Storage Internals & Micro-Partitions

What are Micro-partitions?

Industry Problem Modern data-driven enterprises like Netflix, Amazon, or Meta generate petabytes of data every day—user logs, transactions, content interactions, personalization data, etc.

Traditional row-based and even columnar databases often struggle with:

- Managing massive volumes of data efficiently
- Fast query response on large tables (billion+ rows)
- Poor partitioning logic needing manual tuning
- Expensive storage scanning due to full-table scans

Snowflake's Solution: Micro-partitions

Snowflake **breaks down every table into hundreds to millions of **immutable**, **columnar**, **compressed** files called **micro-partitions**.

Key Characteristics

Feature	Description		
Size	Each micro-partition stores ~16 MB (compressed) of data		
Columnar	Each column is stored independently, enabling selective reads		
Automatic	You don't define partitions—Snowflake handles it automatically		
Metadata-rich	Each micro-partition stores min/max ranges, null counts, and value lists per column		
Immutable	Once written, a micro-partition doesn't change—new changes create new partitions		

How it Works (Behind the Scenes)

- When you insert 10 million rows into a table, Snowflake:
 - Splits it into many micro-partitions
 - Compresses and organizes them by column
 - Stores min/max and statistical metadata
 - Makes them available for pruning

Real Industry Example – Netflix (Content Analytics)

Netflix collects **video watch logs** from every device (TVs, phones, browsers) for personalization, recommendation, A/B testing, etc.

Each event:

```
user_id, video_id, device_type, play_time, watch_duration, region, timestamp
```

They use Snowflake to:

- Store billions of rows per day
- Organize them automatically into micro-partitions
- Enable filtering by region/date/user without full scans

So, a query like:

```
SELECT COUNT(*)
FROM VIDEO_LOGS
WHERE REGION = 'IN' AND DATE(timestamp) >= CURRENT_DATE - 1;
```

Only scans partitions that match the date and region (instead of the entire table)

Columnar Storage Format & Compression

Industry Problem

At scale (think Google Search logs, Amazon transaction data, or Meta ad clicks):

- Reading entire rows (all columns) is extremely inefficient
- Uncompressed storage eats up huge cost and IO
- Legacy systems read 90% irrelevant data

Snowflake's Solution: Columnar + Compression

Snowflake's storage engine is:

- Fully columnar: Data is stored by column, not row
- Heavily compressed: Applies best-fit compression per column
 - o Run-Length Encoding, Dictionary, Delta, Bit-Packing, etc.

Benefits:

- Only columns used in SELECT are read from disk
- Compression ratios of 5x to 15x are common
- Read and write optimized for analytical queries

Example: How It Helps Meta (Facebook Ads Team)

Meta stores ad impressions and clicks with fields:

```
user_id, ad_id, campaign_id, click_time, device_type, impression_cost
```

A marketing analyst querying **daily click-through rates** only needs:

```
SELECT campaign_id, COUNT(*)
FROM ADS_EVENTS
WHERE click_time BETWEEN '2024-01-01' AND '2024-01-02'
GROUP BY campaign_id;
```

Snowflake only reads:

- campaign_id
- click_time

Result: Massive reduction in I/O and query latency despite billions of rows

Automatic Creation & Immutability of Micro-partitions

Industry Problem / Real-World Challenge

In legacy warehouse solutions (e.g., Hadoop Hive or on-prem Teradata):

- You must predefine partitioning keys
- Poor choice leads to query inefficiency
- Data mutations rewrite entire partitions
- Concurrency and time-travel are hard

Snowflake's Solution: Auto & Immutable Micro-partitions

* Automatic Creation:

- Snowflake automatically partitions data as it's loaded
- Based on **natural ingestion order** (no user intervention)
- Adds metadata to aid later pruning

📌 Immutability:

- Once a micro-partition is written, it is never updated
- Updates/deletes result in new partitions
- Ensures:
 - Fast concurrency
 - Snapshot isolation (Time Travel)

Audit safety and compliance

Real-World Example: Amazon Order Data

In Amazon's ecommerce systems:

Every order has: order_id, customer_id, order_date, order_amount, status

Their BI systems might:

- Load millions of orders hourly
- Occasionally correct order statuses or refunds

Snowflake allows:

- Automatic ingestion → partitions created
- Refund updates? → Old partitions closed, new partition created
- Analysts still see **latest view**, but older partitions exist for:
 - Time Travel
 - o Audit trails
 - Reversibility

Example: Observing Partition Evolution

```
-- View table metadata

SELECT *

FROM INFORMATION_SCHEMA.TABLE_STORAGE_METRICS

WHERE TABLE_NAME = 'ORDERS';

-- View clustering statistics

SELECT SYSTEM$CLUSTERING_INFORMATION('ORDERS');
```

- You'll observe micro-partition counts and last update times changing
- But no partition is modified—new data = new partitions



Subtopic	Key Idea	Benefit	Industry Example
What are Micro-partitions?	Snowflake breaks large data into small, columnar blocks	Enables automatic scalability and parallelism	Netflix watch logs
Columnar + Compression	Store columns separately with compression	Only required data read; faster queries, less storage	Meta ad clicks
Auto + Immutable Partitions	Snowflake handles everything + never updates partitions	Enables time travel, ACID compliance, and high concurrency	Amazon order pipeline