A DIVE INTO DELTA LAKE A MODERN FILE FORMAT FOR NEXT-GEN LAKES







ADVANCING ANALYTICS











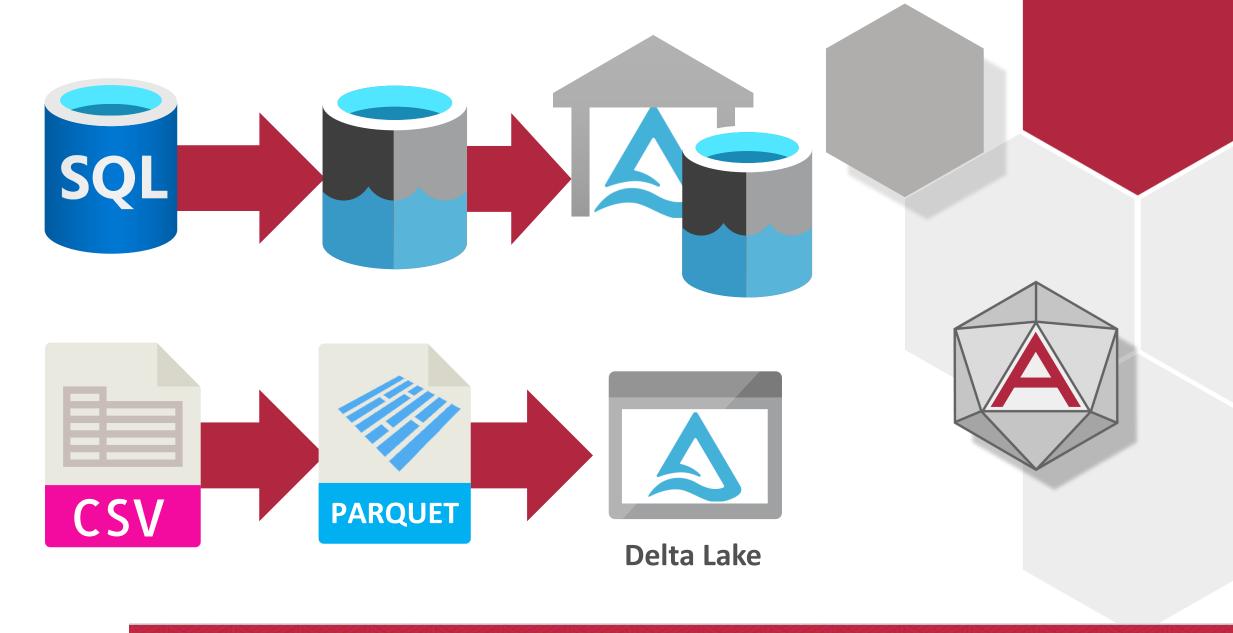


https://delta.io/

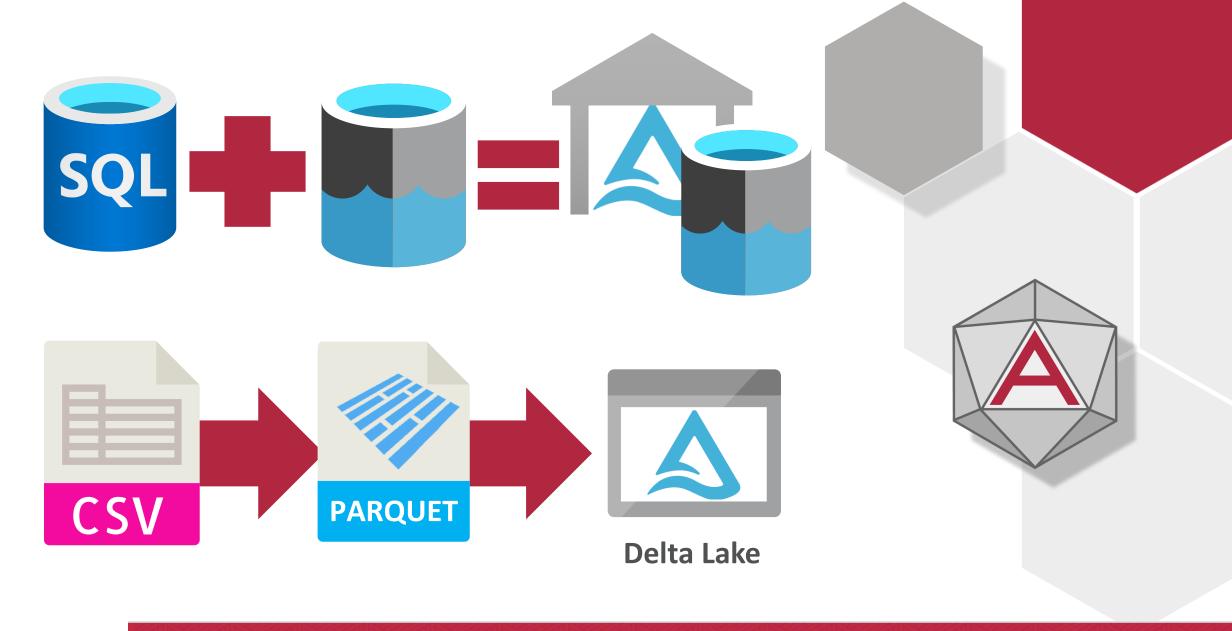




@ADVANALYTICSUK



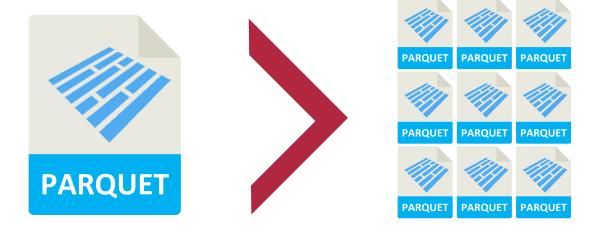
ADVANCING ANALYTICS



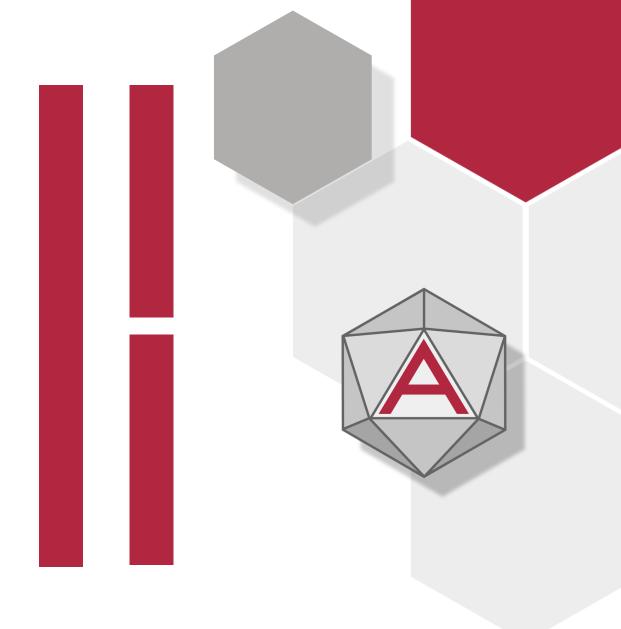
ADVANCING ANALYTICS



THE PROBLEMS OF PARQUET

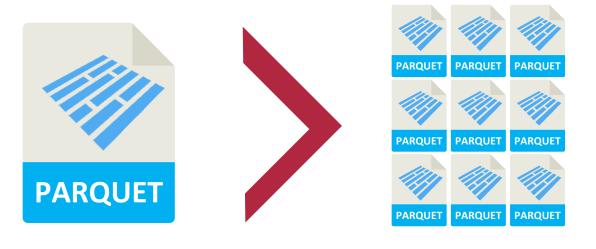


Small files have a heavy performance impact





THE PROBLEMS OF PARQUET

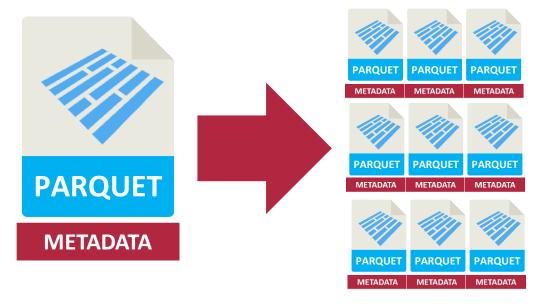


Small files have a heavy performance impact





METADATA IN DATA FILES



Metadata scan = reading all files

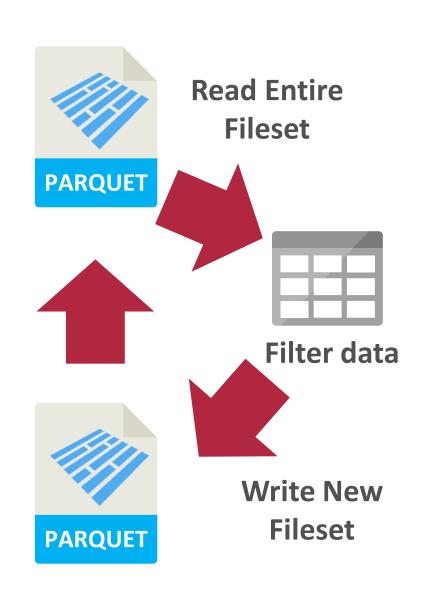




OPERATIONAL COMPLEXITY



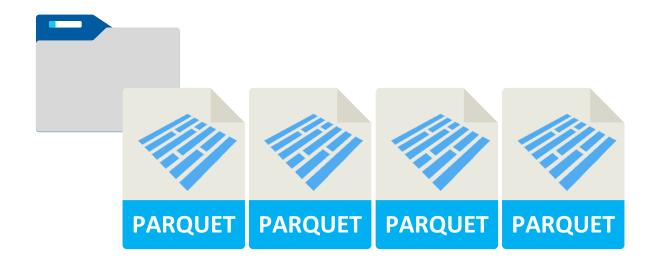
MyTable WHERE...







AUDITING & RECOVERY



DELETE FROM MyTable WHERE ID = 132





NO INDEXES PARQUET **SELECT * FROM** MyTable WHERE... PARQUET PARQUET PARQUET PARQUET

PARQUET PARQUET PARQUET PARQUET







WHAT IS DELTA?

"Databricks Delta is a unified data management system that brings reliability and performance(10-100x faster than Spark on Parquet) to cloud data lakes.



Delta's core abstraction is a Spark table with built-in reliability and performance optimizations."



WHAT IS DELTA?

Delta Lake is an **optimised**, **managed format** for organising &

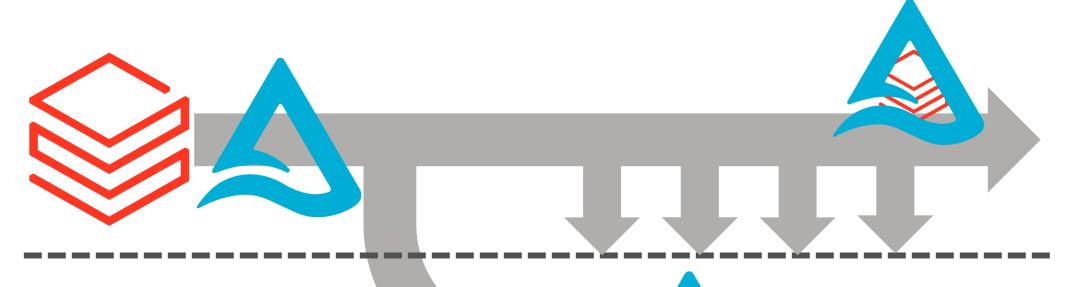
working with **Parquet** files

"It's Parquet, but better"





HOW OPEN SOURCE IS DELTA?









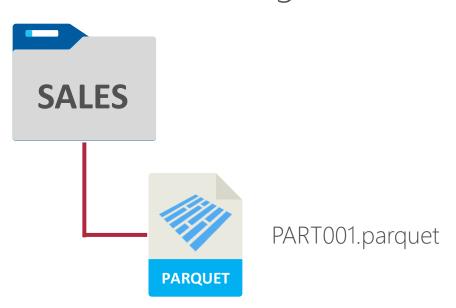


@ADVANALYTICSUK

BEFORE DELTA



DELETE * **FROM** SALES **WHERE** Segment = 3

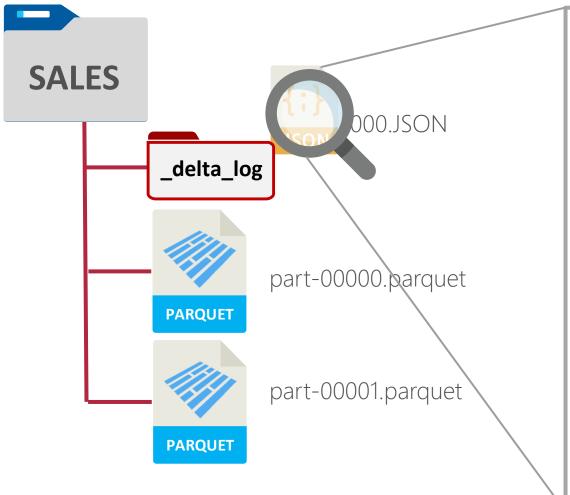


PART002.parquet

Only way to delete is to replace the existing files with a new file containing the non-deleted data



BUT WHAT ACTUALLY IS IT? - WITH DELTA:



```
"add": {
    "path": "part-00000.parquet",
    "partitionValues": {},
    "size": 255520,
    "modificationTime": 1572823237000,
    "dataChange": true,
    "stats":[...]},
"add": {
    "path": "part-00001.parquet",
    "partitionValues": {},
    "size": 242520,
    "modificationTime": 1572823237000,
    "dataChange": true,
    "stats": [...]}
```



BUT WHAT ACTUALLY IS IT? - WITH DELTA:

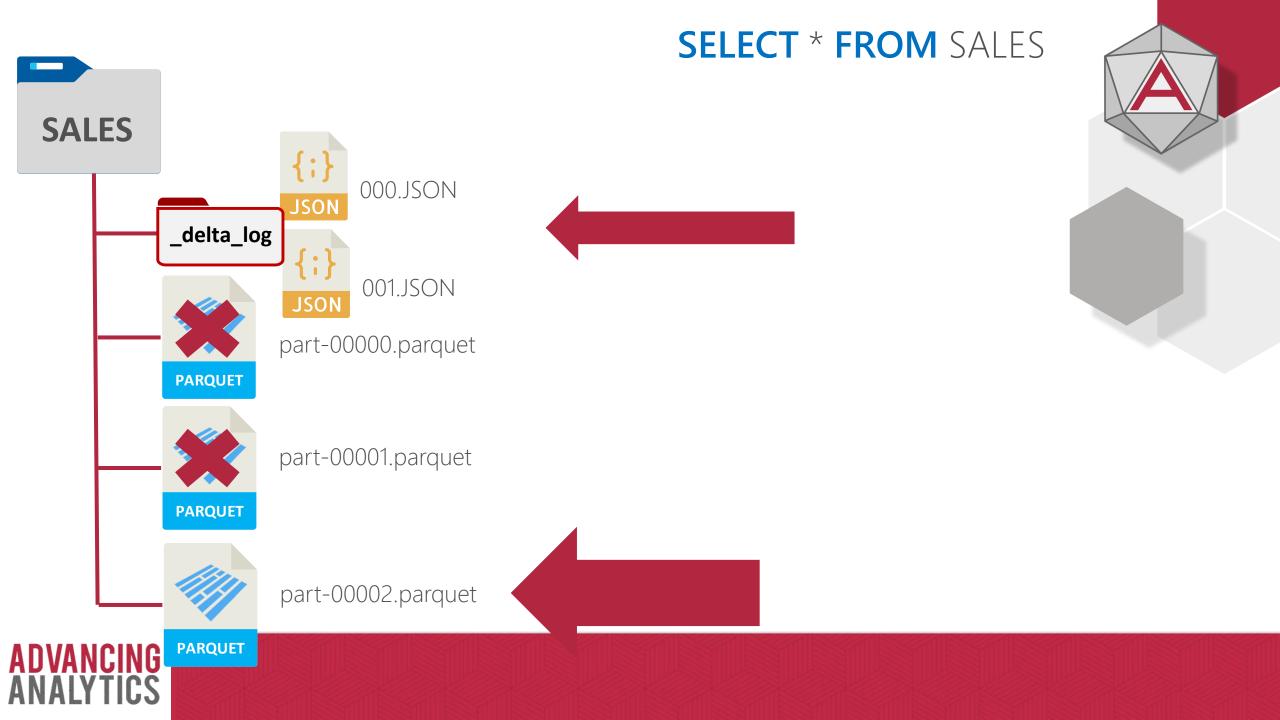
DELETE * FROM SALES WHERE

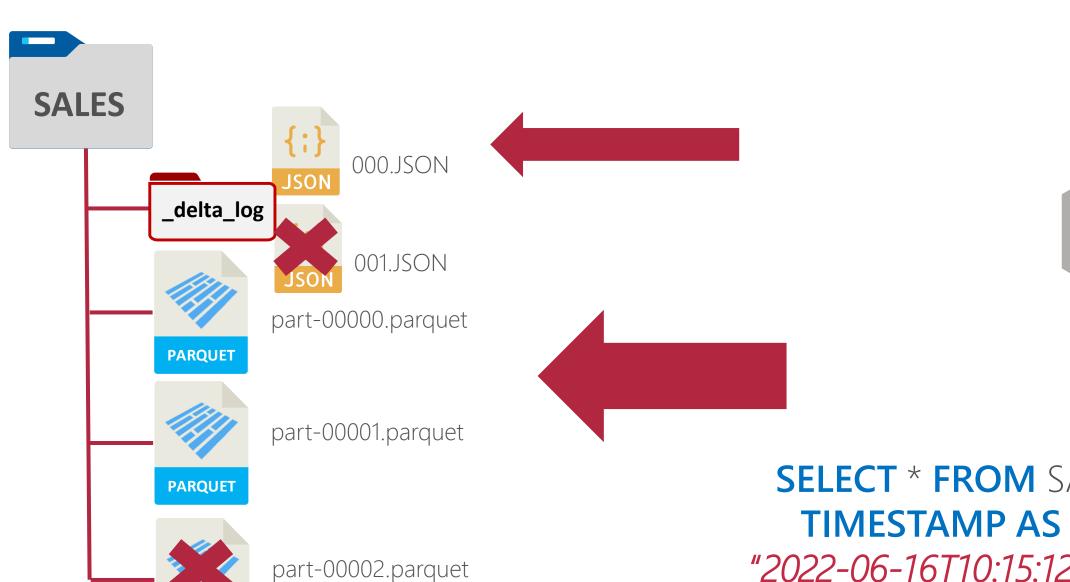
Segment = 3

```
A
```

```
SALES
                              000.JSQ4
                        JSON
            _delta_log
                               001.JSON
                       part-00000.parquet
            PARQUET
                       part-00001.parquet
            PARQUET
                       part-00002.parquet
            PARQUET
```

```
"add": {
   "path": "part-00002.parquet",
   "partitionValues": {},
   "size": 255520,
   "modificationTime": 1572823237000,
   "dataChange": true,
    "stats":[...]},
"remove": {
   "path": "part-00000.parquet",
    "modificationTime": 1572823237000,
    "dataChange": true},
"remove": {
   "path": "part-00001.parquet",
    "modificationTime": 1572823237000,
    "dataChange": true}
```









PARQUET



Local Development

We're assuming Spark will be the main tool uses to interact with Delta. So you need a local spark environment first!

https://delta.io/learn/getting-started

Once your spark environment is configured, delta can be installed during initialisation:

pyspark --packages io.delta:delta-core_2.12:1.0.0 \
 --conf
"spark.sql.extensions=io.delta.sql.DeltaSparkSessionExtens
ion" \
 --conf
"spark.sql.catalog.spark_catalog=org.apache.spark.sql.delt
a.catalog.DeltaCatalog"

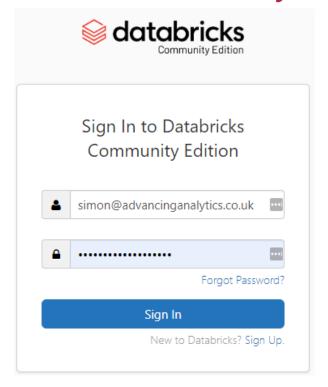
You will need to import delta within your pyspark scripts, then you can use it as you wish!

Hosted Sandbox

However... spark setup can be awkward and has several installation pre-requisites.

For today's examples, we can use a free sandbox version of the Databricks hosted spark to test out Delta features

https://databricks.com/try-databricks

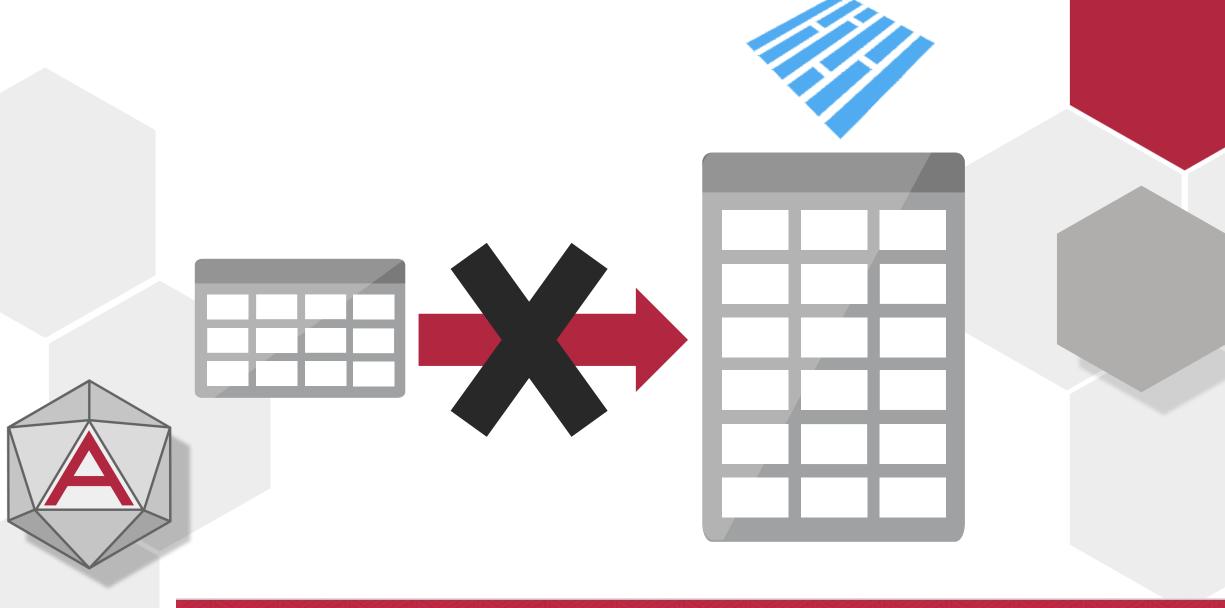




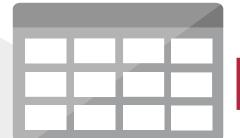
DEMO: WORKING WITH DELTA

- Creating a Delta Table
- The Delta Transaction Log
- Updating Delta Tables





ADVANCING ANALYTICS Delta supports minor schema drift









DELTA MERGE



```
MERGE INTO <target_table>
USING <source_table>
ON <merge_condition>
[ WHEN MATCHED [ AND <condition> ] THEN <matched_action> ]
[ WHEN MATCHED [ AND <condition> ] THEN <matched_action> ]
[WHEN NOT MATCHED [AND <condition>] THEN <not_matched_action>]
where
<matched action> =
 DELETE |
 UPDATE SET * |
 UPDATE SET column1 = value1 [, column2 = value2 ...]
<not matched action> =
 INSERT *
 INSERT (column1 [, column2 ...]) VALUES (value1 [, value2 ...])
```



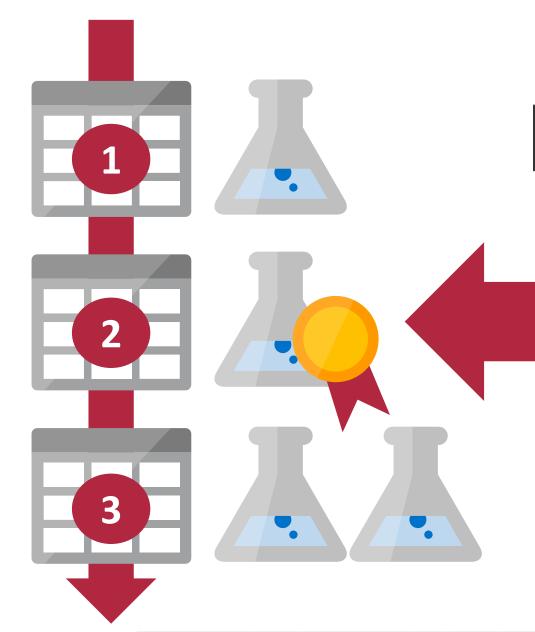






TABLE AS OF VERSION 2

Reduced need for multiple copies of data attached to evolving machine learning experiments





DEMO: DELTA UPDATES

- Schema Drift
- Merge into a Delta Table
- TIME TRAVEL



AUDITING DELTA TABLES

Delta Tables have additional commands for interrogating the transaction log:

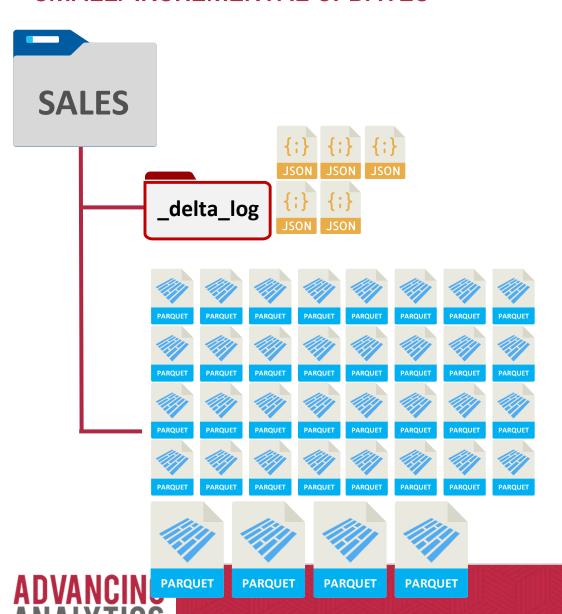
Describe < Table> - Returns the table schema as usual

Describe Detail <Table> – Returns the underlying delta metadata, number of files, storage location etc

Describe History <Table> - Displays the most recent changes made to the table (inserts, updates, optimize, vacuum etc)



SMALL/INCREMENTAL UPDATES

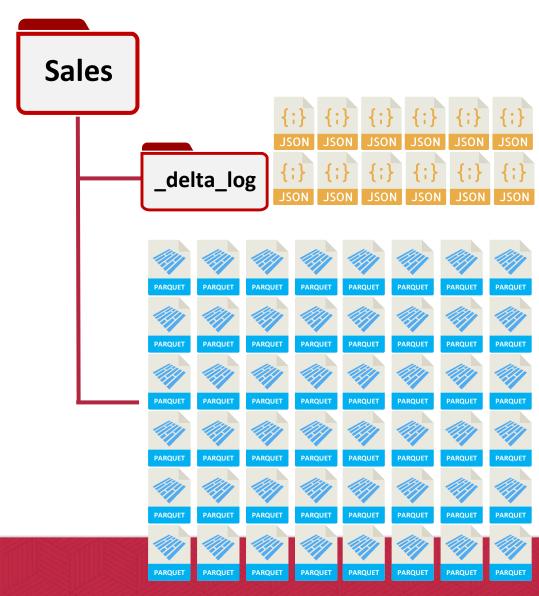


The **Optimize** command compacts small files into larger, better compressed files

This is treated like all other updates, files are NOT deleted

Optimize a Delta Table in Spark SQL OPTIMIZE [TableName]





To remove obsolete history files,

Delta has the **VACUUM** command

This command physically deletes data files older than a specified date

You CANNOT time travel past dates where history has been vacuumed



USING THE VACUUM COMMAND

Vacuuming in SQL:

--Vacuum Table using defaults VACUUM [database].[table]

--Vacuum using path not Hive table VACUUM '/mnt/lake/BASE/myTable/'

--VACUUM for a non-default time period VACUUM [database].[table] RETAIN 168 HOURS

--TEST THE VACUUM BEFORE YOU RUN IT VACUUM [database].[table] RETAIN 168 HOURS DRY RUN

Using the python deltaTable object:

Vacuum Table using defaults deltaTable.vacuum()

Vacuum Table for files older than 7 days (168 hours) deltaTable.vacuum(168)



FILE SKIPPING

SELECT * FROM events WHERE year=2020 AND uid=24000

year: min 2018, max 2019 file1.parquet uid: min 12000, max 23000 year: min 2018, max 2020 file2.parquet uid: min 12000, max 14000 year: min 2020, max 2020 file3.parquet

25000

uid: min 23000, max

skipped as data range outside selected value



DELTA MANAGEMENT

- View Delta Metadata
- Optimize a Delta Table
- Vacuum a Delta Table



THE DATA LAKEHOUSE

- Structured
- Governed
- Familiar
- Fast
- Flexible
- Cheap
- Scalable









https://delta.io/

Build Lakehouses with Delta Lake

<u>Delta Lake</u> is an open-source storage framework that enables building a <u>Lakehouse architecture</u> with compute engines including Spark, PrestoDB, Flink, Trino, and Hive and APIs for Scala, Java, Rust, Ruby, and Python.

Get Started

Github Releases Roadmap





THANKS FOR LISTENING



Twitter: @MrSiWhiteley

youtube.com/c/AdvancingAnalytics

AdvancingAnalytics.co.uk