

LAB 04: Delta Lake Optimization

Duration: ~30 min | **Day:** 2 | **After module:** M04: Delta Lake Optimization I

Difficulty: Intermediate

Scenario

“The Bronze layer is growing with many small files. Your job is to optimize the Delta tables — compact files, apply Z-ORDER for faster queries, clean up with VACUUM, explore Liquid Clustering, and handle a data skew scenario.”

Objectives

After completing this lab you will be able to:

- Inspect table metrics with `DESCRIBE DETAIL`
- Compact small files with `OPTIMIZE`
- Apply Z-ORDER for query optimization
- Clean obsolete files with `VACUUM`
- Create a Liquid Clustered table
- Detect and handle data skew with `broadcast()` join

Prerequisites

- Cluster running and attached to notebook
 - Setup cell creates test tables automatically
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Tasks Overview

Open `LAB_04_code.ipynb` and complete the `# TODO` cells.

Task	What to do	Key concept
Task 1	Inspect Table Metrics	<code>DESCRIBE DETAIL</code> — check <code>numFiles</code> , <code>sizeInBytes</code>
Task 2	OPTIMIZE	Compact small files into larger ones
Task 3	ZORDER BY	<code>OPTIMIZE ... ZORDER BY (customer_id)</code>
Task 4	VACUUM	<code>DRY RUN</code> first, then execute with 0 hours retention
Task 5	Liquid Clustering	<code>CREATE TABLE ... CLUSTER BY (col)</code>
Task 6	Detect and Handle Data Skew	Identify skew with <code>GROUP BY</code> , fix with <code>broadcast()</code> join

Detailed Hints

Task 1: DESCRIBE DETAIL

- Command: `DESCRIBE DETAIL table_name`
- Look at `numFiles` and `sizeInBytes` columns

Task 2: OPTIMIZE

- Command: `OPTIMIZE table_name`
- Compare `numFiles` before and after

Task 3: ZORDER

- Syntax: `OPTIMIZE table_name ZORDER BY (column_name)`

Task 4: VACUUM

- First: `VACUUM table_name RETAIN 0 HOURS DRY RUN` (preview only)
- Then: `VACUUM table_name RETAIN 0 HOURS` (execute)
- Must disable safety check first for 0 hours retention

Task 5: Liquid Clustering

- Syntax: `CREATE TABLE ... CLUSTER BY (col) AS SELECT ...`
- Verify with `DESCRIBE DETAIL` — check `clusteringColumns`

Task 6: Data Skew

- Detect: `SELECT customer_id, COUNT(*) ... GROUP BY ... ORDER BY count DESC`
- Fix: `from pyspark.sql.functions import broadcast` → `df.join(broadcast(small_df), ...)`

Summary

In this lab you: - Inspected table metrics with `DESCRIBE DETAIL` - Compacted small files with `OPTIMIZE` - Applied `Z-ORDER` for query optimization - Cleaned obsolete files with `VACUUM` - Created a Liquid Clustered table - Detected data skew and resolved it with broadcast join

Exam Tip: Liquid Clustering replaces both partitioning and `Z-ORDER`. Use `ALTER TABLE ... CLUSTER BY (new_cols)` to change clustering columns without rewriting data. For data skew, `broadcast()` works when one side is small (< 10MB by default). AQE handles skew automatically in most cases.

What's next: In LAB 05 you will set up Auto Loader for streaming ingestion and explore Change Data Feed (CDF).