



Low latency Change Data Capture (CDC) to your data lake, using Apache Flink and Apache Paimon

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Agenda

Change Data Capture (CDC) use-cases

Challenges when handling CDC

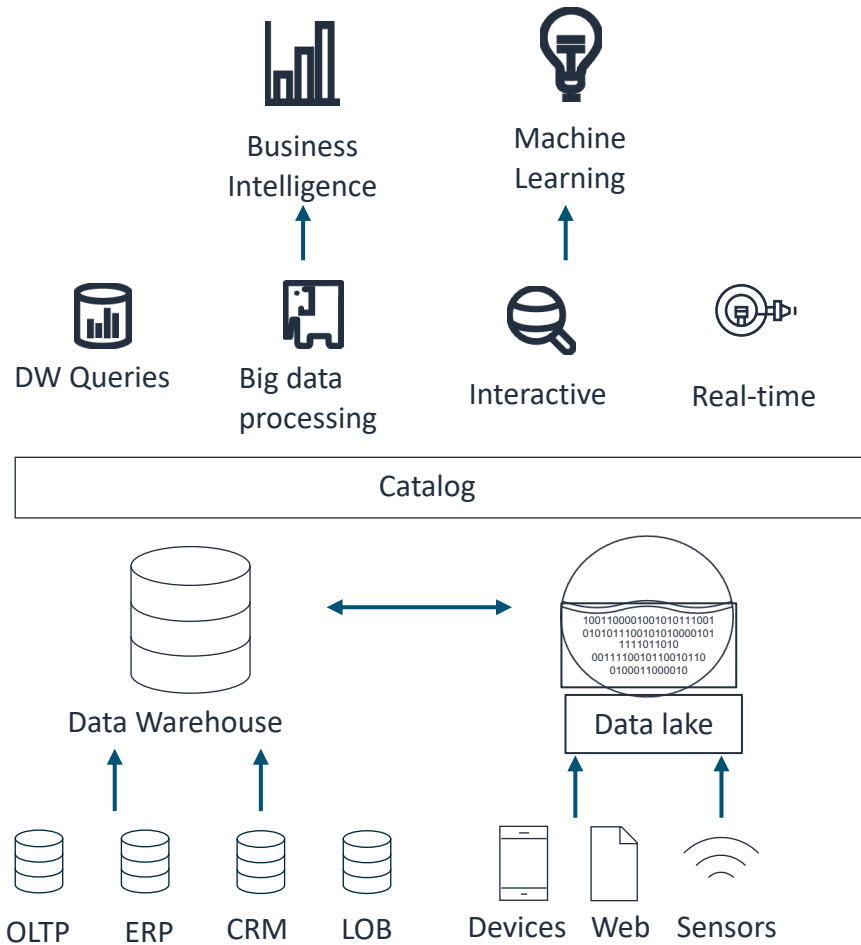
Apache Flink CDC and Apache Paimon

How Apache Paimon solves data pipeline challenges?

Conclusion



Data Lakehouse



Data lakes provide:

Relational and non-relational data

Scale-out to EBs

Diverse set of analytics and machine learning tools

Work on data without any data movement

Designed for low cost storage and analytics

CDC to a message queue



Challenges when handling CDC data



Data volume



Complexity



Data Quality



Integrating CDC data with other Data



Consistency



Data security and compliance

Why Apache Flink for CDC?



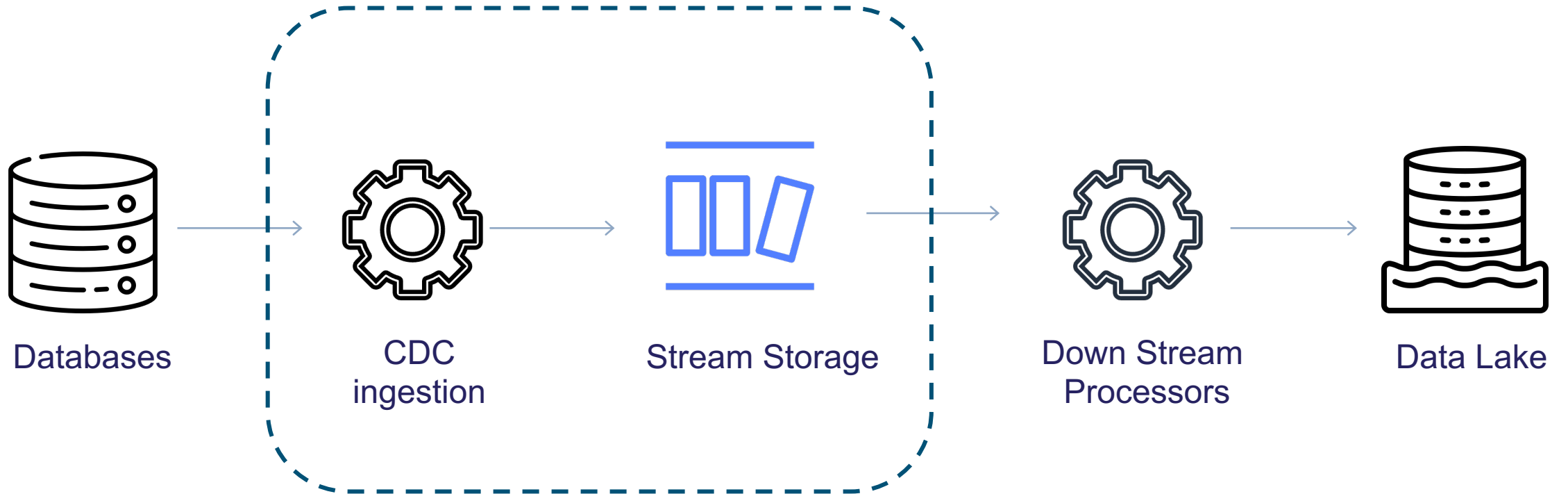
Apache Flink

- Can consume changelog from various sources across variety of database engines: (MySQL, PostgreSQL, SQL Server, and more)
- Reading snapshots, and transaction logs
- Provides Dynamic Table as one abstract for:
 - Temporal processing
 - CDC data operations (Insert, Update, Delete)
 - Bounded Unbounded
- Exactly-once semantics
- Data delivery to various targets, including file system (including: Data lake)

CDC Patterns



CDC to a message queue



CDC to Apache Kafka topic



1. `CREATE TABLE source_table... AS (...)`
2. `CREATE TABLE target_table...AS (...)`
3. `INSERT INTO target_table SELECT * from`
`source_table;`

Pros –

- Low Latency
- Horizontal scalability with Kafka

Challenges –

- Schema Evolution
- Kafka Storage
- Kafka capacity during backfill
- Retention

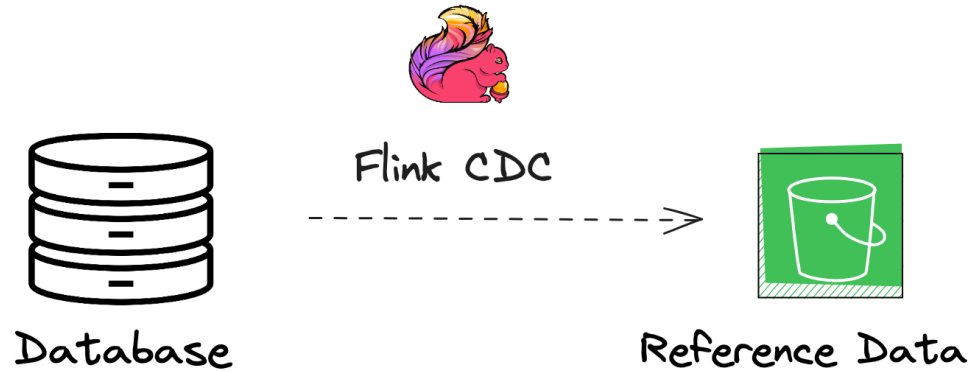


CDC to Apache Kafka topic



- New snapshot when Flink state is lost
- Data inconsistency between source, and Kafka
- Standard topic deletes data after retention
- Compacted topics keeps data forever
- Spike in storage during the backfill
- Separate topic per each table

More cost effective, scalable approach



1. `CREATE TABLE source_table... AS (...)`
2. `CREATE TABLE target_table...AS (...)`
3. `INSERT INTO target_table SELECT * from source_table;`

Pros –

- Scalable storage
- Reduce cost
- Infinite retention




Challenges -

- Higher Latency compared to Kafka
- UPSERT / Delete
- Streaming read
- Schema evolution

Open Table Formats



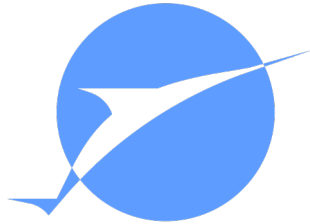
Modern Data lakes use Open Table Format

<i>Metadata</i>	Hive MetaStore				
<i>Computing</i>	Spark	Flink	Trino	Athena	Redshift
Table Format			ICEBERG 		
File format	CSV	JSON	Parquet	Avro	ORC
<i>Storage</i>	HDFS		Amazon S3		

CDC to data lake using Apache Flink



```
INSERT INTO "target_table"  
SELECT * from "source_table";
```



```
INSERT INTO "target_table"  
SELECT * from "source_table";
```



```
INSERT INTO "target_table"  
SELECT * from "source_table"  
/*+ OPTIONS('upsert-enabled'='true') */
```

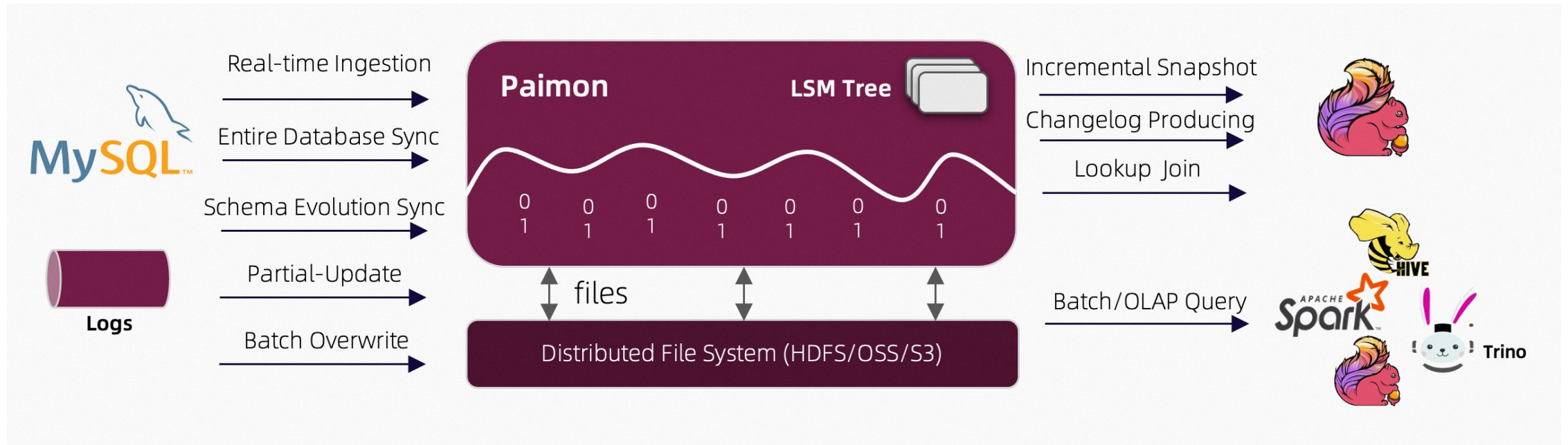

What is Apache Paimon ?



Apache Paimon



Apache Paimon is a streaming data lake platform that supports **high-speed data ingestion**, **change data tracking** and **efficient real-time analytics**.

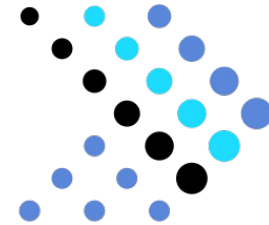
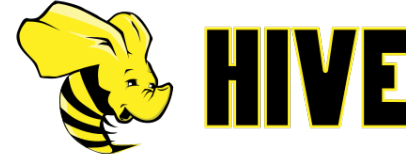


Apache Paimon supports...

File Systems



Engines



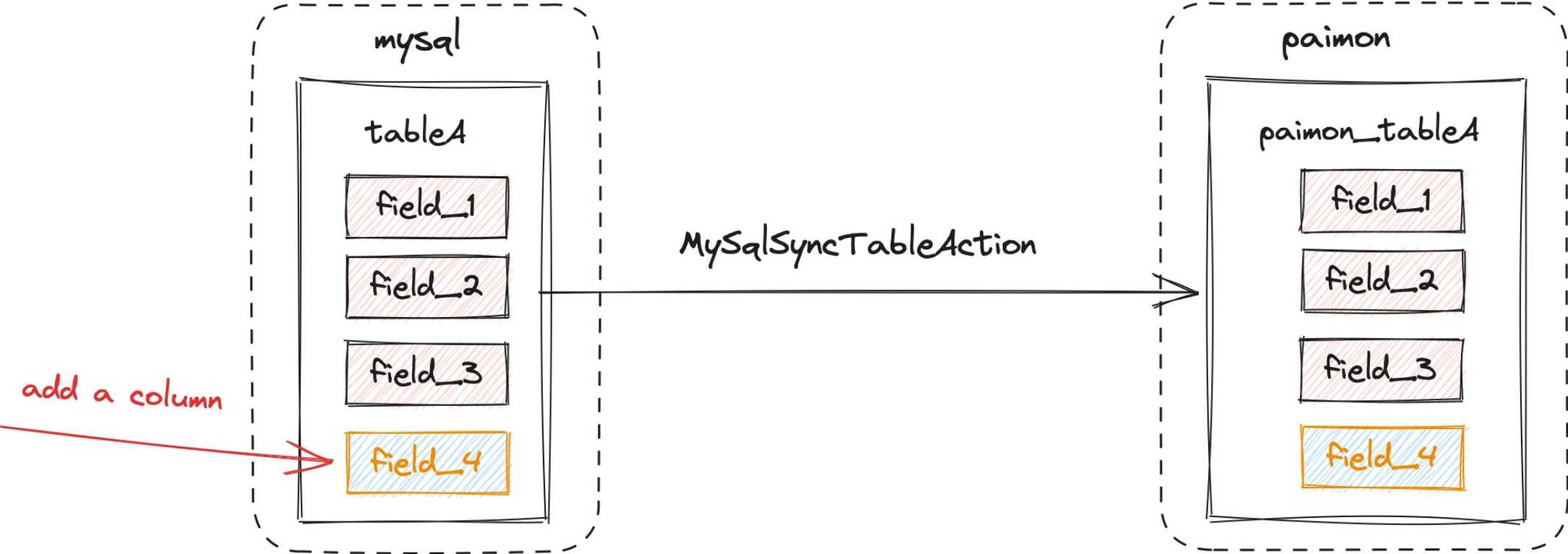
trino



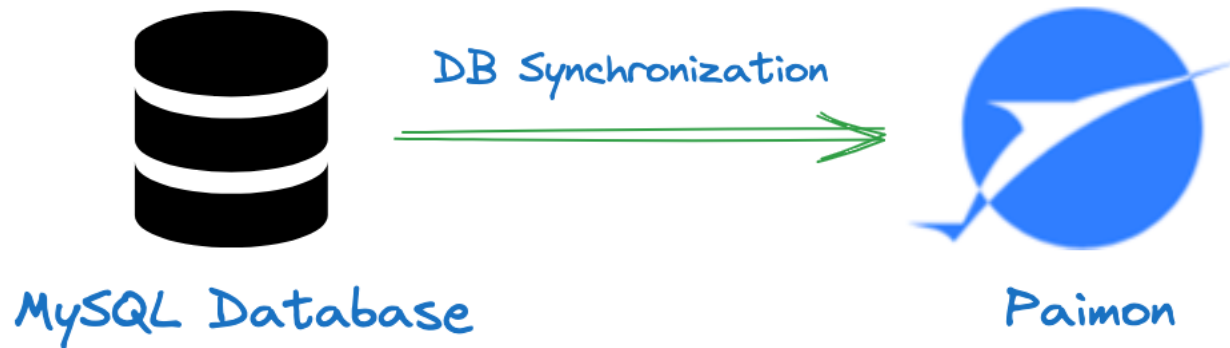
How do we handle schema changes ?



What is schema evolution?



Apache Paimon Actions



- Synchronisation from **MySQL, Kafka, Pulsar** and **MongoDB**
- Supports **table/collection/database** synchronization
- **One Flink Job** sinks all table
- Support **Schema Evolution**
- New tables are created and synced **automatically**
- **Supports UPSERT** for PK table
- Low cost for large number of small tables

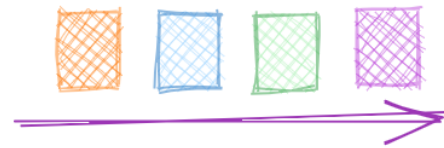
CDC requires small file writes,
random file reads, and many re-
writes



CDC write to partitions



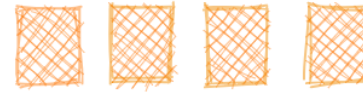
Orders



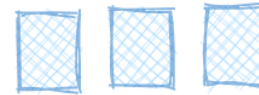
DB Synchronisation



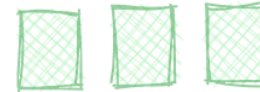
partition=20221215



partition=20221214



partition=20221213

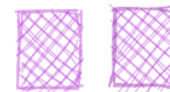


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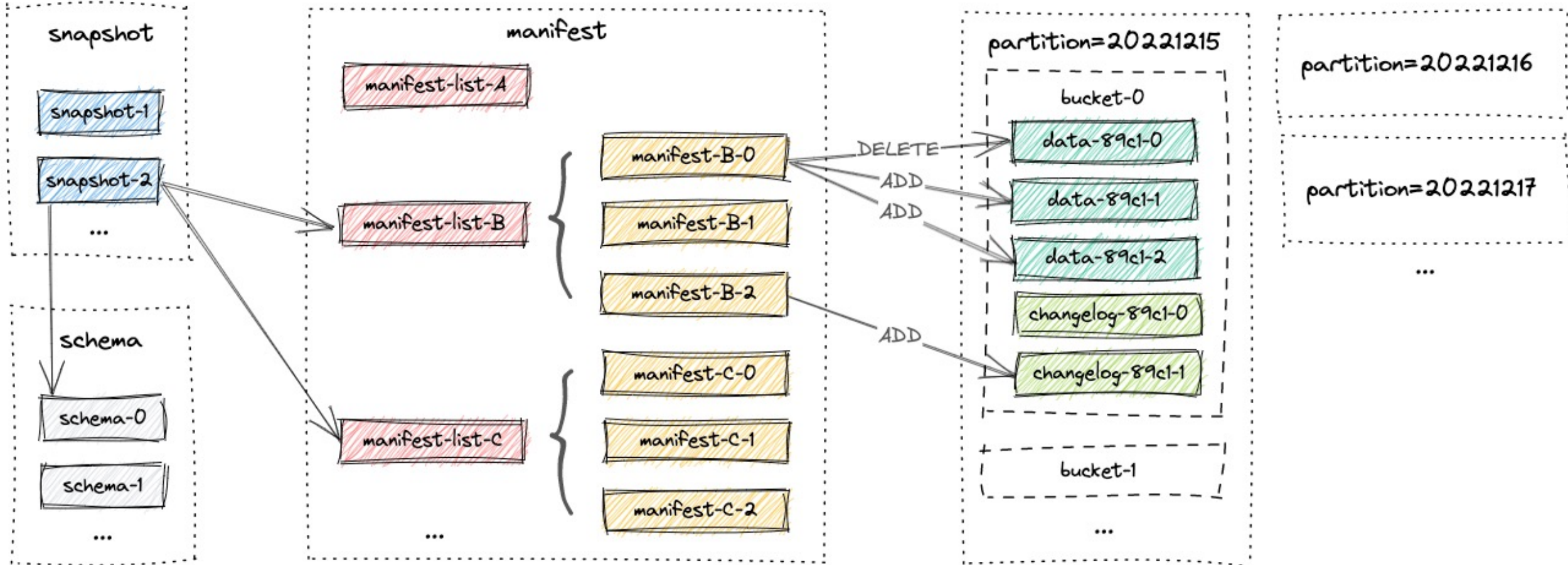
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partition=20231231



Apache Paimon File Layouts

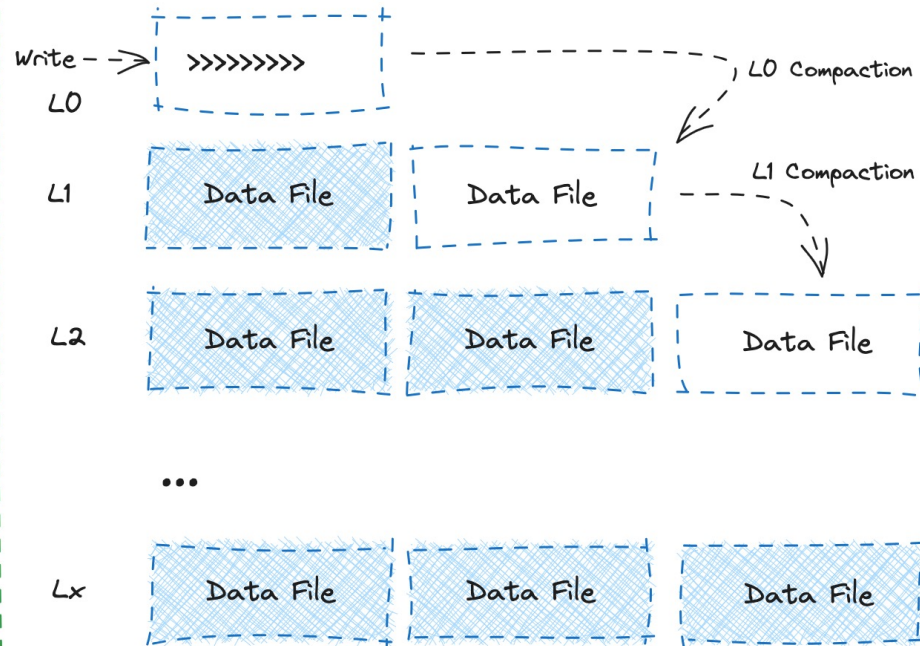


<https://paimon.apache.org/docs/master/concepts/basic-concepts/>

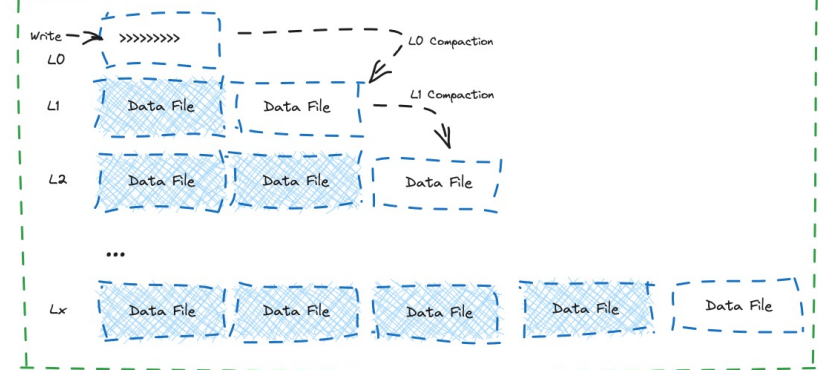
Partition and Bucket

partition=20221215

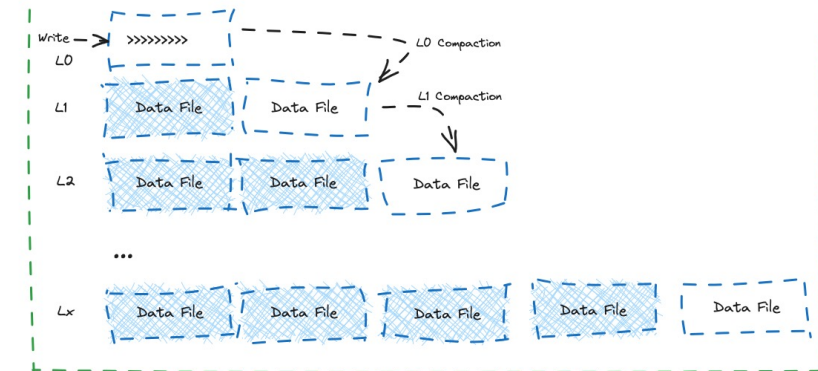
bucket-0



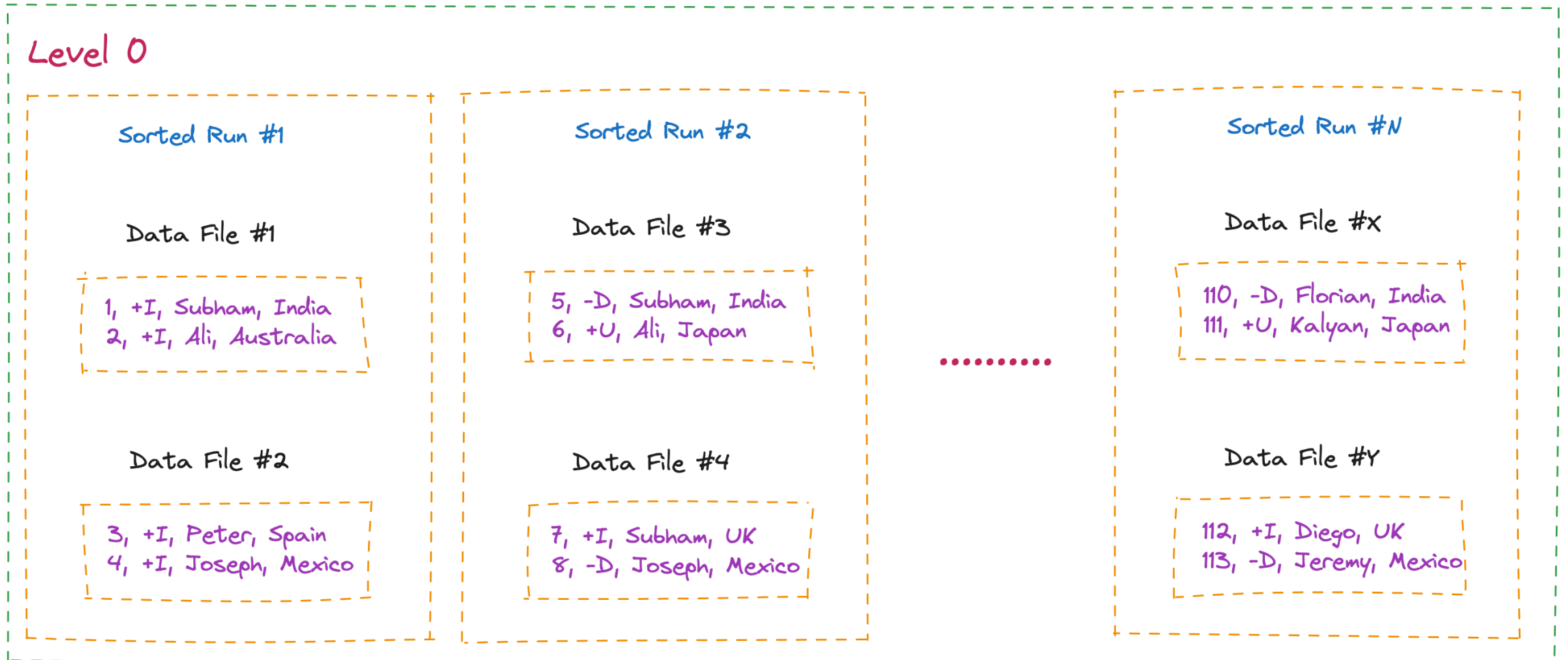
bucket-1



bucket-2

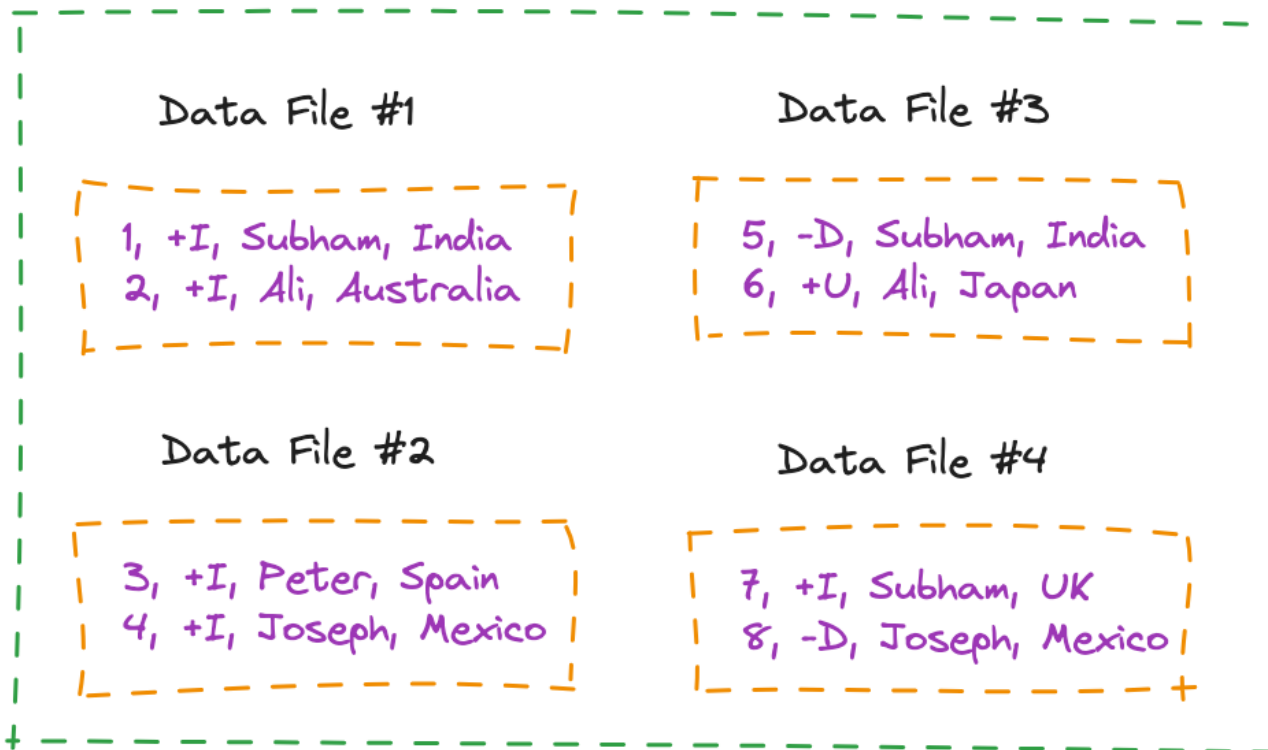


Level 0 Sorted Runs

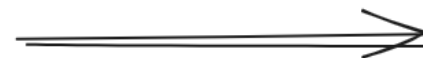


Compaction

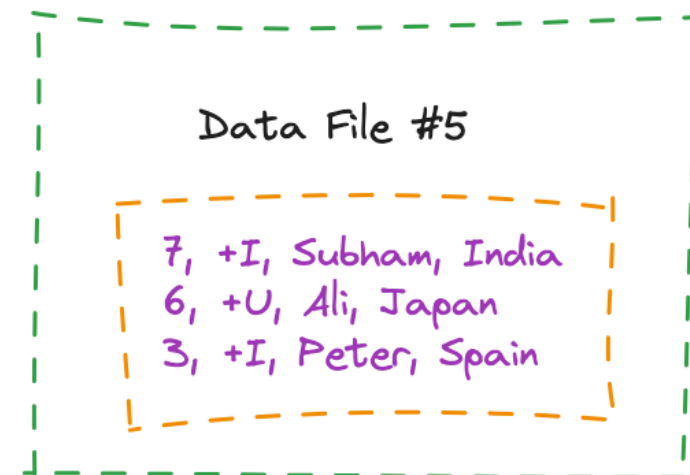
Level 0



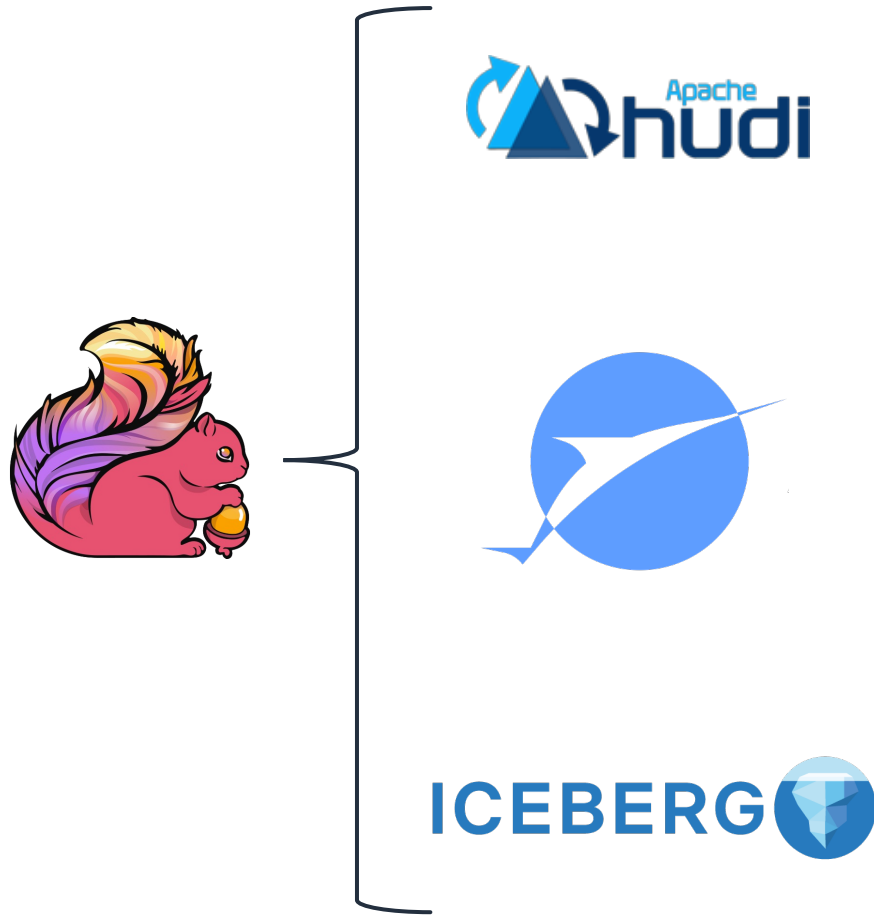
Merge datafiles



Level 1



Compaction



Same job compaction

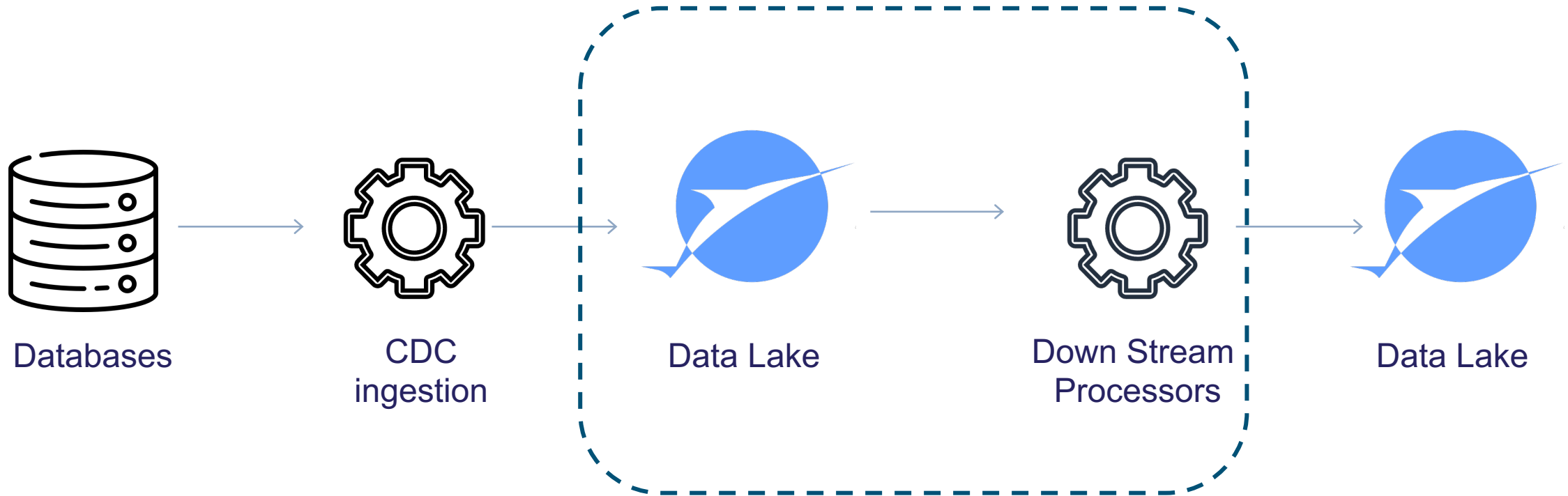
Dedicated compaction
Job



Streaming Read from Apache Paimon



Read from Apache Paimon



Downstream ETL



1. `CREATE TABLE source_table... AS (...)`
2. `CREATE TABLE target_table...AS (...)`
3. `INSERT INTO target_table SELECT * from`
`source_table;`

- Streaming read
- Read from Consumer Offset
- Configure Consumer-ID

Read from Consumer ID

```
SELECT * FROM word_count /*+ OPTIONS (  
    'consumer-id' = 'myconsumer-1',  
    'consumer.expiration-time' = '60000000'  
) */;
```

Read from Consumer Offset

```
SELECT * FROM word_count /*+ OPTIONS (  
    'scan.mode' = 'latest') */;
```

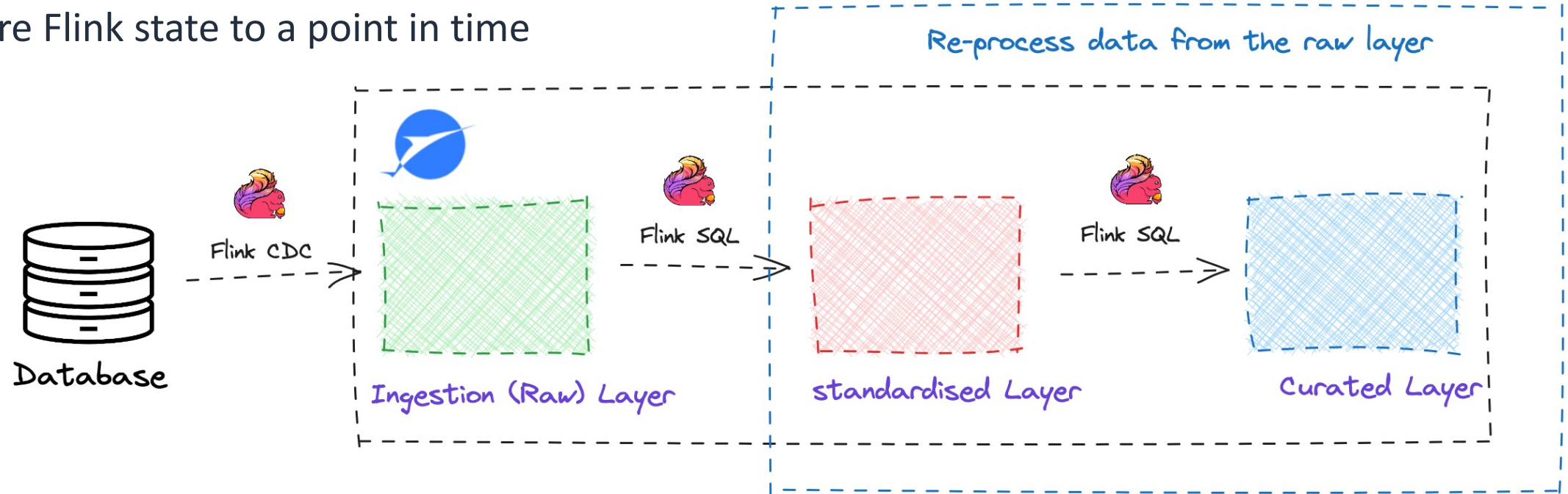
```
SELECT * FROM word_count /*+ OPTIONS (  
    'scan.timestamp-millis' = '1678883047356'  
) */;
```

Data backfill - when and how?



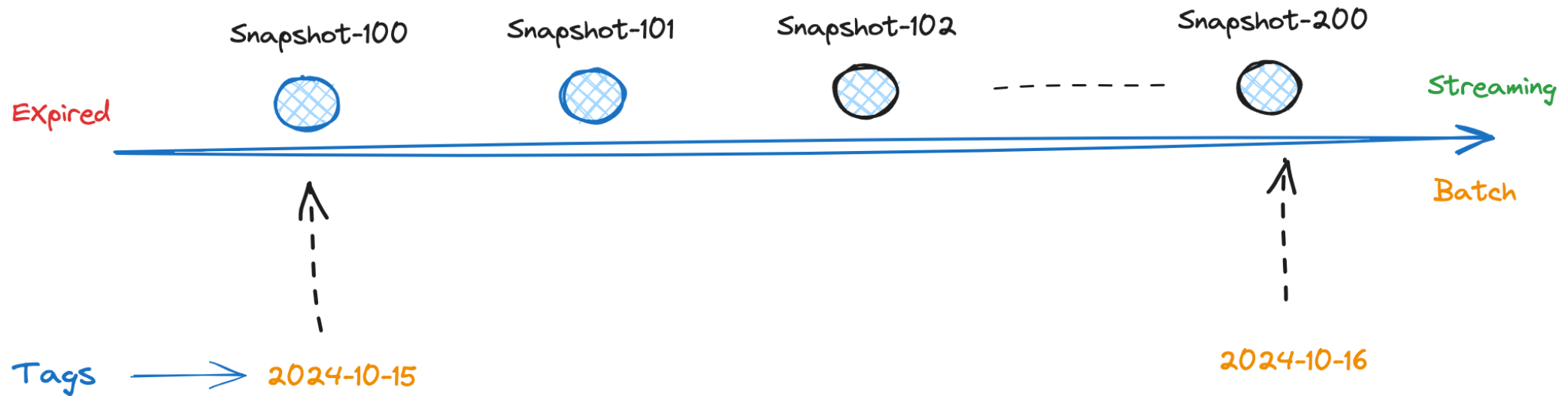
Backfill

- Historical Data correction
- Business logic changes
- Restore downstream table to a point in time
- Restore Flink state to a point in time

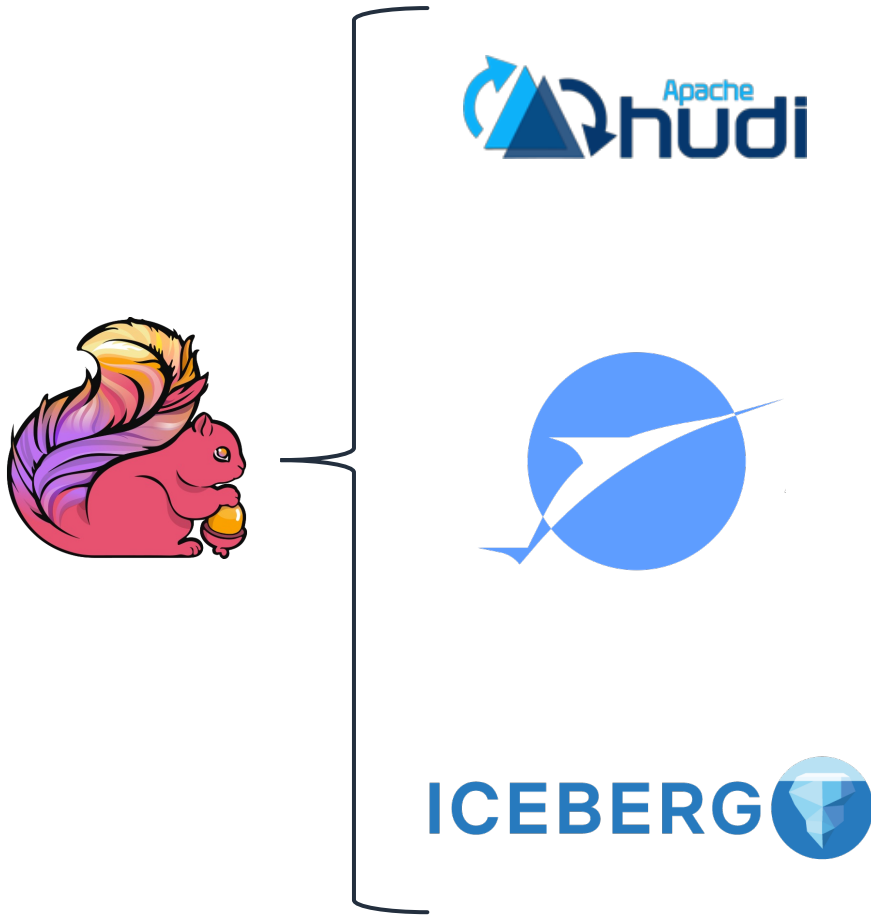


Apache Paimon Tags

- Paimon allows creation of tags to query data from previous snapshots.
- Tag contain the manifest and data files of a snapshot
- Tags can be automatically created and expired.
- You can also rollback a table to a specific tag



Read patterns



Start / End commit time or EARLIEST

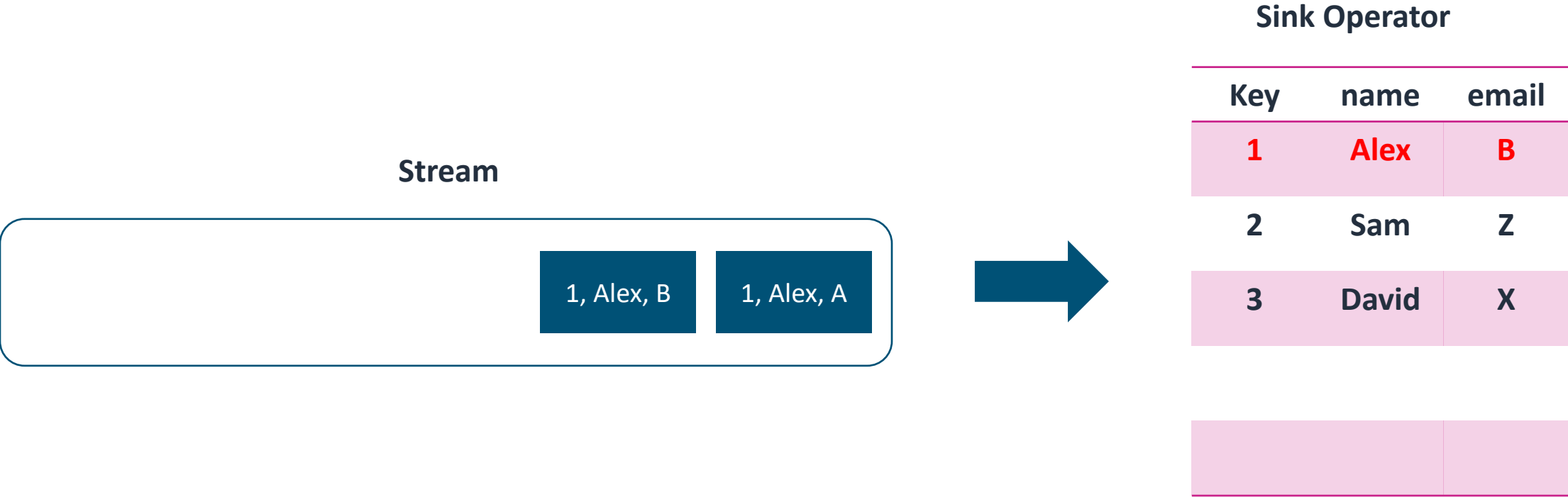
Timestamp, Snapshot_id, tag, branch, consumer-id

Snapshot_id, tag, branch

What can a robust merge engine do?



Merge engine - Deduplicate



Merge engine - Partial Update

'merge-engine' = 'partial-update'

Stream

1, Alex,



Sink Operator

Key	name	email
1	Alex	
2	Sam	Z
3	David	X



Merge engine – Partial Update

'merge-engine' = 'partial-update'

Stream



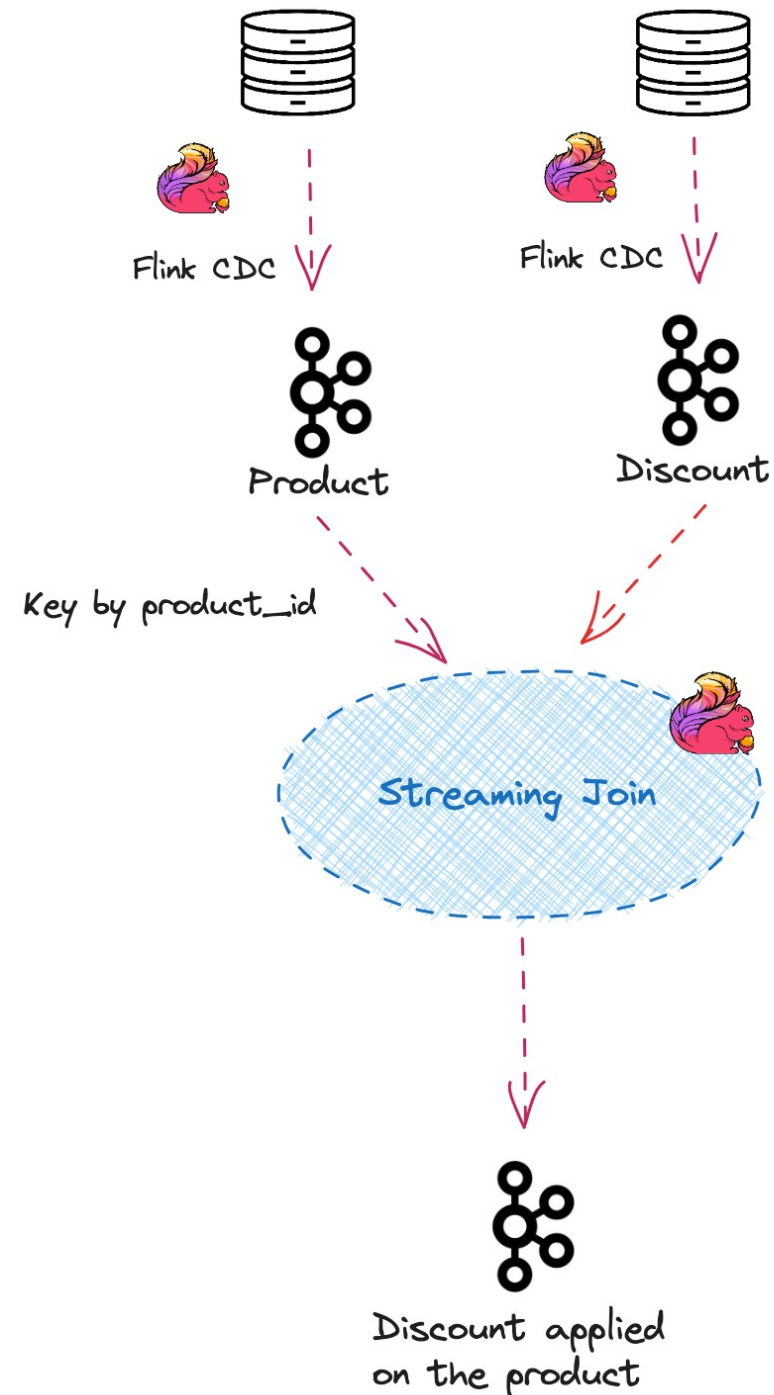
Sink Operator

Key	name	email
1	Alex	B
2	Sam	Z
3	David	X

Before Apache Paimon

Challenges with

- Schema Evolution
- Kafka Storage scaling
- Kafka capacity during backfill
- Longer retention
- Large Flink state
- Higher Cost



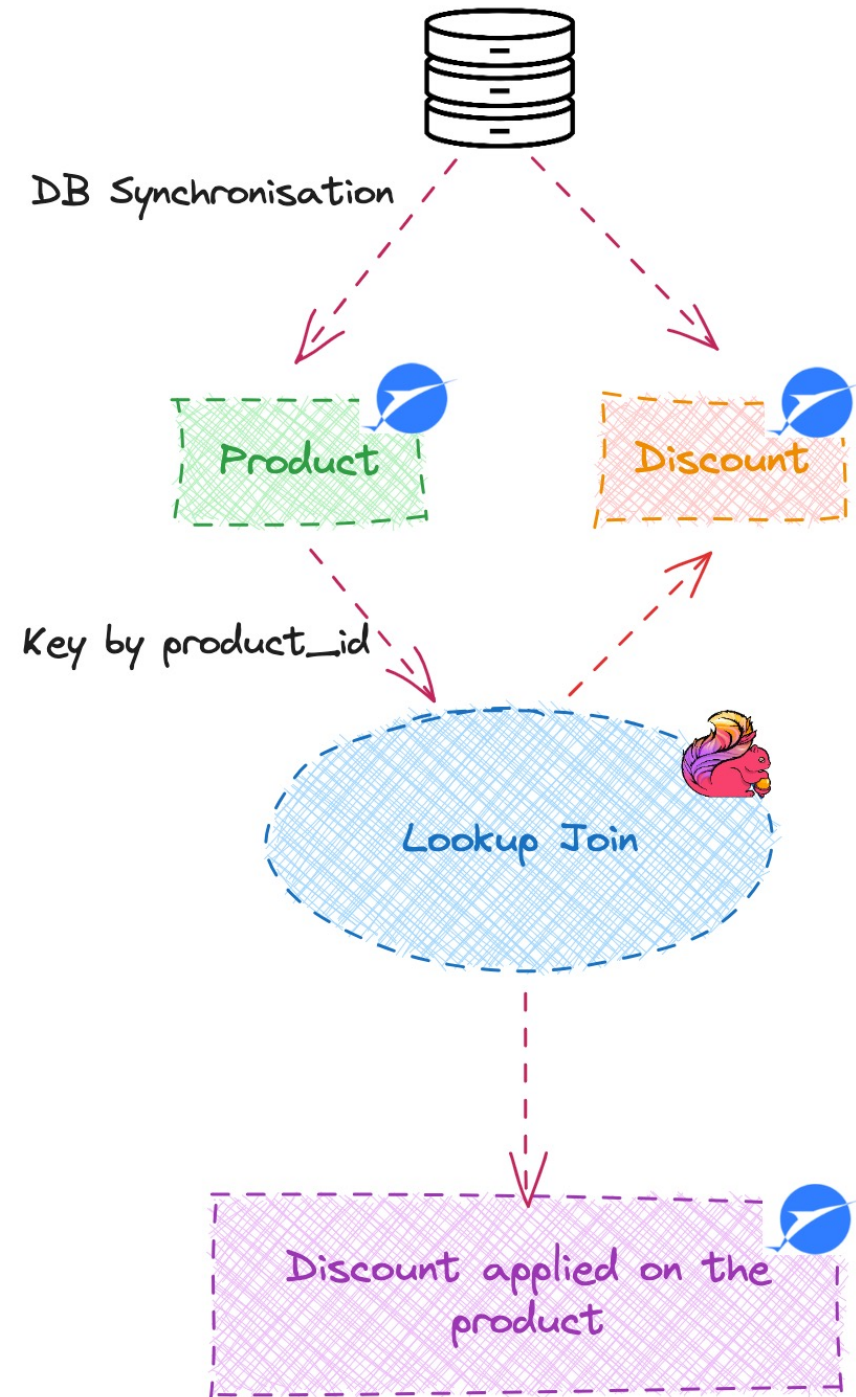
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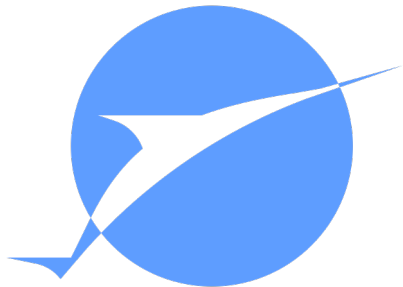
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After Apache Paimon

- Full support for Schema Evolution
- Scalable storage with Amazon S3
- Infinite retention
- Backfill is much easier
- Reduced Flink state by leveraging lookup join with Paimon table
- Low cost



Conclusion



- Data lake solves storage challenges with latency trade-off
- Apache Kafka for low-latency, data lake for under a minute
- Paimon Actions for database to lake synchronization
- Use the power of data lake, instead of Flink State
- Iceberg compatibility eliminates tooling challenges
- Additionally Paimon supports
 - Additional rich Merge Engines – Partial Update with Aggregation, Aggregation, First Row
 - Kafka like behavior for append only table



Q&A

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