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

Simple aggregate, short
GROUP BY key with few values

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

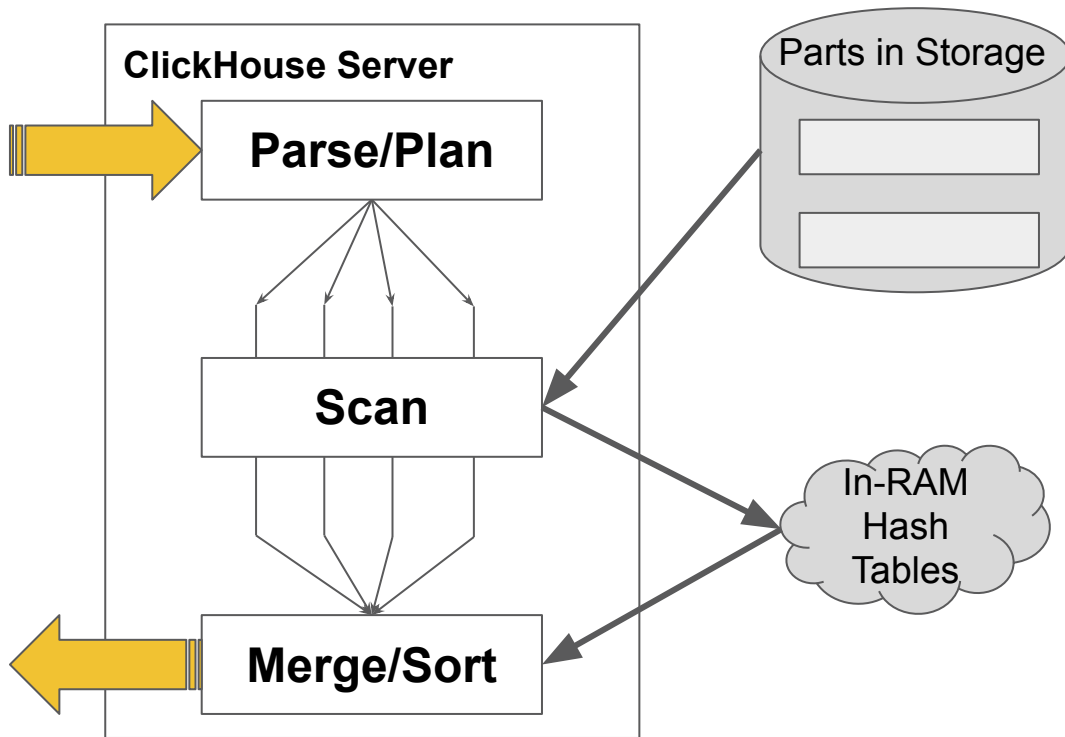
More complex aggregates, longer
GROUP BY with more values

How does ClickHouse process a query with aggregates?

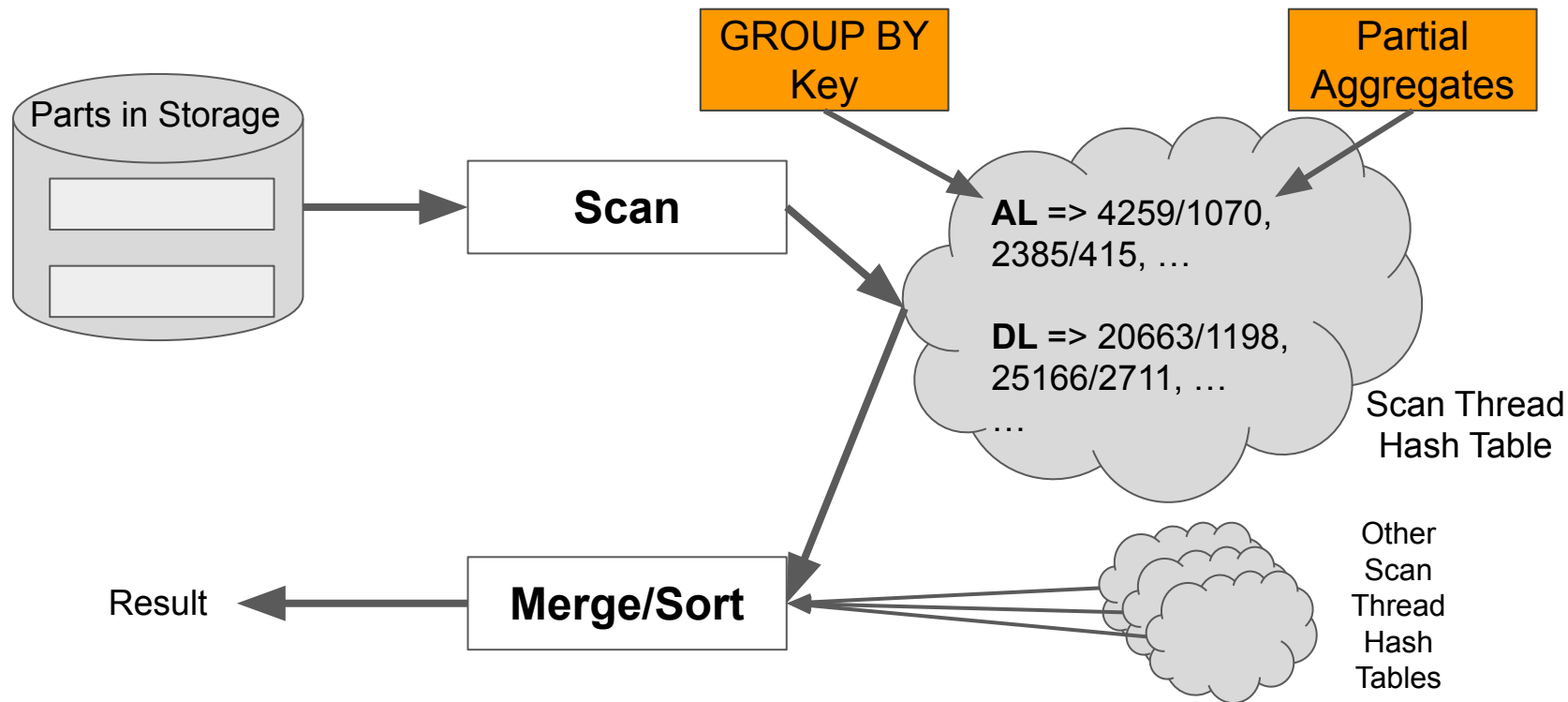
```
SELECT Carrier,  
       avg(DepDelay)AS Delay  
FROM ontime  
GROUP BY Carrier  
ORDER BY Delay DESC
```

Carrier	Delay
B6	12.058290698785067
EV	12.035012037703922
NK	10.437692933474269

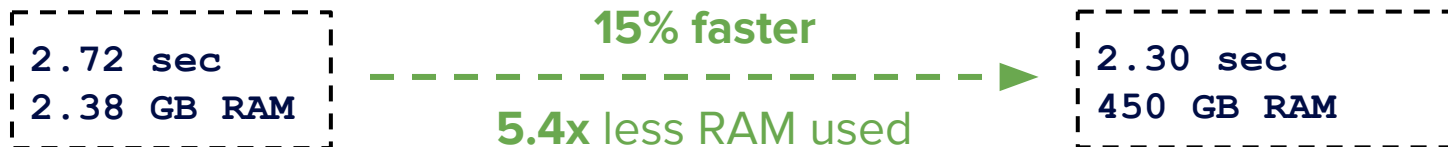
...



How does a ClickHouse thread do aggregation?



We can optimize queries by choosing better aggregates



```
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

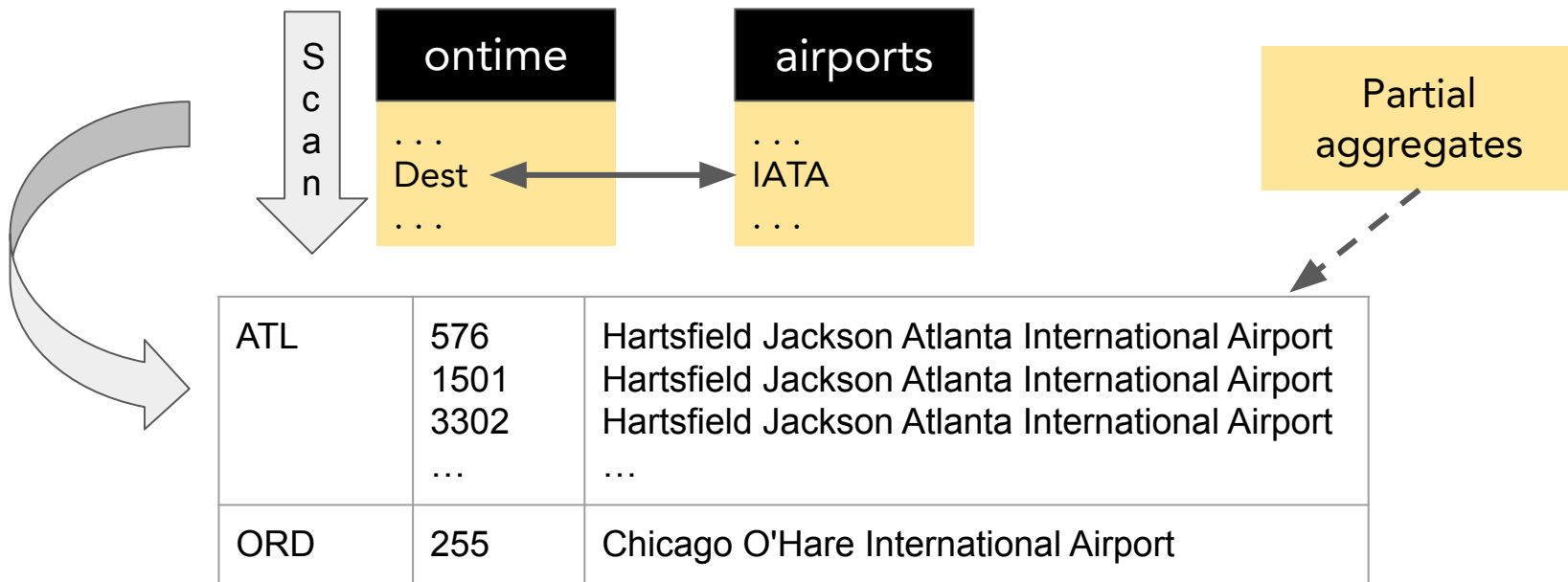
```
SELECT o.Dest, a.Name AS AirportName, o.Flights  
FROM (  
  SELECT Dest, count(Dest) AS Flights  
  FROM ontime_ref GROUP BY Dest ) AS o  
JOIN default.airports a ON a.IATA = o.Dest  
ORDER BY Flights DESC LIMIT 10
```

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]
```

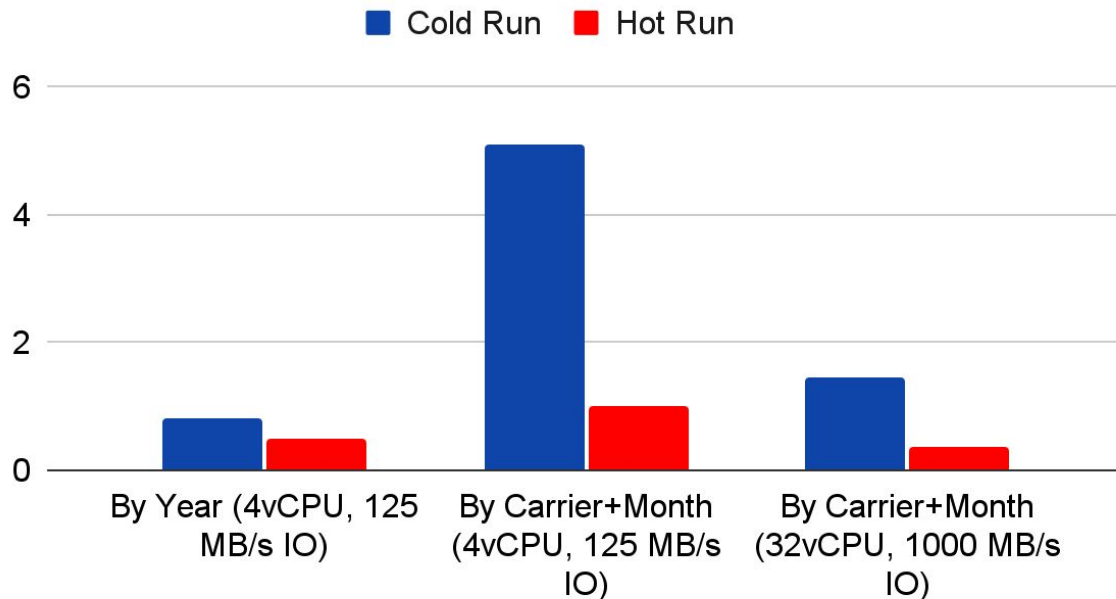
	Carrier	Year	cancelled_pct
1.	G4	2020	16.733186040434276

. . .

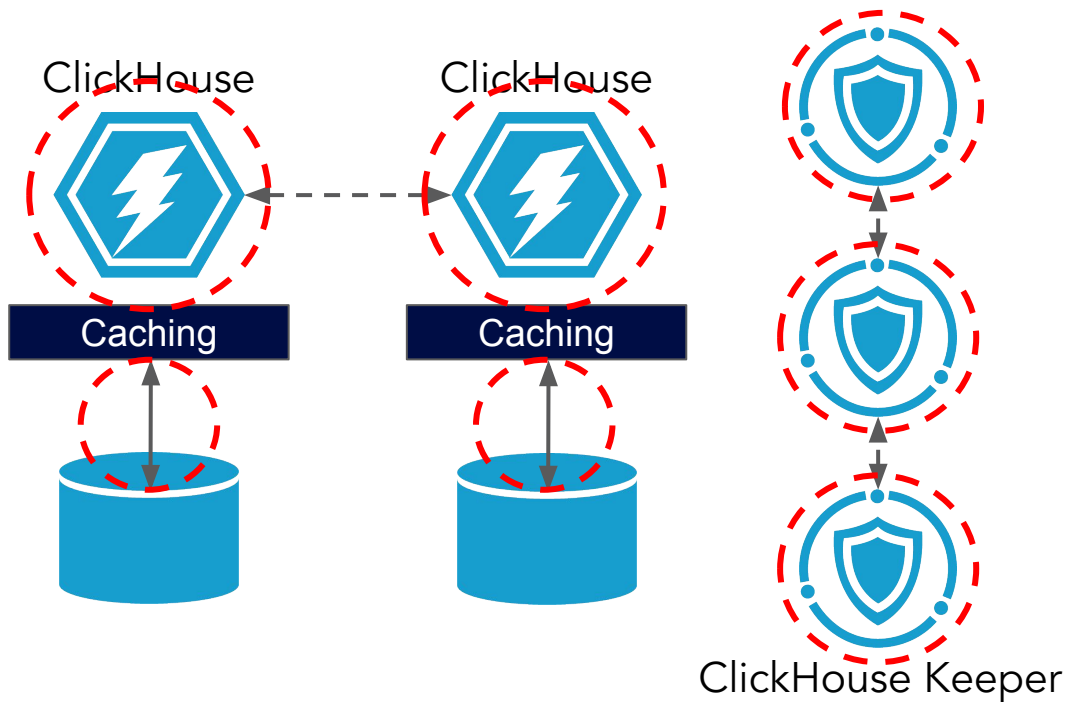
10 rows in set. Elapsed: **5.092 sec.** Processed 196.51 million
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Better hardware can make slow queries fast...

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

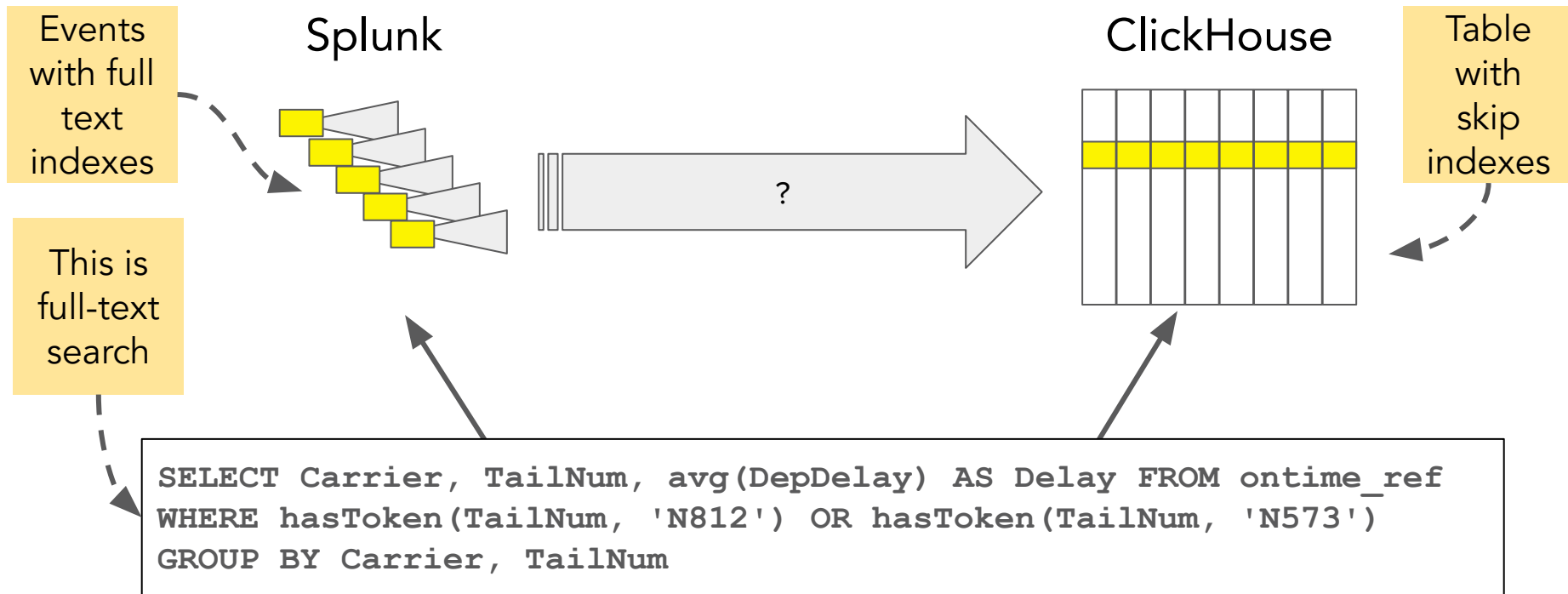
PostgreSQL, MySQL



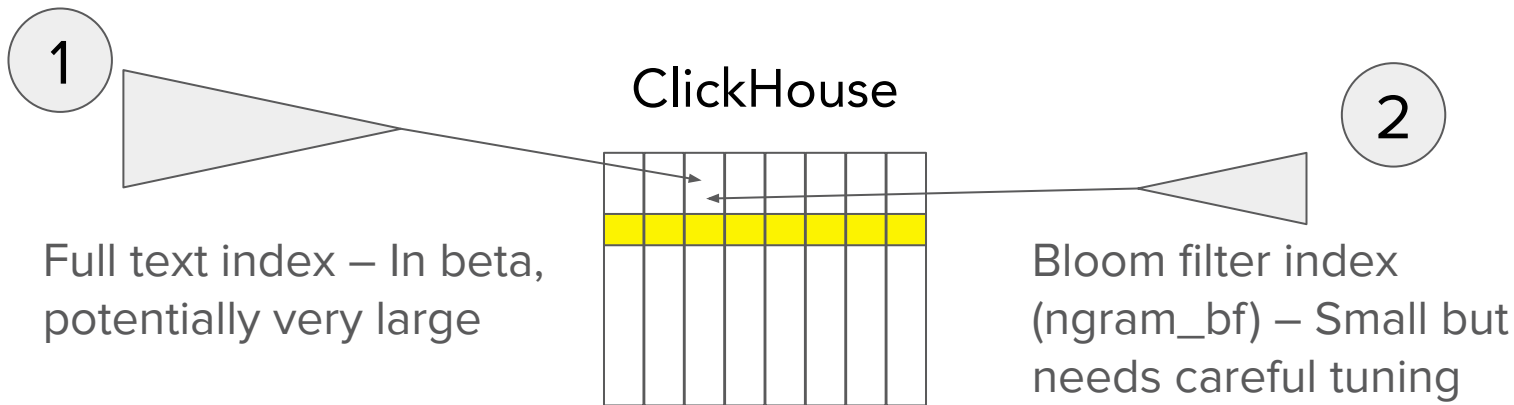
ClickHouse


```
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



```
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  
rhodges_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>  
rhodges_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

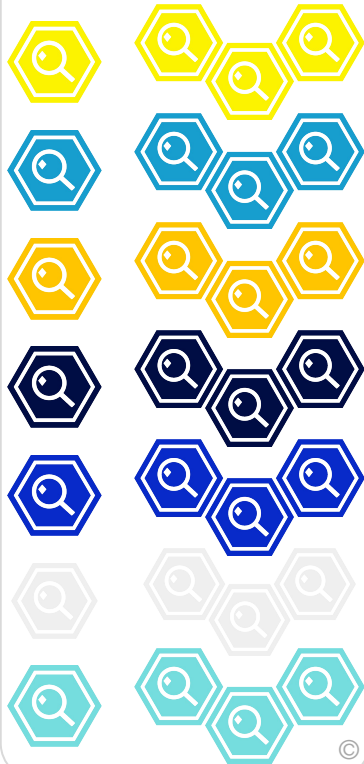


Icons- Transparent

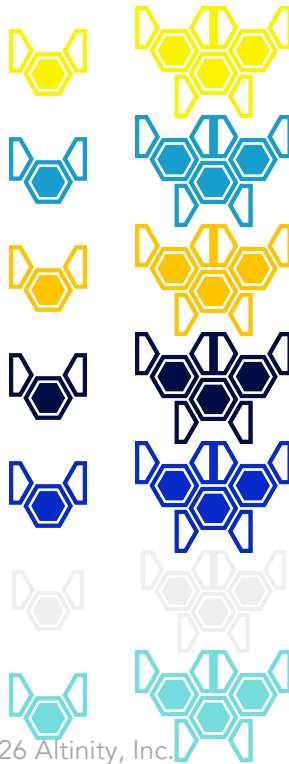
Clickhouse (Native)
Cluster



Director Cluster



Swarm Cluster



Keeper



Other

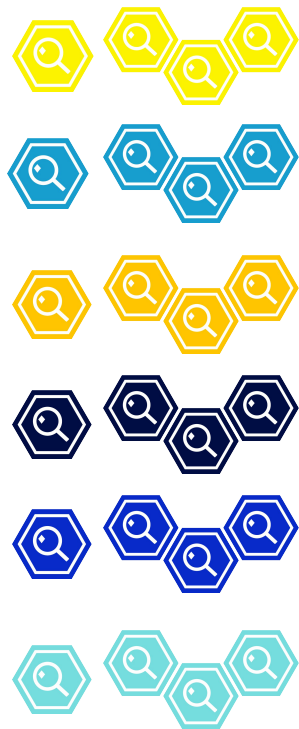


Icons-Stackable on White Backgrounds

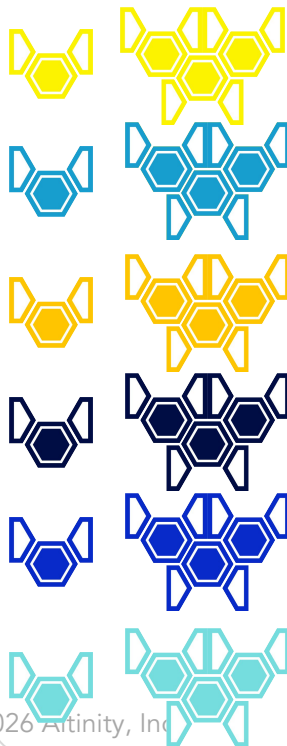
Clickhouse (Native) Cluster



Director Cluster



Swarm Cluster



Keeper



Other



Icons-Stackable on Dark Backgrounds

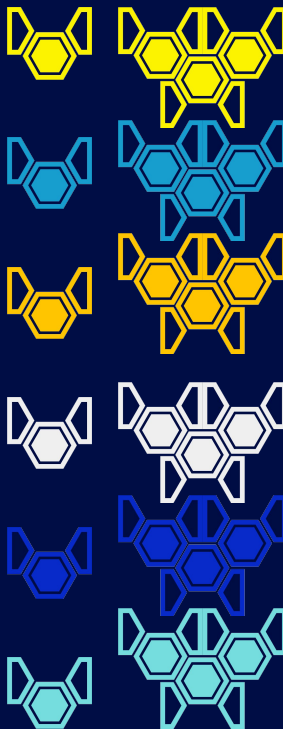
Clickhouse (Native) Cluster



Director Cluster



Swarm Cluster



Keeper



Other

