# Accelerating CDC Data Ingestion with Apache Beam: A Qlik-to-BigQuery Journey

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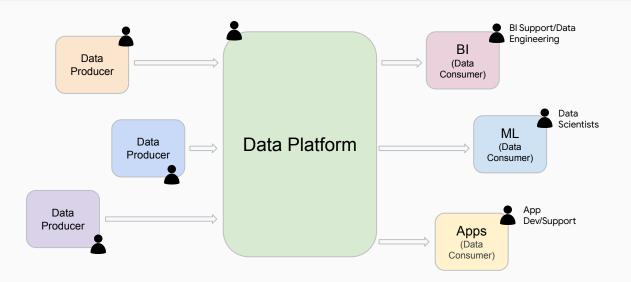
## Agenda

- Overview
- Guiding Principles
- Overall Architecture
- Ingesting CDC to BigQuery
  - Qlik Data Format
  - Error Handling
- Observation and Optimization



### Overview

- Ingest data from different RDBMS sources (e.g., change data captures)
- Vend out curated data useful to multiple eligible consumers.
- Decoupled systems & clean
   Operational Ownership.





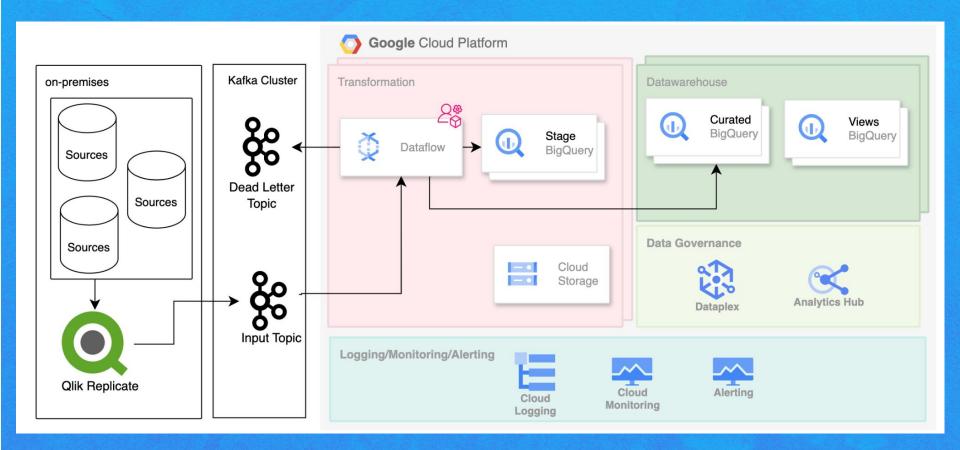
## **Guiding Principles**

- Managed services
  - Scalable framework
  - Petabyte scale storage and Queries
- Code Reusability
  - o Build Once. Run for all CDC ingestion
- Data security and privacy first
- Making data available for Business Analysts within seconds
- Infrastructure as Code
- Operational observability



## **Overall Architecture**







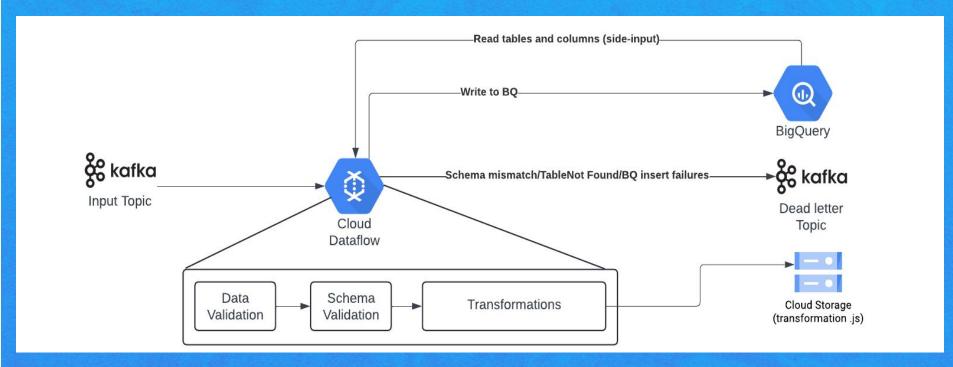
## Collaborate & Prep

- Cross collaboration between Data governance team, Data privacy team, Data owners and Data stewards.
- Extract Schema from the sources
  - Create BigQuery dataset and table (use SQL translation service)
    - Append extra columns (such as replicate\_timestamp, sequence#)
- Access controls
  - Define policy tags for the columns
  - o Define row level access controls
- Define tag templates for data discoverability



## Ingesting CDC Changes to BigQuery







#### **Qlik Data Format**

```
"data": {
   "{ColumnName} REPL int": 36507787,
    "{ColumnName} REPL nvarchar": "TPHONE",
    "_replicate_schemaname": "{schema}",
    "_replicate_tablename": "{tableName}",
    "_replicate_userid": null,
    " replicate commit_ts": "2024-08-09 09:24:17.601000"
"beforeData": null,
"headers": {
    "operation": "INSERT",
    "changeSequence": "20240809142417600000000000000013145",
    "timestamp": "2024-08-09T14:24:17.601",
    "streamPosition": "0000C09A:00003188:0002",
    "transactionId": "0000000000000000000000000371046D6",
    "changeMask": "1FFFFF",
    "columnMask": "1FFFFF",
    "transactionEventCounter": 1,
    "transactionLastEvent": true
```

#### Iteration I

{columnName}\_\_REPL\_\_{format}

- Schema validation to check the Column Name exists in BigQuery table
- Dataflow pipeline casts the value to proper format

#### **Iteration II**

{columnName}

- Drop REPL part.
- Use the schema information from BQ tables

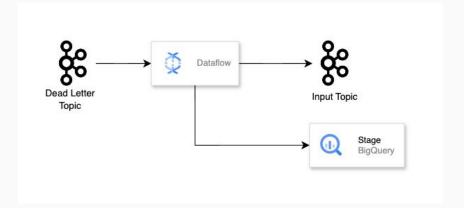


## **Error handling**

Records with data-related errors are persist in BQ stage dataset (Manual inspection required)

- Bad data/Unserializable data
- Format conversion Errors

Other errors such as table not found and schema mismatch are resolved during replay.



## Observations and Optimizations

- Enable Autosharding in BigQuery
- Enable cloud profiling (in dev/test) to identifying any bottlenecks in your functions
- Leverage DoFn lifecycle to speed up per element processing when external API is involved.
- Number of partitions in kafka topic



## Observations and Optimizations

- Kafka consumer configurations
  - o unboundedReaderMaxReadTimeSec: Use lower for low latency pipeline
  - unboundedReaderMaxElements: Use higher number if pipeline performs aggregation
- Restrict excessive logging in Dataflow pipelines
  - defaultWorkerLogLevel
  - --workerLogLevelOverrides={"<package/class>":"<level>","<package/class>":"<level>"}
- Similar deployment configurations
  - Run ingestion pipelines using the same worker type configuration
  - Capping the maximum workers number to avoid very large fleets
- [Future] Use BigQuery CDC



## Thank you!

Questions?

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