

Data Lineage with Apache Airflow using OpenLineage

Julien Le Dem and Willy Lulciuc, Datakin | March 2022

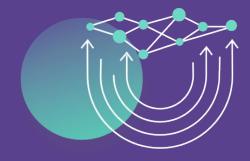




The principal sponsors of OpenLineage & Marquez

ASTRONOMER

The company behind **Apache Airflow**



ASTRONOMER



Agenda

The need for lineage metadata

OpenLineage and Marquez

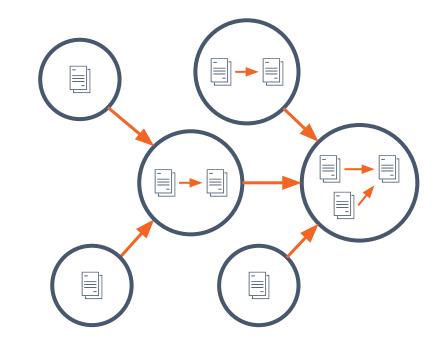
- OpenLineage, an open standard for lineage collection
- Marquez, its reference implementation

Airflow observability with OpenLineage

The key = data lineage

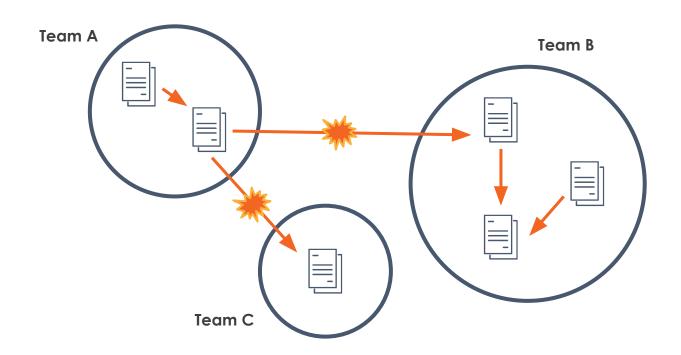
Data lineage contains what we need to know to solve our most complicated problems.

- Producers & consumers of each dataset
- Inputs and outputs of each job





Building a healthy data ecosystem





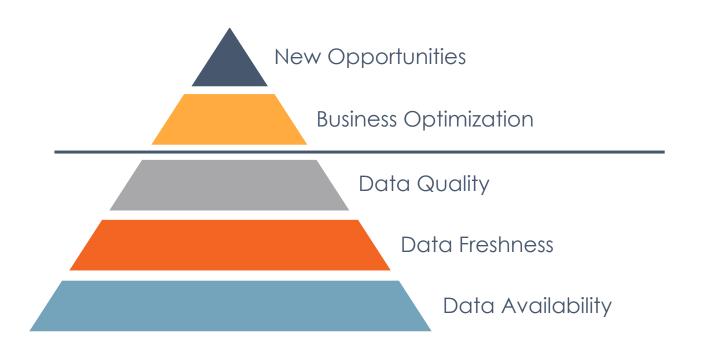
Limited metadata = limited context



- What is the data source?
- What is the schema?
- Who is the owner?
- How often is it updated?
- Where does it come from?
- Who is using it?
- What has changed?



Maslow's Data hierarchy of needs





OpenLineage

Mission

To define an **open standard** for the collection of lineage metadata from pipelines **as they are running**.





OpenLineage contributors

























The snowball effect



The best time to collect metadata



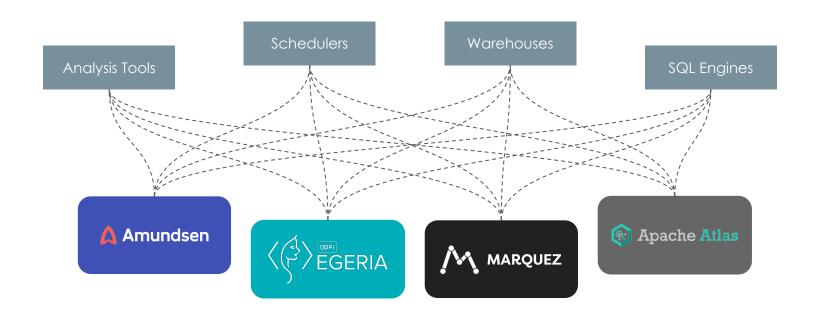
You can try to infer the date and location of an image after the fact...



...or you can capture it when the image is originally created!

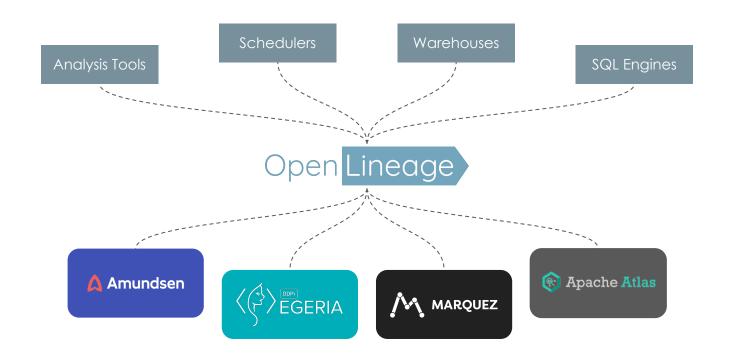


Before OpenLineage



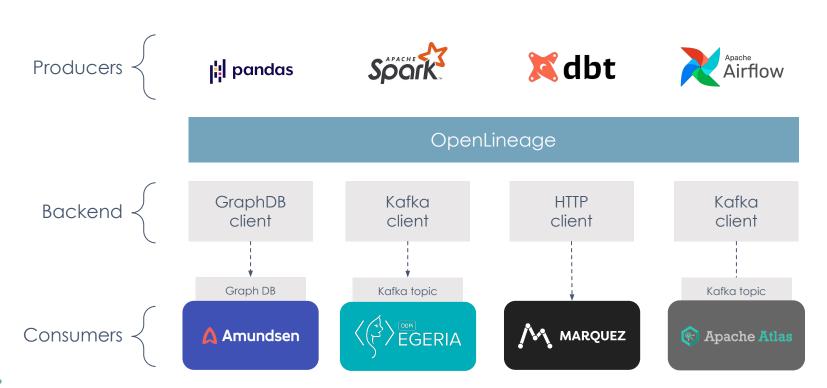


With OpenLineage



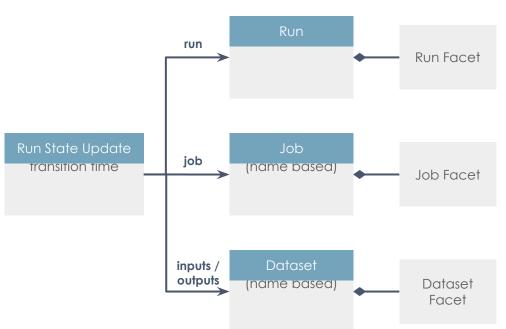


Where OpenLineage potentially fits





Data model



Built around core entities: Datasets, Jobs, and Runs

Defined as a JSONSchema spec

Consistent naming for:
Jobs (scheduler.job.task)
Datasets (instance.schema.table)



How OpenLineage events work

Lineage is reported as a series of asynchronous run events.

Each event passes a unique client-generated run ID to:

- identify the run
- correlate events

Typical event series:

Send start event

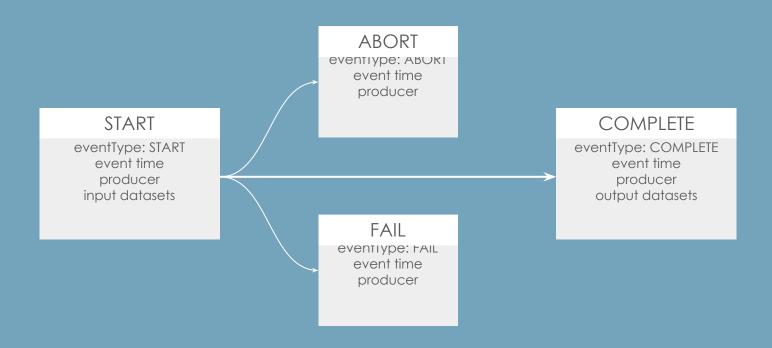
- source code version
- run parameters

Send complete event

- input dataset
- output dataset version
- output schema



Lifecycle of a job run





Extending the model with Facets

Facets are atomic pieces of metadata attached to core entities.

Self-documenting

Facets can be given unique, memorable names

Flexible

Facets can be attached to any core entity: Job, Dataset & Run

Familiar

Facets are defined using JSON schema objects

Scalable

Prefixes on names are used to establish discrete namespaces



Facet examples

Dataset:

- Stats
- Schema
- Version

Job:

- Source code
- Dependencies
- Source control
- Query plan

Run:

- Scheduled time
- Batch ID
- Query profile
- Params



OMG the possibilities are endless

Dependency tracing
Root cause identification
Issue prioritization
Impact mapping
Precision backfills
Anomaly detection
Change management
Historical analysis
Compliance



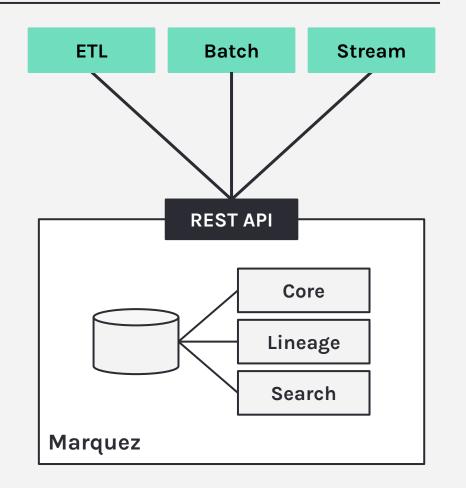


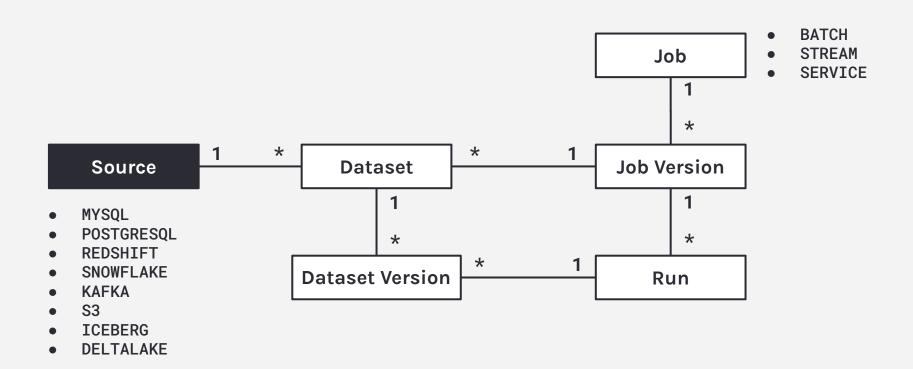


Marquez: Design

Metadata Service

- Centralized metadata management
 - Sources
 - Datasets
 - Jobs
- Features
 - Data governance
 - Data lineage
 - Data discovery + exploration

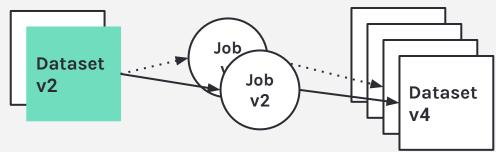




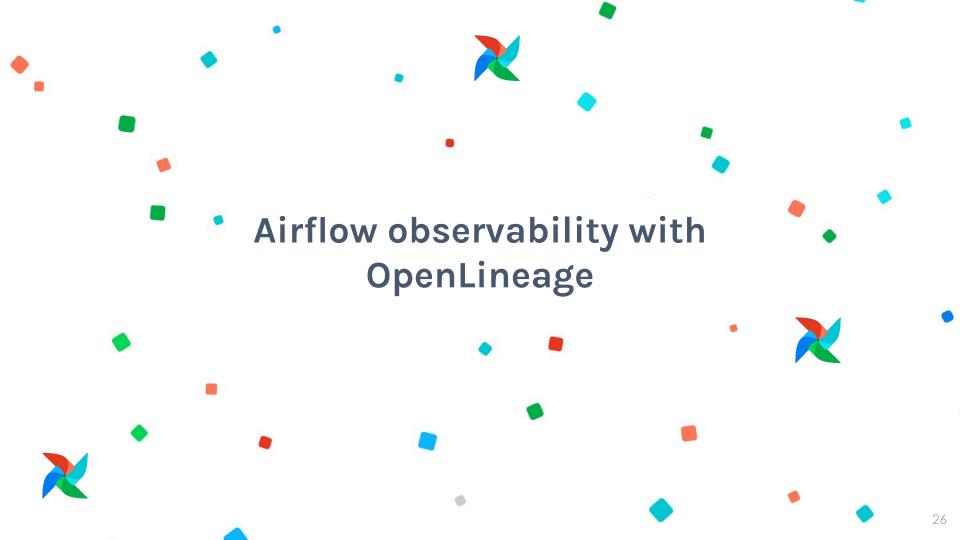


Design benefits

- Debugging
 - What job version(s) produced and consumed dataset version X?



- Backfilling
 - Full / incremental processing



Airflow support for Marquez



Metadata

- Task lifecycle
- Task parameters
- Task runs linked to versioned code
- Task inputs / outputs

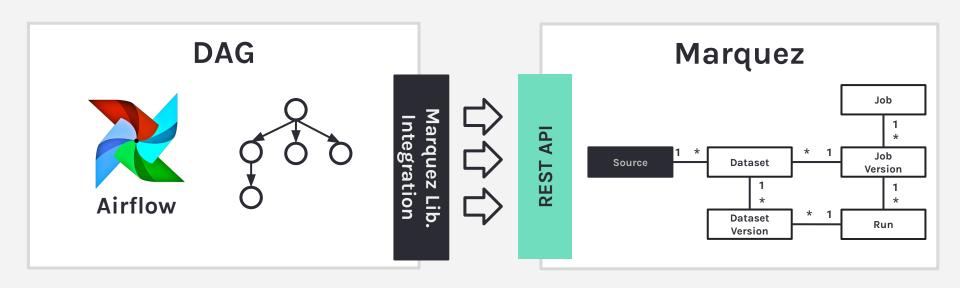
Lineage

Track inter-DAG dependencies

Built-in

- SQL parser
- Link to code builder (GitHub)
- Metadata extractors

Capturing task-level metadata in a nutshell



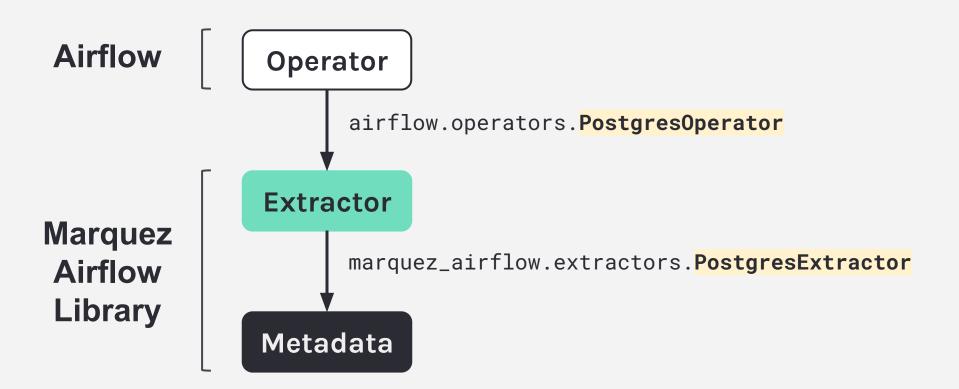
Marquez: Airflow

Marquez Airflow Lib.

- Open source! 5
- Enables global task-level metadata collection
- Extends Airflow's DAG class

```
from marquez_airflow import DAG
from airflow.operators.postgres_operator import PostgresOperator
...
```

Marquez: Airflow



Operator Metadata

```
new_room_booking_dag.py
t1=PostgresOperator(
task_id='new_room_booking',
postgres_conn_id='analyticsdb',
 sql='''
   INSERT INTO room_bookings VALUES(%s, %s, %s)
 , , ,
 parameters=... # room booking
```

01 Source

Operator Metadata

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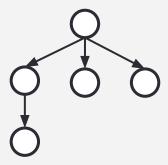
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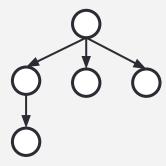
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Managing inter-DAG dependencies

new_room_bookings_dag.py

top_room_bookings_dag.py



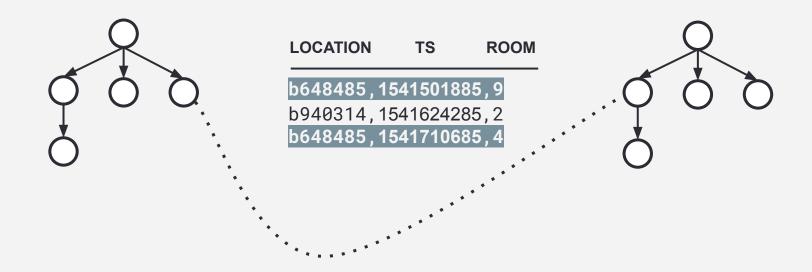


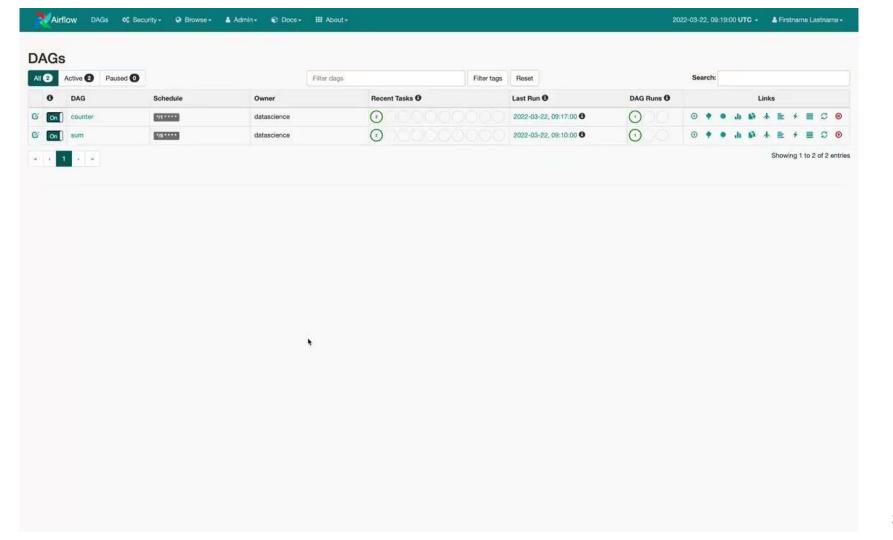
Managing inter-DAG dependencies

new_room_bookings_dag.py

public.room_bookings

top_room_bookings_dag.py





Join the conversation



github.com/openlineage

openlineage.slack.com

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groups.google.com/g/openlineage



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Thanks:)

