x-gov Data Engineering ETL & Espresso

Evaluating Iceberg and Athena for curation

MoJ Analytical Platform

Codebase: iceberg-evalution

- Context
- Challenges
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- Outcome
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Context

- Use AWS DMS to extract full and changed data
- Use AWS Glue to dedupe and version data (SCD2)
- Use AWS Athena to derive new tables
- Process data incrementally, but replace target table every time
- Tables can very in size from 10 - 3B rows (~100GB)

Challenges

- Cost of AWS Glue Job quadrupled!
- Performance of AWS Glue job degraded
- Large volumes of intermittent missing data and duplicates
- Difficult to debug/maintain complex glue job

Scope

Things we needed

1. Reliable / Resilient curation architecture
2. Cost effective
3. Easy to implement and for new data engineers to pick up

Things we liked

1. Compatible with other data pipelines
2. Compatible with existing tech stack

Options - Table Format

Table formats provide a table-like abstraction on top of native file formats like Parquet by storing additional metadata.

Today, no single table format is the best fit for all use cases, and each format has its own unique strengths for specific requirements (source)

Iceberg was the clear winner for our use case because of enhanced Athena support.

Options - Compute

Option 1: Use Glue PySpark and Iceberg orchestrated using Step Functions

Option 2: Use Athena and Iceberg orchestrated using dbt

Evaluation

Curate 0.1TB TPCDS stores_sales table (~0.3B rows, 21 GB)

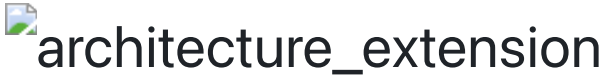
Evaluation

Curate 3TB TPCDS stores_sales table (~8B rows, 440 GB)

Out-of-the-box, Athena + Iceberg is **cheaper** and more **performant** for our use cases than Glue PySpark + Iceberg

Outcome

- We have already migrated 2 pipelines, and are in the process of migrating 2 more
- Achieved a cost reduction of 99%, and estimate cost savings of ~380K this year



Lessons Learnt

dbt vs Step Functions

Don't make any assumptions! The downstream pipelines were already using Athena and dbt for scheduling.

dbt is better suited for customised models, as opposed to running the same process against a large number of tables

Scale-out vs scale-up

We evaluated the impact of scale-up i.e. what happens as the number of rows increases. We did not evaluate the impact of scale-out i.e. what happens as the number of tables increases.

Further Work

- Support snapshots (in progress)
- Build incremental pipelines to speed up run-time
- Analyse and publish run and test metrics