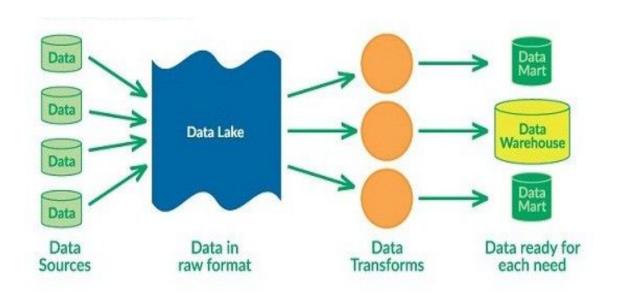


Agenda

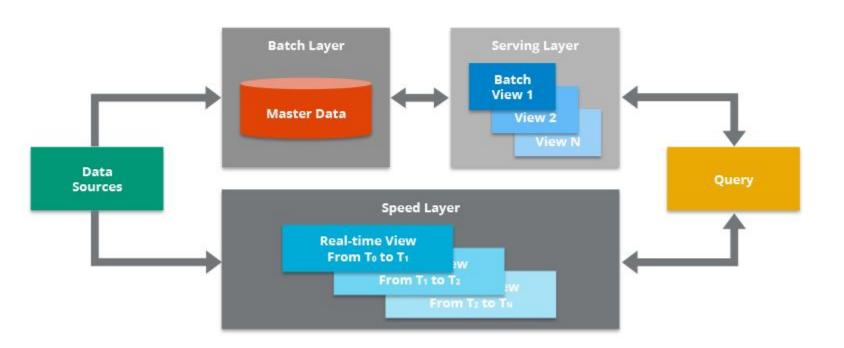
- Towards Delta
 - Traditional Data Lakes
 - Lambda architecture
 - Current challenges
- Delta Lake
 - Owner of the control of the contr
 - Architecture
 - Delta tables
- Features of Delta Engines
 - ACID Transactions
 - Schema enforcement & evolution
 - Time Travel
 - Performance Optimization, Zorder etc

Traditional Data Lake (Handling Volume)



- 1. Input Data from many sources.
- 2. Collected in a data lake.
- Processed using some distributed processing engine.
- 4. Served for analysis.

Lambda Architecture (Handling Velocity)



Challenges with current architecture

1. Query Performance

- ETL process can add latency
- With growing scale, query becomes slower.

2. System Complexity

- Unifying stream and batch doubles infra cost.
- Low level Code intervention needed to sync two pipeline data.
- Need to reset pipelines for stream failures.

3. Data Reliability

- Failed jobs can corrupt data
- Schema changes can break joins, aggregates, transformations.
- Hard to update/ delete data
- Concurrent access suffer inconsistent query results.

Delta to the rescue!

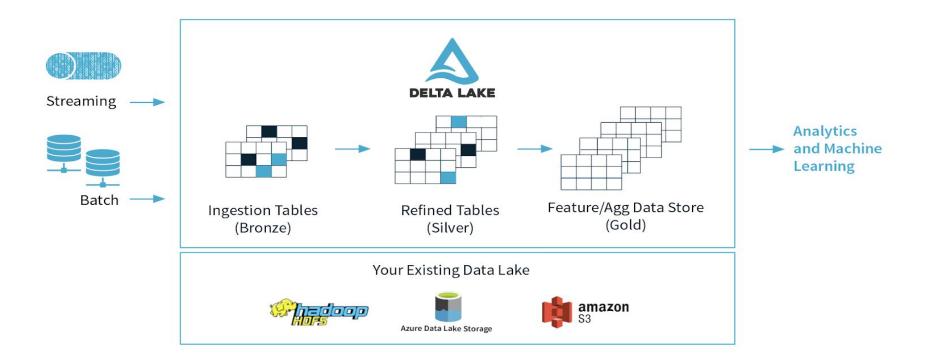


https://blog.knoldus.com/why-databricks-delta-called-as-unified-data-management/

- Delta stores data in versioned Apache Parquet format.
- Uses Delta Transaction log protocol to provide consistency.
- Provide Automatic data indexing.

Incoming data is processed as "delta" records rather than the append-only new records.

Delta Lake Architecture



Delta Table

To provide ACID guarantees we require the storage system to provide the following.

- 1. **Atomic visibility:** There must be a way for a file to be visible in its entirety or not visible at all.
- 2. **Mutual exclusion:** Only one writer must be able to create (or rename) a file at the final destination.
- 3. **Consistent listing:** Once a file has been written in a directory, all future listings for that directory must return that file.



Storage systems do not necessarily provide all of these guarantees out-of-the-box.

Delta Lake transactional operations typically go through the **LogStore API** instead of accessing the storage system directly.

https://github.com/delta-io/delta/blob/master/PROTOCOL.md

Delta Lake Features

Delta Lake is an open source storage layer that brings reliability to data lakes.

- 1. ACID transactions
- 2. Scalable metadata handling
- 3. Time travel/ Data versioning
- 4. Streaming and batch unification
- 5. Schema enforcement
- 6. Schema evolution
- 7. Updates and deletes
- 8. 100% compatible with Apache spark

Dive into Delta Lake!

- 1. ACID Transactions Demo
- 2. Schema Enforcement and Evolution Demo

Demo Summary

- 1. Delta table is a **single serial history of atomic versions**.
- 2. The state of a table at a given version is called a **snapshot**.
- 3. Delta Lake uses *optimistic concurrency control* for concurrent read-write access.

```
/mytable/_delta_log/000000000000000000.json
/mytable/_delta_log/000000000000000001.json
/mytable/_delta_log/000000000000000003.json
/mytable/_delta_log/000000000000000003.checkpoint.parquet
/mytable/_delta_log/_last_checkpoint
/mytable/part-00000-3935a07c-416b-4344-ad97-2a38342ee2fc.c000.snappy.parquet
```

Data reliability by Delta

- **ACID Transactions**: "all or nothing" transactions
- **Snapshot Isolation** : multiple readers & writers support
- **Schema Enforcement**: Delta provides schema and prevent writes that do not align
- **Schema Evolution**: via merge schema or overwrite schema as needed.
- **Exactly Once**: Delta employs checkpointing to ensure data is not missed or repeated.
- **Upserts and Deletes support**: Spark tables are write once, modifications need to be done explicitly, while delta has implicit support for them.

Data Versioning

We have seen Delta Transactions & its Logs. Now, lets see how to use them for data versioning?

- Query older snapshot of delta table
- Revert to earlier versions of data for audits
- Trace transactions on the delta lake
- Reproducible machine learning experiments



https://www.thesun.co.uk/tech/8624077/time-machine1@xperiment-quantum-physics-electrons/

Performance Optimization

So far, delta lake supported us with ACID, data versioning, schema enforcement, etc. But,

Is there any downside of this architecture?

1. Query times?

a. Delta scans so many files

2. Storage?

a. Delta maintain all versions of data

Now, lets see how delta helps improve query times.

Optimize via File Management

Data Compaction or Bin-Packing

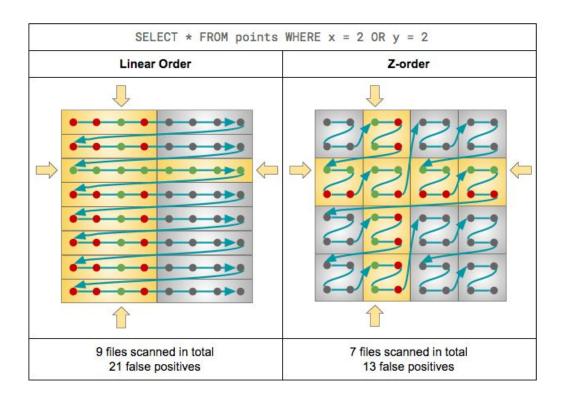
- Coalesce small files into larger ones.
- Produce evenly-balanced data files w.r.t their size on disk.
- No data related changes to the table.
- Idempotent.

Data Skipping

- Automatically collect statistics when you write data into a Delta table.
 - Parquet metadata supports this.
- Improve query execution time by skipping unrelated files.

For best results, apply **Z-Ordering**. ()

Z-ordering (multi-dimensional clustering)



- **Dot** = data points
- Box = data files; in this example, we aim for files of 4 points each
- **Yellow boxes** = data file that's read for the given query
- **Green dot** = data point that passes the query's filter and answers the query
- Red dot = data point that's read, but doesn't satisfy the filter; "false positive"

vacuum

- Remove files older than the retention threshold. Default is 7 days.
- Deletes only **data files, not log files.**
- Not triggered automatically.
- Ability to time travel back to a version older than retention period is lost after running vacuum.

Delta Cache

- Accelerates data reads by creating copies of remote files in nodes' local storage.
- Supports only Parquet files, not other storage formats.
- Stores data entirely on the local disk.
- Any stale entries are automatically invalidated and evicted from the cache.
- uses efficient decompression algorithms and outputs data in the optimal format

References

- https://docs.delta.io/latest/delta-batch.html#language-scala Demo
- https://docs.microsoft.com/en-us/azure/databricks/delta/
- https://akashrehan.wordpress.com/2019/07/11/anatomy-of-databrick s-delta-lake/
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Thank You

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