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Over 10 years of experience in the IT Industry

• Currently, Principal Software Engineer at Electrum.id



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@A_Pangeran

electrum 🔰

We aim to create a sustainable future where electric mobility is accessible, efficient, and integral to urban life. We offer electric motorcycles & mobility solutions with convenient charging battery swap technology.

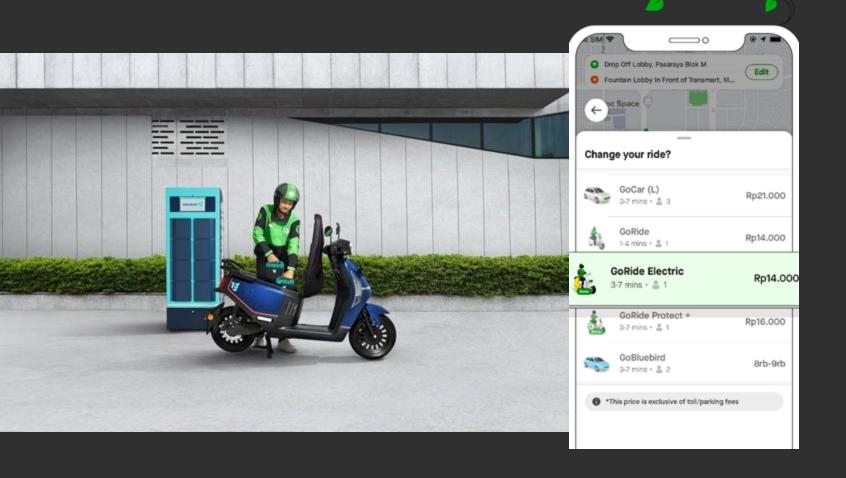
3,000+

electric motorcycles

220k

swap stations across Jakarta km travelled per day

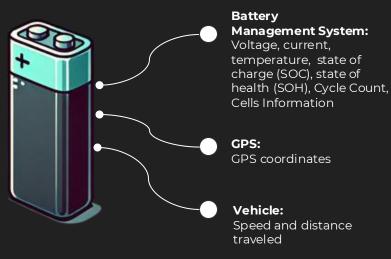




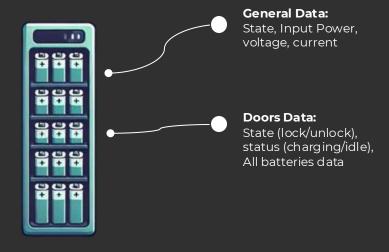




Data: Internet of Things (IoT)

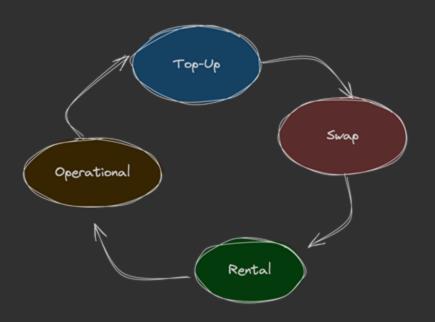


Battery



Swap Station

Data: Transactions (Order)



• Top-Up Transactions:

Data on account recharges, including amounts, payment methods, and status.

Swap Transactions:

Details on battery swaps (station ID, door-number, battery ID, completion status).

Rental Transactions:

- Onboarding: Start data (user, vehicle, battery).
- Offboarding: End data (user, vehicle, vehicle condition).

Operational Activity:

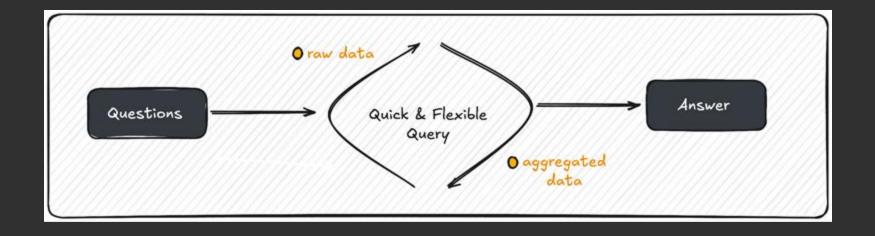
Maintenance, logistics, and fleet management data (e.g., technician actions, station repairs).

So what do we look for?

Flexible, Scaleable, & Low-Cost

Data Stack

Need: Quick & Flexible Query



Question: Why do one of our battery swap stations was charging slower than others?

Answer: Query showed that the station's temperature is 4°C hotter than others, which caused slower charging to avoid overheating. Later, we learned the station was positioned under direct sunlight.

Need: Scaleable & Low-Cost

Resource Efficiency

Efficiently use resources so we're not **over- provisioning**

Expandable Infrastructure

Able to **scale linearly** with resources; as datasets grow, adding more resources will keep response times fast.



Why ClickHouse?







Why: It's columnar based





Excess disk Reads We need data in column, but it's stored in rows. So entire rows are read.



Column-Based

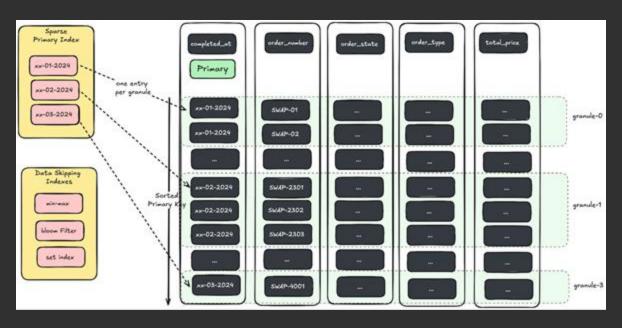
Column Pruning it only pick the subset of columns that are openly declared in the SQL query, thus eliminating 99% columns (990 of 1000 columns) from the disk read consideration.

Vectorized Execution

data processed in **large arrays**, which could be the entire length of the column loaded into RAM.

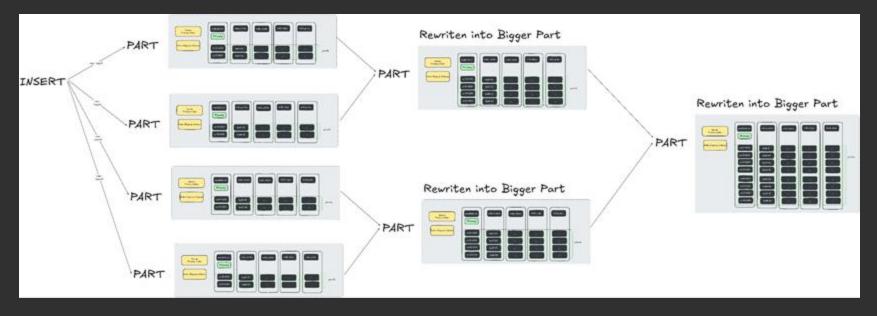
+ reduces CPU cache miss rates

Why: Multiple Layers of Indexes



- Granules are chunks of rows, and ClickHouse groups rows into granules based on the index_granularity setting.
- Sparse Primary Index: ClickHouse uses the <u>sparse primary index</u> to quickly jump to the relevant granules, skipping over large portions of the data that don't need to be read.
- **Skip Indexes**: These allow **ClickHouse** to skip entire granules if it can determine from the index that no rows in the granule satisfy the query's filter.

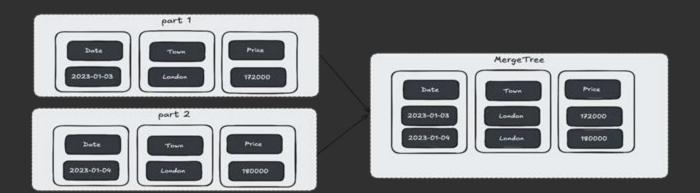
Why: Efficient Merging



Parts: Each time data is inserted, a new part is created. Multiple parts can accumulate over time, which can slow down query performance.

To improve performance and reduce the number of parts, **ClickHouse** periodically runs merging operations.

What Happens During a Merge?

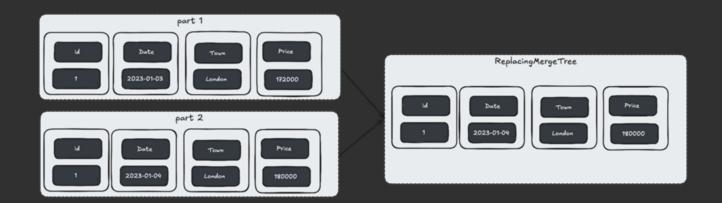


```
CREATE TABLE uk_price_paid
(
   date Date,
   town String,
   price UInt32
)
   ENGINE = MergeTree
   ORDER BY (town, date);
```

MergeTree:

- Combines parts without deduplication or transformations.
- Simply reduces the number of parts but keeps all rows (including duplicates).
- Ideal for append-only data.

What Happens During a Merge?

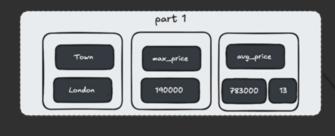


```
CREATE TABLE uk_listings
(
   id UInt32,
   date Date,
   town String,
   price UInt32
)
   ENGINE = ReplacingMergeTree(date)
   ORDER BY id;
```

ReplacingMergeTree:

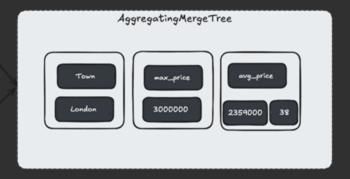
- Combines parts and deduplicates based on a key (or version).
- Replaces older rows with newer ones during a merge.
- Suitable for upserts and scenarios where data is frequently updated.

What Happens During a Merge?





```
CREATE TABLE uk_price_paid_aggregates
(
   town String,
   max_price SimpleAggregateFunction(max, UInt32),
   avg_price AggregateFunction(avg, UInt32)
)
ENGINE = AggregatingMergeTree
   ORDER BY town;
```



AggregatingMergeTree:

- Combines parts by aggregating rows with the same key using predefined functions like sum, count, avg, etc.
- Ideal for pre-aggregating data to reduce query time on large datasets.

Another thing we love about ClickHouse is **materialized view.**

Let's deep-dive

Materialized view 101

```
CREATE TABLE IF NOT EXISTS trx_orders
(
   completed_at DateTime64(3, 'UTC'),
   order_number String,
   order_type LowCardinality(String),
   total_price Int64
)
ENGINE = MergeTree
    PARTITION BY toYYYYMM(completed_at)
   ORDER BY (completed_at, order_type, order_number)
   SETTINGS index granularity = 8192;
```

Raw-Table

Aggregated-Table

SYNC?

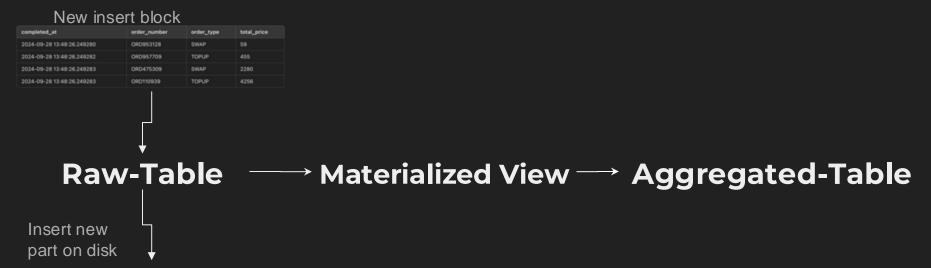
Materialized view 101

```
CREATE TABLE IF NOT EXISTS trx_orders
(
   completed_at DateTime64(3, 'UTC'),
   order_number String,
   order_type   LowCardinality(String),
   total_price Int64
)
ENGINE = MergeTree
   PARTITION BY toYYYYMM(completed_at)
   ORDER BY (completed_at, order_type, order_number)
   SETTINGS index_granularity = 8192;
```

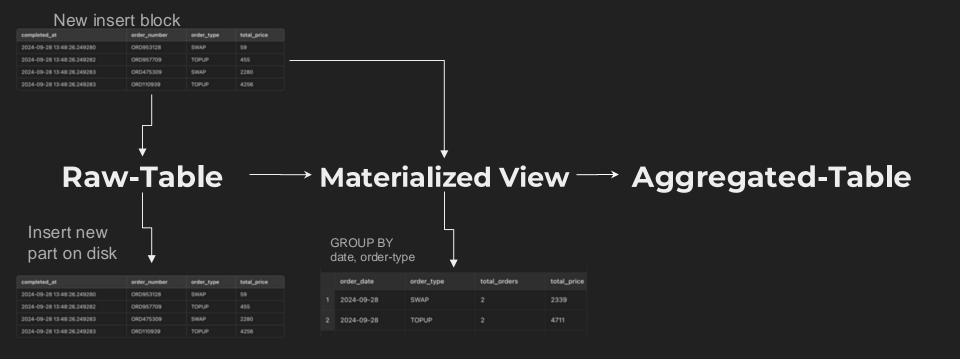
Raw-Table → Materialized View → Aggregated-Table

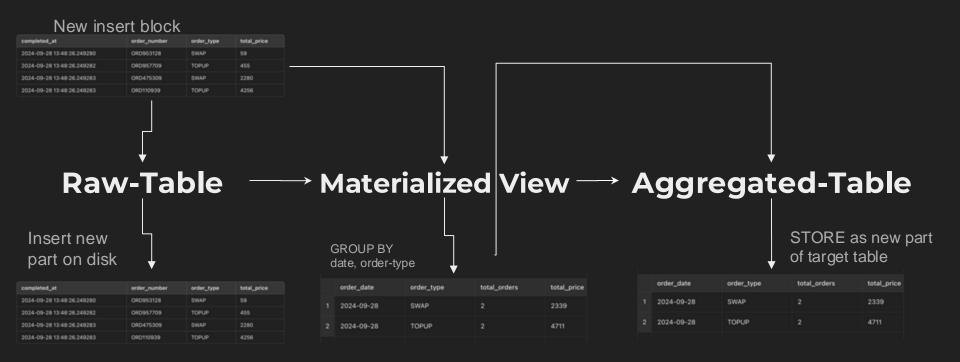


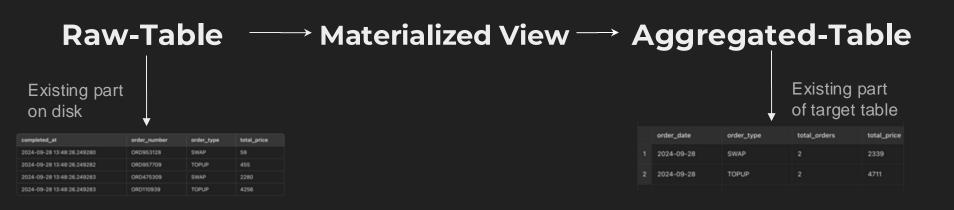
Raw-Table → Materialized View → Aggregated-Table

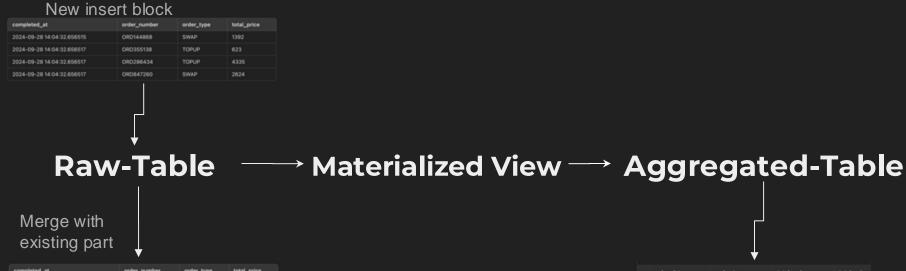


completed_at	order_number	order_type	total_price
2024-09-28 13:48:26:249280	ORD953128	SMAP	
2024-09-28 13-48-26.249282	ORD957709	TOPUP	455
2024-09-28 13:48:26:249283	ORD475309	SMAP	2280
2024-09-28 13:48:26.249283	ORD110939	TOPUP	4256





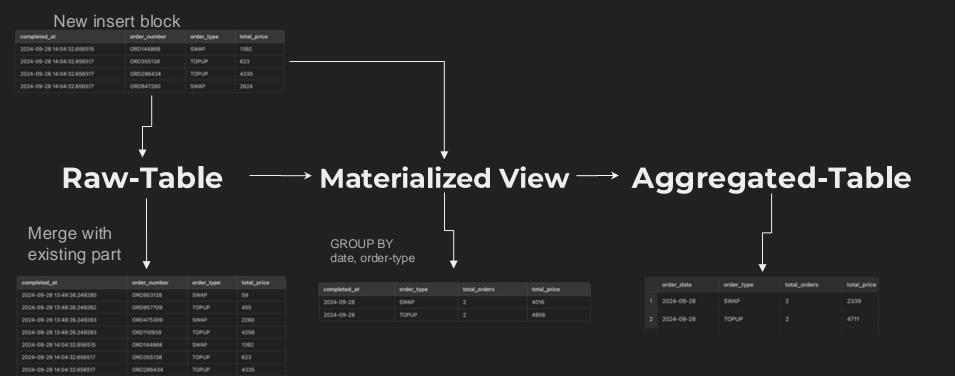




2339

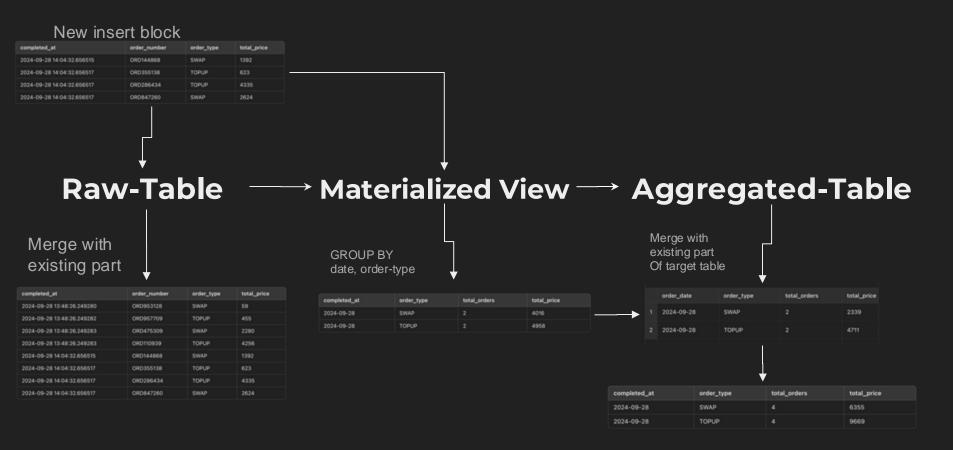
1 2024-09-28

completed_at	order_number	order_type	total_price
2024-09-28 13:48:26:249280	ORD953128	SINAP	
2024-09-28 13:48:26.249282	ORD957709	TOPUP	455
2024-09-28 13:48:26.249283	ORD475309	SWAP	2280
2024-09-28 13:48:26.249283	ORD110939	TOPUP	4256
2024-09-28 14:04:32.656515	ORD144868	SINAP	1392
2024-09-28 14:04:32.656517	ORD355138	TOPUP	623
2024-09-28 14:04:32.656517	ORD286434	TOPUP	4335
2024-09-28 14:04:32:656517	ORD847260	SWAP	2624



2024-09-28 14:04:32:656517

ORD847260



Materialized view 101 - Query Comparison

Raw-Table → Materialized View → Aggregated-Table

completed_at	order_number	order_type	total_price
2024-09-28 13:48:26.249280	ORD953128	SWAP	
2024-09-28 13:48:26.249282	ORD957709	TOPUP	455
2024-09-28 13:48:26.249283	ORD475309	SWAP	2280
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2024-09-28 14:04:32.656515	ORD144868	SWAP	1392
2024-09-28 14:04:32:656517	ORD355138	TOPUP	623
2024-09-28 14:04:32.656517	ORD286434	TOPUP	4335
2024-09-28 14:04:32:656517	ORD847260	SWAP	2624

Processed 8 rows

```
        completed_at
        corper_type
        total_price

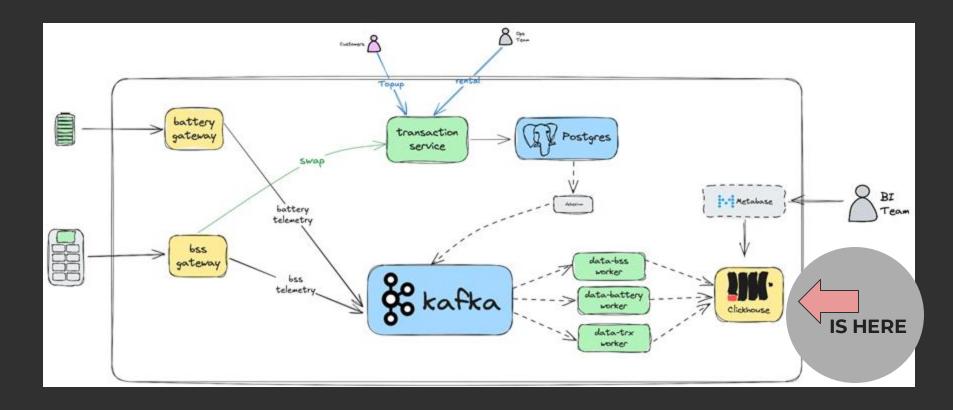
        2024-09-28
        SWAP
        4
        6355

        2024-09-28
        TOPUP
        4
        9669
```

Processed 2 rows

How ClickHouse Powers Our Data Platform

How ClickHouse Powers Our Data Platform



DEMO

https://github.com/meong1234/clickhouse-meetup-10-2024

Where are we?





"ClickHouse enables us to have a data analytics capability that grows with our needs, while keeping costs low and predictable" - Electrum.id

Key Takeaways

Columnar Format:

ClickHouse stores data in **columns**, optimizing read performance for analytical queries by reading only the relevant columns, reducing I/O and improving query speed.

Sparse Primary Index:

ClickHouse uses a sparse index to store pointers to data blocks (granules), allowing it to skip irrelevant data during queries, improving performance when filtering by the primary key.

Skip Index:

The **skip index** allows **ClickHouse** to **bypass granules** based on non-primary key column values, helping to avoid unnecessary data scanning and speeding up queries on non-primary key columns.

Merging Process in Different Engines:

- MergeTree: Performs basic merging of data parts without deduplication or aggregation.
- ReplacingMergeTree: Deduplicates data during merging, keeping only the most recent version of rows with the same primary key.
- **Aggregating Merge Tree**: Aggregates data during merging based on predefined aggregation functions, storing the results in summarized form.

ClickHouse Materialized Views:

- Triggered on Inserts: Automatically updates the materialized view whenever new data is inserted into the source table.
- **Real-Time Pre-Aggregation**: Performs pre-aggregation or data transformation in real time, allowing for faster query performance by reducing the need for complex computations at query time.

Ref

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ChistaDATA Inc. (2023). Why ClickHouse Is So Fast. https://chistadata.com/why-clickhouse-is-so-fast

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