



OpenLineage: from Operators to Hooks



Maciej Obuchowski



in

Astronomer

Maciej is a software engineer, Airflow and OpenLineage committer. He loves rock climbing, contributing to open source data projects and playing with cats.

Sessions by Maciej Obuchowski

- [OpenLineage: From Operators to Hooks](#) (2024)
- [OpenLineage in Airflow: A Comprehensive Guide](#) (2023)
- [OpenLineage & Airflow - data lineage has never been easier](#) (2022)



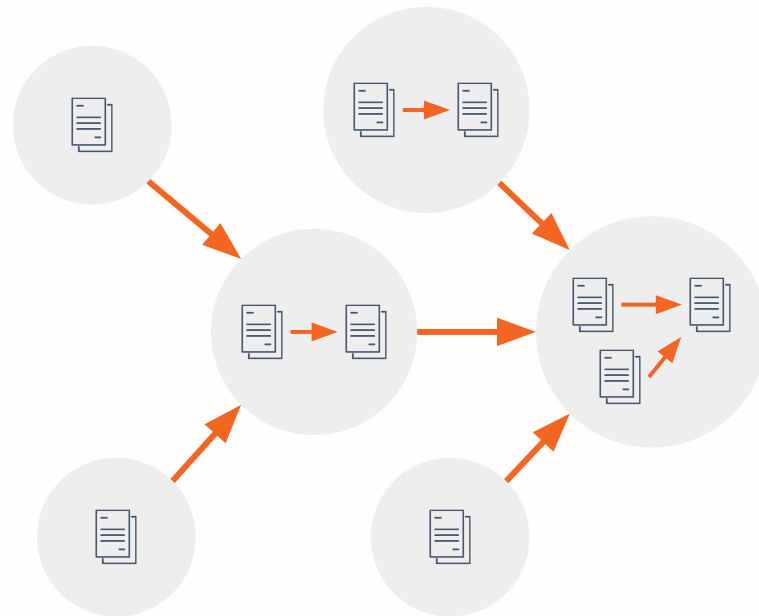
Agenda

- What is OpenLineage
- OpenLineage Airflow Integration
- Getting Lineage From Hooks
- Peek into the future?

What is Data Lineage?

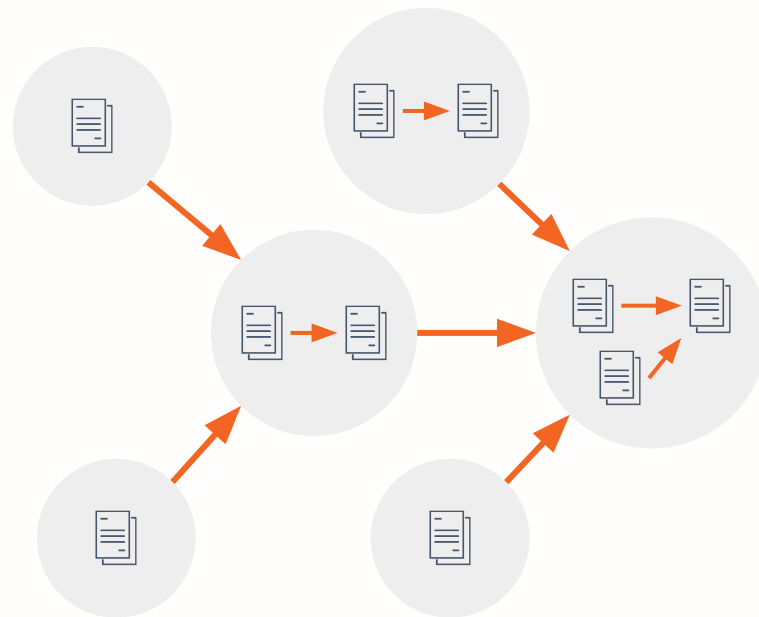
Data lineage is the set of complex relationships between datasets and jobs in data pipelines.

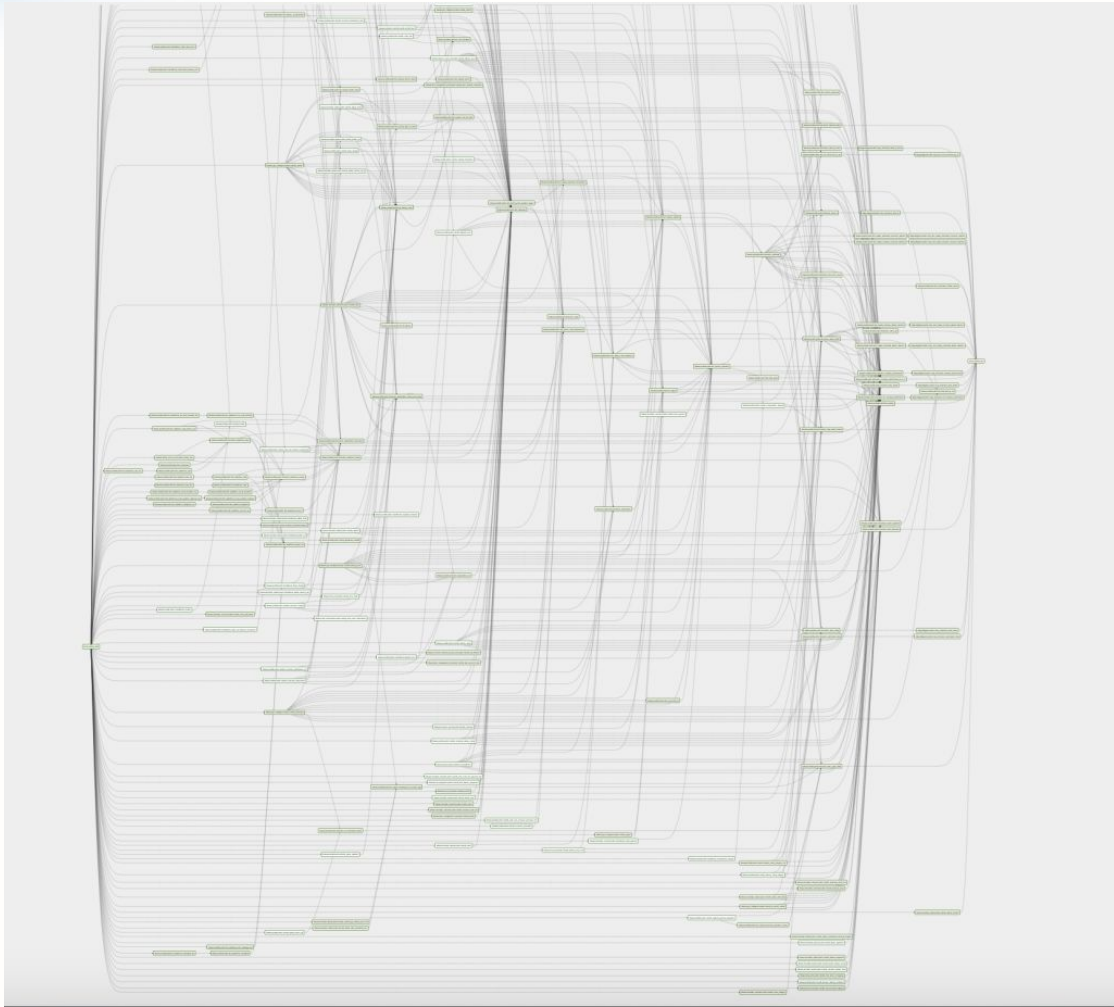
- Producers & consumers of each dataset
- Inputs and outputs of each job



What problems Data Lineage solves

- Holistic view on data flowing through organization
- Security and Compliance
 - A German bank suffered data breach from vendor - but was wholly unaware that PII data was being send there.
 - Prod data send to dev env
- Impact analysis - pipeline failed, which other datasets it affects







Working with data in 2024

The image is a comprehensive grid of logos, categorized into several main sections:

- INFRASTRUCTURE:** Includes STORAGE (Amazon S3, Google Cloud Storage, Microsoft Azure Storage, Oracle Cloud Infrastructure, IBM Cloud Object Storage, AWS S3, Oracle Cloud Infrastructure, IBM Cloud Object Storage), HADOOP (Cloudera, Databricks, Amazon EMR, Microsoft Azure HDInsight, Google Cloud Dataproc, AWS EMR, Oracle Cloud Infrastructure, IBM Cloud Object Storage), DATA LAKES (Databricks, Amazon Lake Formation, Microsoft Azure Data Lake Storage, Google Cloud BigLake, AWS Lake Formation, Oracle Cloud Infrastructure, IBM Cloud Object Storage), DATA WAREHOUSES (Snowflake, Amazon Redshift, Microsoft Azure Synapse Analytics, Google Cloud BigQuery, AWS Redshift, Oracle Cloud Infrastructure, IBM Cloud Object Storage), STREAMING / IN-MEMORY (Amazon Kinesis, Microsoft Azure Stream Analytics, Google Cloud Dataflow, AWS Kinesis, Oracle Cloud Infrastructure, IBM Cloud Object Storage).
- ANALYTICS:** Includes BI PLATFORMS (Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), VISUALIZATION (Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), DATA ANALYST PLATFORMS (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), AUGMENTED ANALYTICS (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics).
- MACHINE LEARNING & ARTIFICIAL INTELLIGENCE:** Includes DATA SCIENCE NOTEBOOKS (Databricks, Amazon SageMaker, Microsoft Azure ML, Google Cloud AI Platform, AWS SageMaker, Oracle Cloud Infrastructure, IBM Analytics), DATA SCIENCE PLATFORMS (Databricks, Amazon SageMaker, Microsoft Azure ML, Google Cloud AI Platform, AWS SageMaker, Oracle Cloud Infrastructure, IBM Analytics), ML PLATFORMS (Databricks, Amazon SageMaker, Microsoft Azure ML, Google Cloud AI Platform, AWS SageMaker, Oracle Cloud Infrastructure, IBM Analytics), DATA GENERATION & LABELLING (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), MODEL BUILDING (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), FEATURE STORE (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), DEPLOYMENT & PRODUCTION (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), MODEL MONITORING & OBSERVABILITY (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics).
- APPLICATIONS - ENTERPRISE:** Includes SALES (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), MARKETING - B2B (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), MARKETING - B2C (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), CUSTOMER EXPERIENCE / SERVICE (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), HUMAN CAPITAL (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), LEGAL (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), REGTECH & COMPLIANCE (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), FINANCE (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), AUTOMATION & RPA (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), SECURITY (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), PARTNERSHIPS (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud).
- APPLICATIONS - INDUSTRY:** Includes ADVERTISING (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), EDUCATION (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), REAL ESTATE (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), GOVT & INTELLIGENCE (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), COMMERCE (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), FINANCE - LENDING (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), INSURANCE (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud), FINANCE - INVESTING (Salesforce, HubSpot, Microsoft Dynamics 365, SAP Sales Cloud, Oracle Sales Cloud, IBM Sales Cloud).
- Other categories include:** MPP DBs (Teradata, Vertica, Oracle Exadata, SAP HANA, Microsoft Azure Synapse Analytics, Google Cloud BigQuery, AWS Redshift, Oracle Cloud Infrastructure, IBM Cloud Object Storage), REVERSE ETL (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), DATA INTEGRATION (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), DATA GOVERNANCE & ACCESS (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), DATA CATALOG AND DISCOVERY (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), METRICS STORE (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), LOG ANALYTICS (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), COMPUTER VISION (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), SPEECH (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), NLP (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), SYNTHETIC MEDIA (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), HORIZONTAL AI (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), GPU DBS & CLOUD (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics), AI HARDWARE (Alteryx, Tableau, Power BI, Qlik, SAP Analytics Cloud, Microsoft Power BI, Google Analytics, Amazon QuickSight, Oracle Analytics Cloud, IBM Analytics).



Open Lineage

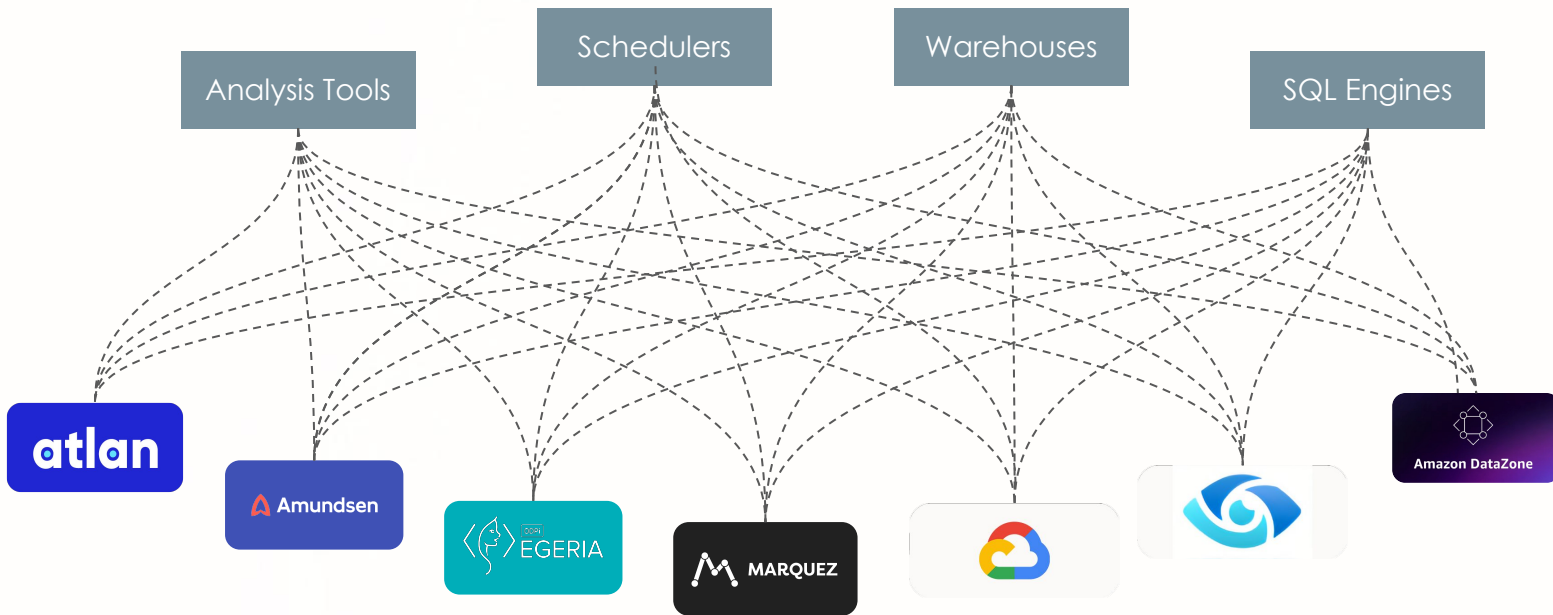
Mission:

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

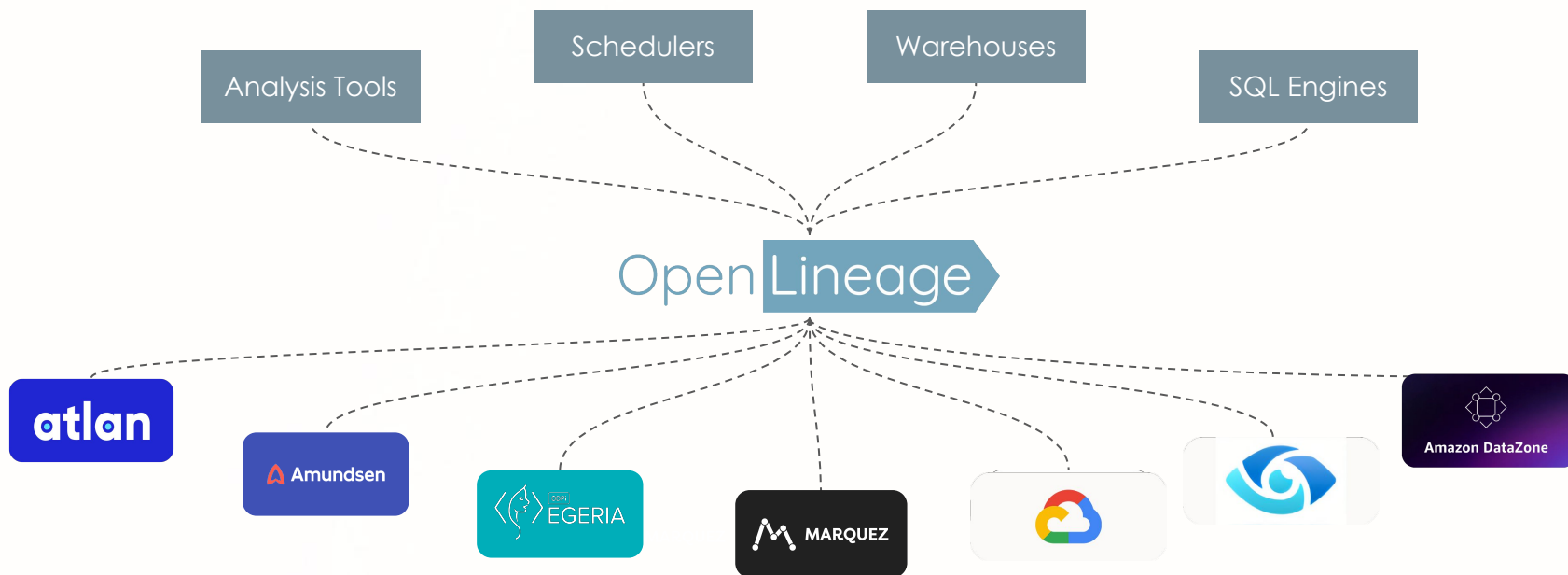
LF AI & DATA



The Data World Without OpenLineage



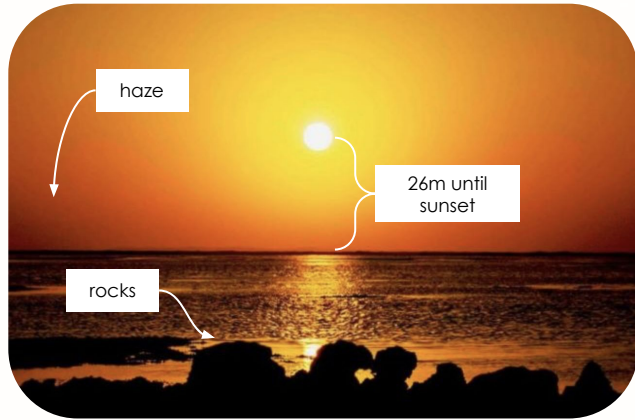
The Data World With OpenLineage



<https://openlineage.io/ecosystem>



Why runtime?



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!



OpenLineage Integrations

Metadata producers



Metadata consumers





OpenLineage Contributors

ASTRONOMER





Astro Observe

ID: cm0o61jps081i01ou1bfvxz6u | ASSETS: 1 | OVERALL SLA HIT RATE: 0% | OWNER: Julian LaNeve | UPDATED: 6 hours ago by Julian LaNeve | CREATED: 6 days ago by Julian LaNeve

Overview | **Graph** | Assets | SLA Evaluations | Details

Graph not filtered

```
graph LR; subgraph Sources; S1[GCS://BUCKET6 data6.json]; S2[GCS://BUCKET5 data5.json]; S3[GCS://BUCKET4 data4.json]; S4[AWS S3://BUCKET3 data3.json]; S5[AWS S3://BUCKET2 data2.json]; S6[AWS S3://BUCKET1 data1.json]; end; subgraph Extract; E1[DATA ENGINEERING (CE OBSERVAB... extract_data6)]; E2[DATA ENGINEERING (CE OBSERVAB... extract_data5)]; E3[DATA ENGINEERING (CE OBSERVAB... extract_data4)]; E4[DATA ENGINEERING (CE OBSERVAB... extract_data3)]; E5[DATA ENGINEERING (CE OBSERVAB... extract_data2)]; E6[DATA ENGINEERING (CE OBSERVAB... extract_data1)]; end; subgraph Transform; T1[DATA ENGINEERING (CE OBSERVAB... transform_bronze_3)]; T2[DATA ENGINEERING (CE OBSERVAB... transform_bronze_2)]; T3[DATA ENGINEERING (CE OBSERVAB... transform_bronze_1)]; T4[DATA ENGINEERING (CE OBSERVAB... transform_silver_2)]; T5[DATA ENGINEERING (CE OBSERVAB... transform_silver_1)]; end; subgraph Destination; D1[SNOWFLAKE://GP21A11US-EAST-1 prod.dwh.gold_1]; end; S1 --> E1; S2 --> E2; S3 --> E3; S4 --> E4; S5 --> E5; S6 --> E6; E1 --> T1; E2 --> T1; E3 --> T1; E4 --> T2; E5 --> T2; E6 --> T2; T1 --> T4; T2 --> T4; T3 --> T5; T4 --> D1; T5 --> D1;
```

Asset	Task	Observed	
GCS://BUCKET6 data6.json	DATA ENGINEERING (CE OBSERVAB... extract_data6) + 3 TASKS	Last observed an hour ago	
GCS://BUCKET5 data5.json	DATA ENGINEERING (CE OBSERVAB... extract_data5) + 3 TASKS	Last observed an hour ago	
GCS://BUCKET4 data4.json	DATA ENGINEERING (CE OBSERVAB... extract_data4) + 3 TASKS	Last observed an hour ago	
AWS S3://BUCKET3 data3.json	DATA ENGINEERING (CE OBSERVAB... extract_data3) + 3 TASKS	Last observed an hour ago	
AWS S3://BUCKET2 data2.json	DATA ENGINEERING (CE OBSERVAB... extract_data2) + 3 TASKS	Last observed an hour ago	
AWS S3://BUCKET1 data1.json	DATA ENGINEERING (CE OBSERVAB... extract_data1) + 3 TASKS	Last observed an hour ago	
...			
DATA ENGINEERING (CE OBSERVAB... transform_bronze_3) + 4 TASKS			Last observed 44 minutes ago
DATA ENGINEERING (CE OBSERVAB... transform_bronze_2) + 4 TASKS			Last observed an hour ago
DATA ENGINEERING (CE OBSERVAB... transform_bronze_1) + 4 TASKS			Last observed 44 minutes ago
DATA ENGINEERING (CE OBSERVAB... transform_silver_2) + 4 TASKS			Last observed 42 minutes ago
DATA ENGINEERING (CE OBSERVAB... transform_silver_1) + 4 TASKS			Last observed 42 minutes ago
DATA ENGINEERING (CE OBSERVAB... transform_gold_1) + 4 TASKS			Last observed 40 minutes ago
SNOWFLAKE://GP21A11US-EAST-1 prod.dwh.gold_1			Last observed 40 minutes ago



Astro Observe

ID: cm0o624mu081m01ounrnn154v | ASSETS: 1 | OVERALL SLA HIT RATE: 93.38% | OWNER: Julian LaNeve | UPDATED: 6 hours ago by Julian LaNeve | CREATED: 6 days ago by Julian LaNeve

Overview | Graph | Assets | SLA Evaluations | Details

Data Product

CURRENT STATUS: **SUCCESS** | LAST UPDATED: Sep 10, 2024 5:11 PM

SLAs

SLA NAME	STATUS	EVALUATED
hourly freshness policy	Miss (0 / 1)	4 hours ago
hourly freshness policy	Miss (0 / 1)	5 hours ago
hourly freshness policy	Miss (0 / 1)	18 hours ago
hourly freshness policy	Miss (0 / 1)	3 days ago
hourly freshness policy	Miss (0 / 1)	4 days ago

1-5 of 5 SLA Evaluations

Assets

SUCCESS: 4 | RUNNING: 0 | FAILED: 0

TYPE	NAME	STATUS	NAMESPACE
No assets currently failed or running.			

Insights

No insights.



OpenLineage Airflow Integration



What does it do

The screenshot shows the Apache Airflow web interface. At the top, there's a navigation bar with 'Airflow' logo and menu items: DAGs, Cluster Activity, Datasets, Security, Browse, Admin, Docs. The current page is 'DAGs'. The breadcrumb is 'snowflake - Grid - Airflow'. The URL is 'localhost:28080/dags/snowflake/grid?dag_run_id=manual__2024-09-05T12%3A20%3A28.495732%2B00%3A'. The date and time are '09 / 05 / 2024 12:20:45 PM'. There are filters for 'All Run Types' and 'All Run States', and a 'Clear Filters' button. The 'Auto-refresh' is set to '25'. Below the filters, there's a legend for task states: deferred, failed, queued, removed, restarting, running, scheduled, skipped, success, up_for_reschedule, up_for_retry, upstream_failed, no_status. The main content area shows a DAG 'snowflake' with a task 'snowflake_insert' running at '2024-09-05, 12:20:28 UTC'. There are buttons for 'Clear task', 'Mark state as...', and 'Filter DAG by task'. Below this, there are tabs for 'Details', 'Graph', 'Gantt', 'Code', 'Event Log', 'Logs', 'XCom', and 'Task Duration'. The 'Code' tab is active, showing the DAG definition. The code is as follows:

```
41     ,
42     dag=dag
43 )
44
45 t2 = SnowflakeOperator(
46     task_id='snowflake_insert',
47     snowflake_conn_id=CONNECTION,
48     sql='''
49     INSERT INTO test_orders (ord, str, num) VALUES
50     (1, 'b', 15),
51     (2, 'a', 21),
52     (3, 'b', 7);
53     '''
54     dag=dag
55 )
56
57 t3 = SnowflakeOperator(
58     task_id='snowflake_truncate',
59     snowflake_conn_id=CONNECTION,
60     sql="TRUNCATE TABLE test_orders;",
61     dag=dag
62 )
63
64
```

On the left side, there's a 'Duration' chart showing the execution time for each task. The tasks are: snowflake_if_not_exists (00:02:44), snowflake_insert (00:01:22), snowflake_truncate (00:00:00), and uwal_sie (00:00:00). The 'snowflake_insert' task is highlighted in blue.



What does it do

The screenshot shows a web browser window with the URL `localhost:23100/lineage/job/airflow/snowflake.snowflake_insert`. The interface features a search bar at the top with the text "Search your Jobs and Datasets...". Below the search bar, there are tabs for "Jobs", "Mode" (set to "Table Level"), "Namespace" (set to "airflow"), and "Name" (set to "snowflake.snowflake_insert"). On the right side, there are controls for "Depth" (set to 2), "Full Graph" (checked), and "Compact Nodes" (unchecked). The main area displays a lineage graph with a single node: a job icon labeled "JOB snowflake.snowfl..." connected to a dataset icon labeled "DATASET PLAYGROUND.SOME...". The dataset node lists its schema as "- ORD", "- STR", and "- NUM". A sidebar on the left contains icons for settings, datasets, and search. A bottom-left corner shows a small thumbnail of the current view and the text "en".

2.7+ AIP-53 Implementation

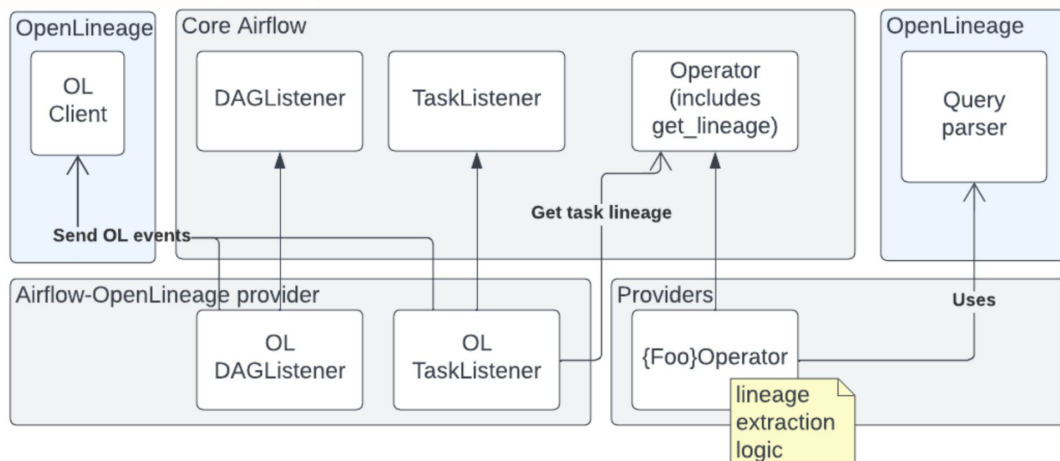
- OpenLineage is part of Airflow since 2.7+ introduced in AIP-53
- Part of implementation happens in Operators
- START, COMPLETE, FAIL states are exposed via different `get_openlineage_facets_*` methods returning `OperatorLineage` class

Ex. on BigQueryToGCSOperator

```
302     def get_openlineage_facets_on_complete(self, task_instance):
303         parsed_uri = urllib.parse.urlparse(self.destination_cloud_storage_uris[0])
304         return OperatorLineage(
305             inputs=[Dataset(
306                 namespace="bigquery",
307                 name=f"{self.conf['datasetId']}.{self.conf['projectId']}.{self.conf['tableId']}"),
308             ],
309             outputs=[
310                 # Caveat: single URI
311                 Dataset(namespace=f"gs://{parsed_uri.hostname}", name=parsed_uri.path)
312             ]
313         )
314
```

How does it work

- OpenLineage Provider has listener that gets called by Airflow by Listener API
 - On worker, before task runs, and after task completes or fails
- Checks if Operator has implemented lineage methods, calls them
- Enrich data returned from Operator with common Airflow facets
- More on that in last year's presentation <https://youtu.be/SZBVgREqets>





Supported Operators

https://airflow.apache.org/docs/apache-airflow-providers-openlineage/stable/supported_classes.html

Caveat - does not make sense for OpenLineage to cover all Operators

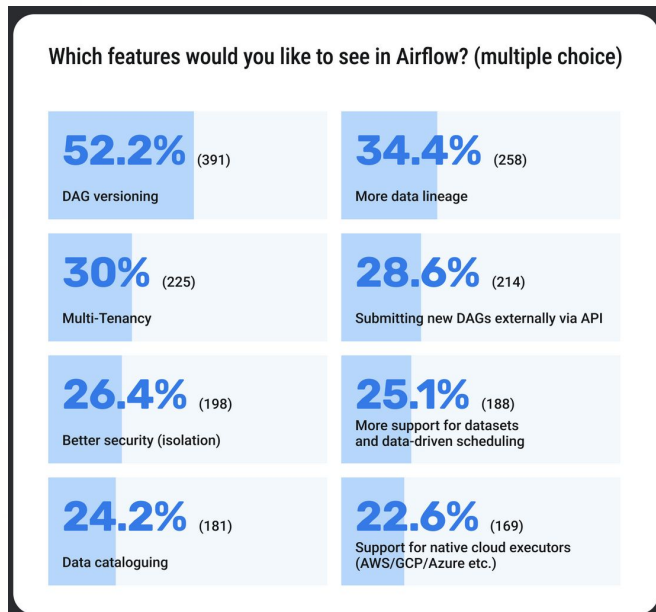
The screenshot shows the Apache Airflow documentation page for "Supported operators". The page title is "Supported operators". Below the title, there is a paragraph: "Below is a list of Operators that support OpenLineage extraction, along with specific DB types that are compatible with the SQLExecuteQueryOperator." There are two callout boxes: a blue one labeled "Important" stating "While we strive to keep the list of supported operators current, please be aware that our updating process is automated and may not always capture everything accurately." and a green one labeled "Tip" stating "You can easily implement OpenLineage support for any operator. See [Implementing OpenLineage in Operators](#)." The page is divided into sections: "Core operators" (with a note that two core operators support OpenLineage) and "SQLExecuteQueryOperator" (with a note that it uses SQL parsing for lineage extraction). A sidebar on the right lists "Supported operators" under "Provider's operators", including: apache-airflow-providers-amazon, apache-airflow-providers-common-io, apache-airflow-providers-dbt-cloud, apache-airflow-providers-ftp, apache-airflow-providers-google, apache-airflow-providers-microsoft-azure, apache-airflow-providers-sftp, and apache-airflow-providers-snowflake. The left sidebar contains navigation links for BASICS, GUIDES, REFERENCES, and RESOURCES.



Is this enough?



Airflow Survey 2023 results





How people use Airflow?

what operators are most popular in Apache Airflow?



Here are some of the most popular operators in Apache Airflow:

1. PythonOperator

- **Description:** Executes Python functions or callables, offering great flexibility and is widely used for various tasks.
- **Use Cases:** Data processing, API integrations, and executing custom Python logic.



TaskFlow API

```
1 from airflow.decorators import dag, task
2 from airflow.utils.dates import days_ago
3 from airflow.providers.amazon.aws.hooks.s3 import S3Hook
4
5 @dag(schedule_interval=None, start_date=days_ago(1), catchup=False)
6 def s3_data_processing_dag():
7
8     @task
9     def read_data_from_s3(bucket_name: str, key: str) -> str:
10         s3_hook = S3Hook(aws_conn_id='my_aws_conn')
11         data = s3_hook.read_key(key=key, bucket_name=bucket_name)
12         return data
13
14     @task
15     def process_data(data: str) -> str:
16         processed_data = data.upper()
17         return processed_data
18
19     @task
20     def write_data_to_s3(bucket_name: str, key: str, data: str):
21         s3_hook = S3Hook(aws_conn_id='my_aws_conn')
22         s3_hook.load_string(string_data=data, key=key, bucket_name=bucket_name, replace=True)
23
24     raw_data = read_data_from_s3(bucket_name='my_bucket', key='input_data.txt')
25     processed_data = process_data(raw_data)
26     write_data_to_s3(bucket_name='my_bucket', key='output_data.txt', data=processed_data)
```



Object Storage

```
1 from __future__ import annotations
2 from datetime import datetime
3
4 from airflow.decorators import dag, task
5 from airflow.io.path import ObjectStoragePath
6
7
8 base = ObjectStoragePath("s3://aws_default@openlineage-test/")
9
10
11 @dag(
12     schedule=None,
13     start_date=datetime(2022, 1, 1),
14     catchup=False,
15 )
16 def objectstorage_lineage():
17     @task
18     def read_s3_data(**kwargs):
19         from pandas import DataFrame as pd
20         storage_path = base / "input" / "dataset"
21         with storage_path.open("r") as file:
22             data = file.read()
23
24         df = pd.DataFrame(data)
25         return df
26
27     @task
28     def transform_data(df):
29         ...
30
31     @task
32     def write_to_s3(df):
33         storage_path = base / "output" / "dataset"
34         with storage_path.open("w") as file:
35             df.to_parquet(file)
36
37     data = read_s3_data()
38     transformed_data = transform_data(data)
39     write_to_s3(transformed_data)
```



Let's look at another DAG

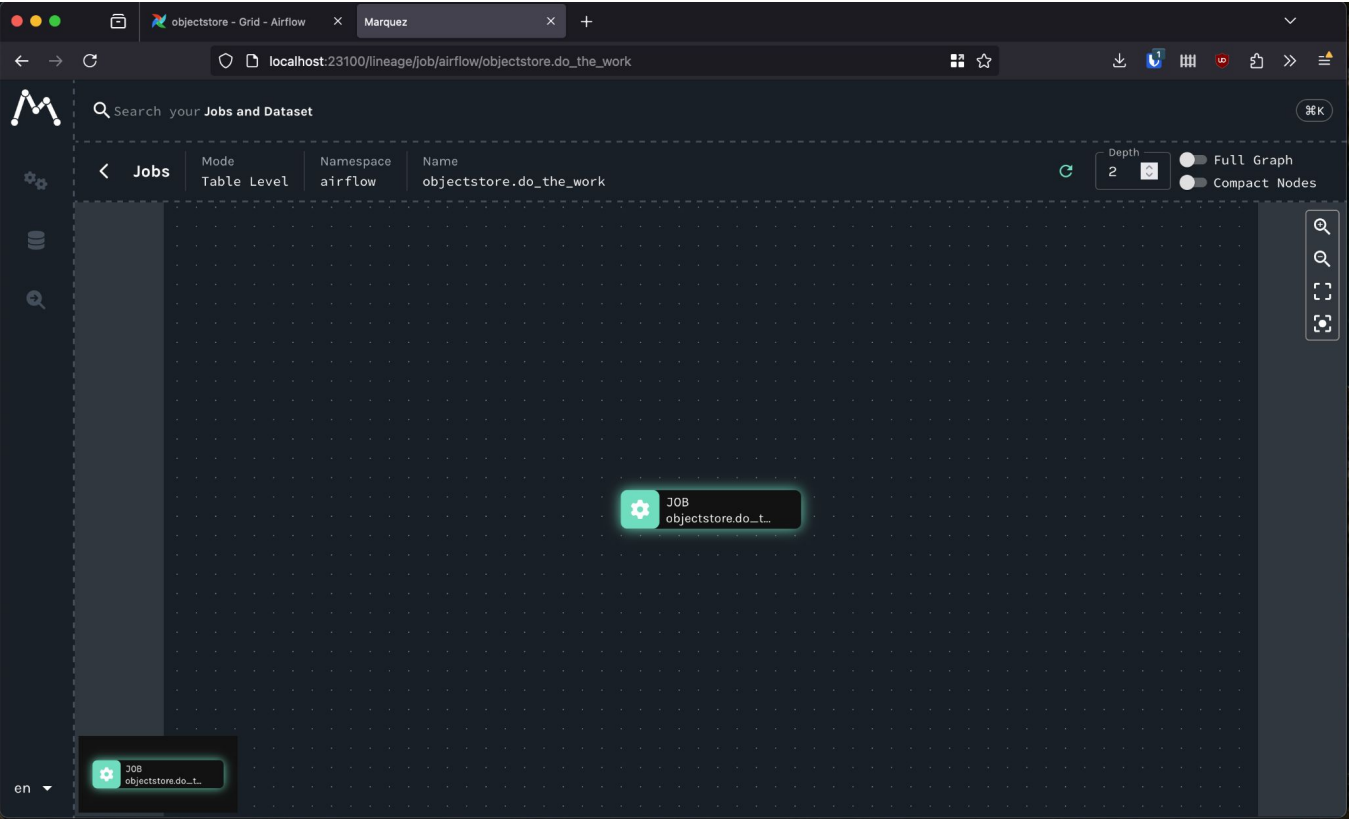
Airflow 2.9.3

The screenshot shows the Airflow web interface for a DAG named 'do_the_work'. The interface includes a navigation bar with options like 'DAGs', 'Cluster Activity', 'Datasets', 'Security', 'Browse', 'Admin', and 'Docs'. The current view is for a DAG run on 2024-09-05 at 12:40:54 UTC. The DAG is in a 'running' state. The code editor shows the following Python code:

```
31 @dag(
32     dag_id="objectstore",
33     start_date=datetime(2022, 1, 1),
34     schedule=None
35 )
36 def taskflow_object():
37     @task
38     def do_the_work(*args, **kwargs):
39         import pandas as pd
40         import numpy as np
41         # ensure the bucket exists
42         base.mkdir(exist_ok=True)
43         execution_date = kwargs["logical_date"]
44         formatted_date = execution_date.format("YYYYMMDD")
45
46         df = pd.DataFrame(np.random.randint(0, 100, size=(100, 4)), columns=list('ABCD'))
47         path = base / f"air_quality_{formatted_date}.parquet"
48
49         with path.open("wb") as file:
50             df.to_parquet(file)
51
52     return path
53 do_the_work()
```



Let's look at another DAG



The problem: arbitrary code

- People love writing their own code!
- Non-code operators: we generally know what they do (*)
- Code operators: can be anything!
- Is there anything we can do?

```
1 from airflow import DAG
2 from airflow.operators.python import PythonOperator
3 from datetime import datetime
4 from my_awesome_code import do_the_work
5
6 def extract_load_transform(**kwargs):
7     # What happens there?
8     do_the_work()
9
10 default_args = {
11     'owner': 'airflow',
12     'start_date': datetime(2023, 9, 1),
13     'retries': 1,
14 }
15
16 with DAG(
17     dag_id='example_python_operator',
18     default_args=default_args,
19     schedule_interval='@daily',
20     catchup=False,
21 ) as dag:
22     the_task = PythonOperator(
23         task_id='extract_load_transform',
24         python_callable=extract_load_transform,
25         provide_context=True,
26     )
```



Instrument the hooks



What are hooks?

- Hooks - a mechanism for communicating with external systems
- Operations on hooks closely resemble those exposed by external systems
- Ex. copy file between object storage buckets, execute SQL on some database, or even create a Kubernetes job

```
def copy(
    self,
    source_bucket: str,
    source_object: str,
    destination_bucket: str | None = None,
    destination_object: str | None = None,
) -> None:
    """
    Copy an object from a bucket to another, with renaming if requested.

    destination_bucket or destination_object can be omitted, in which case
    source bucket/object is used, but not both.

    :param source_bucket: The bucket of the object to copy from.
    :param source_object: The object to copy.
    :param destination_bucket: The destination of the object to copied to.
    | Can be omitted; then the same bucket is used.
    :param destination_object: The (renamed) path of the object if given.
    | Can be omitted; then the same name is used.
    """
    destination_bucket = destination_bucket or source_bucket
    destination_object = destination_object or source_object

    if source_bucket == destination_bucket and source_object == destination_object:
        raise ValueError(
            f"Either source/destination bucket or source/destination object must be different, "
            f"not both the same: bucket={source_bucket}, object={source_object}"
        )
    if not source_bucket or not source_object:
        raise ValueError("source_bucket and source_object cannot be empty.")

    client = self.get_conn()
    source_bucket = client.bucket(source_bucket)
    source_object = source_bucket.blob(source_object) # type: ignore[attr-defined]
    destination_bucket = client.bucket(destination_bucket)
    destination_object = source_bucket.copy_blob( # type: ignore[attr-defined]
        blob=source_object, destination_bucket=destination_bucket, new_name=destination_object
    )

    self.log.info(
        "Object %s in bucket %s copied to object %s in bucket %s",
        source_object.name, # type: ignore[attr-defined]
        source_bucket.name, # type: ignore[attr-defined]
        destination_object.name, # type: ignore[union-attr]
        destination_bucket.name, # type: ignore[union-attr]
    )
```



How to instrument hooks?

- We can't copy the "pull" approach we have for Operators
- One method vs many methods
- We can "push" it from the instrumented method
- Where to?

```
def copy(
    self,
    source_bucket: str,
    source_object: str,
    destination_bucket: str | None = None,
    destination_object: str | None = None,
) -> None:
    """
    Copy an object from a bucket to another, with renaming if requested.

    destination_bucket or destination_object can be omitted, in which case
    source bucket/object is used, but not both.

    :param source_bucket: The bucket of the object to copy from.
    :param source_object: The object to copy.
    :param destination_bucket: The destination of the object to copied to.
        Can be omitted; then the same bucket is used.
    :param destination_object: The (renamed) path of the object if given.
        Can be omitted; then the same name is used.
    """
    destination_bucket = destination_bucket or source_bucket
    destination_object = destination_object or source_object

    if source_bucket == destination_bucket and source_object == destination_object:
        raise ValueError(
            f"Either source/destination bucket or source/destination object must be different, "
            f"not both the same: bucket={source_bucket}, object={source_object}"
        )
    if not source_bucket or not source_object:
        raise ValueError("source_bucket and source_object cannot be empty.")

    client = self.get_conn()
    source_bucket = client.bucket(source_bucket)
    source_object = source_bucket.blob(source_object) # type: ignore[attr-defined]
    destination_bucket = client.bucket(destination_bucket)
    destination_object = source_bucket.copy_blob( # type: ignore[attr-defined]
        blob=source_object, destination_bucket=destination_bucket, new_name=destination_object
    )

    self.log.info(
        "Object %s in bucket %s copied to object %s in bucket %s",
        source_object.name, # type: ignore[attr-defined]
        source_bucket.name, # type: ignore[attr-defined]
        destination_object.name, # type: ignore[union-attr]
        destination_bucket.name, # type: ignore[union-attr]
    )
)
```


What does the instrumentation look like?

- Added HookLineageCollector on Worker with .add_input_dataset and .add_output_dataset methods
- Those methods accept dataset_kwargs that are used to construct the compliant dataset later
- Methods in hooks that modify datasets call those methods to register dataset changes
- Data is then deduplicated - we don't want to see hundreds of writes to same dataset

```
@unify_bucket_name_and_key
@provide_bucket_name
def load_file(
    self,
    filename: Path | str,
    key: str,
    bucket_name: str | None = None,
    replace: bool = False,
    encrypt: bool = False,
    gzip: bool = False,
    acl_policy: str | None = None,
) -> None:
    filename = str(filename)
    if not replace and self.check_for_key(key, bucket_name):
        raise ValueError(f"The key {key} already exists.")

    extra_args = self.extra_args
    if encrypt:
        extra_args["ServerSideEncryption"] = "AES256"
    if gzip:
        with open(filename, "rb") as f_in:
            filename_gz = f"{f_in.name}.gz"
            with gz.open(filename_gz, "wb") as f_out:
                shutil.copyfileobj(f_in, f_out)
            filename = filename_gz
    if acl_policy:
        extra_args["ACL"] = acl_policy

    client = self.get_conn()
    client.upload_file(filename, bucket_name, key, ExtraArgs=extra_args, Config=self.transfer_conf)
    get_hook_lineage_collector().add_input_dataset(
        context=self, scheme="file", dataset_kwargs={"path": filename}
    )
    get_hook_lineage_collector().add_output_dataset(
        context=self, scheme="s3", dataset_kwargs={"bucket": bucket_name, "key": key}
    )
```

What does the instrumentation look like?

- Object Storage: track reads and writes across file-like objects
- No additional instrumentation needed for different types of paths

```
44 class TrackingFileWrapper(LoggingMixin):
45     """Wrapper that tracks file operations to intercept lineage."""
46
47     def __init__(self, path: ObjectStoragePath, obj):
48         super().__init__()
49         self._path: ObjectStoragePath = path
50         self._obj = obj
51
52     def __getattr__(self, name):
53         attr = getattr(self._obj, name)
54         if callable(attr):
55             # If the attribute is a method, wrap it in another method to intercept the call
56             def wrapper(*args, **kwargs):
57                 self.log.debug("Calling method: %s", name)
58                 if name == "read":
59                     get_hook_lineage_collector().add_input_dataset(context=self._path, uri=str(self._path))
60                 elif name == "write":
61                     get_hook_lineage_collector().add_output_dataset(context=self._path, uri=str(self._path))
62                 result = attr(*args, **kwargs)
63                 return result
64
65             return wrapper
66         return attr
67
68     def __getitem__(self, key):
69         # Intercept item access
70         return self._obj[key]
71
72     def __enter__(self):
73         self._obj.__enter__()
74         return self
75
76     def __exit__(self, exc_type, exc_val, exc_tb):
77         self._obj.__exit__(exc_type, exc_val, exc_tb)
```

Works only if you use it

- Not OpenLineage specific - you can write your own plugin, register HookLineageReader and use the gathered hook lineage for your own purposes
- If there are no HookLineageReaders, the .add_input_datasets and .add_output_datasets just send data to /dev/null

```
class HookLineageReader(LoggingMixin):
    """Class used to retrieve the hook lineage information collected by HookLineageCollector."""

    def __init__(self, **kwargs):
        self.lineage_collector = get_hook_lineage_collector()

    def retrieve_hook_lineage(self) -> HookLineage:
        """Retrieve hook lineage from HookLineageCollector."""
        hook_lineage = self.lineage_collector.collected_datasets
        return hook_lineage

def get_hook_lineage_collector() -> HookLineageCollector:
    """Get singleton lineage collector."""
    global _hook_lineage_collector
    if not _hook_lineage_collector:
        from airflow import plugins_manager

        plugins_manager.initialize_hook_lineage_readers_plugins()
        if plugins_manager.hook_lineage_reader_classes:
            _hook_lineage_collector = HookLineageCollector()
        else:
            _hook_lineage_collector = NoOpCollector()
    return _hook_lineage_collector
```

AIP-60 compliant datasets

- Operators don't always work on datasets they know - GCSToS3Operator
- Object Storage does not explicitly know what the file is, it operates on abstract files
- AIP-60 Dataset is an URI, OpenLineage dataset is name+namespace
- Providers know the dataset abstractions they own
- Providers can provide good factory method for AIP-60 URI, and how to translate it to OpenLineage dataset

```
dataset-uris:  
- schemes: [s3]  
  handler: airflow.providers.amazon.aws.datasets.s3.sanitize_uri  
  to_openlineage_converter: airflow.providers.amazon.aws.datasets.s3.convert_dataset_to_openlineage  
  factory: airflow.providers.amazon.aws.datasets.s3.create_dataset
```



Same Object Storage DAG in Airflow 2.10

The screenshot shows the Airflow web interface in a browser window. The browser address bar shows the URL: `localhost:28080/dags/objectstore/grid?dag_run_id=manual__2024-09-06T13%3A01`. The interface includes a navigation bar with tabs for DAGs, Cluster Activity, Datasets, Security, Browse, Admin, and Docs. The current view is for a DAG named 'do_the_work', which is shown as a single task in a Gantt chart on the left. The main area displays the DAG's code in a code editor, with the 'Code' tab selected. The code is as follows:

```
34     schedule=None
35 )
36 def taskflow_object():
37     @task
38     def do_the_work(*args, **kwargs):
39         import pandas as pd
40         import numpy as np
41         # ensure the bucket exists
42         base.mkdir(exist_ok=True)
43         execution_date = kwargs["logical_date"]
44         formatted_date = execution_date.format("YYYYMMDD")
45
46         df = pd.DataFrame(np.random.randint(0, 100, size=(100, 4)), columns=
47         path = base / f"air_quality_{formatted_date}.parquet"
48
49         with path.open("wb") as file:
50             df.to_parquet(file)
51
52         return path
53     do_the_work()
54
```



Same Object Storage DAG in Airflow 2.10

The screenshot shows the Airflow web interface in a browser window. The address bar indicates the URL is `localhost:23100/lineage/job/airflow/objectstore.do_the_work`. The interface features a search bar at the top with the text "Search your Jobs and Datasets...". Below the search bar, there is a table with the following columns: "Mode", "Namespace", and "Name". The table contains one entry: "Table Level", "airflow", and "objectstore.do_the_work". To the right of the table, there are controls for "Depth" (set to 2), "Full Graph" (checked), and "Compact Nodes" (unchecked). The main area of the interface displays a DAG graph with a single job node labeled "JOB objectstore.do_t..." and a dataset node labeled "DATASET airflow-objects...". The job node is highlighted with a green glow. The interface also includes a sidebar with navigation icons and a bottom status bar.



Lineage For Python Tasks

- It does not matter how you author your Python tasks, it captures hook lineage whether you use PythonOperator, TaskFlow or Custom Operators
- However, it has to be the same process as worker - python-based KubernetesOperator job won't work



Peek into the future

As in, what's happening with OL in Airflow 3?



Datasets Assets

- Despite all those changes, not everything can be detected.
- OpenLineage integration will be able to take advantage of that and expose annotated lineage

[Pages](#) / [Airflow Home](#) / [Airflow Improvement Proposals](#)

AIP-73 Expanded Data Awareness

Created by Constance Martineau, last modified by Tzu-ping Chung on Aug 02, 2024

Status

State	Accepted
Discussion Thread	https://lists.apache.org/thread/6rp4jhflwg3czhtvjszoctdry85vfv8r
Vote Thread	https://lists.apache.org/thread/9570fr5b2jv6hb2fd5z43jmsws42ls1z
Vote Result Thread	https://lists.apache.org/thread/f0rvbdbpq7bylt3kv0v6gn90qr3f98ng
Progress Tacking (PR/GitHub Project/Issue Label)	
Date Created	2024-06-26
Version Released	
Authors	@Constance Martineau @Tzu-ping Chung

Motivation

Airflow has become the standard for orchestrating complex data workflows. However, it operates with limited visibility into the actual data it processes or produces. While it understands task execution order and attributes like operators and parameters in use, it lacks insight into the nature of data inputs and outputs. This link between data and tasks is fundamental to data engineering, and vital for providing insights into the state and health of data as they move through the workflow. Orchestrators are the heart of data platforms, and if they can understand this link, they can make orchestration decisions based on data quality and freshness, while also providing data engineers with insights about system and data reliability in one place.



Datasets Assets

Seeing Clearly with Airflow: Bridging Task-Centric and Data-Aware Orchestration

Speaker(s):




[Constance Martineau](#)



[Tzu Ping Chung](#)

Sep-12 12:30-13:15 in Elizabethan A+B

 [Add to Calendar](#)

As Apache Airflow evolves, a key shift is emerging: the move from task-centric to data-aware orchestration. Traditionally, Airflow has focused on managing tasks efficiently, with limited visibility into the data those tasks manipulate. However, the rise of data-centric workflows demands a new approach—one that puts data at the forefront.

This talk will explore how embedding deeper data insights into Airflow can align with modern users' needs, reducing complexity and enhancing workflow efficiency. We'll discuss how this evolution can transform Airflow into a more intuitive and powerful tool, better suited to today's data-driven environments.



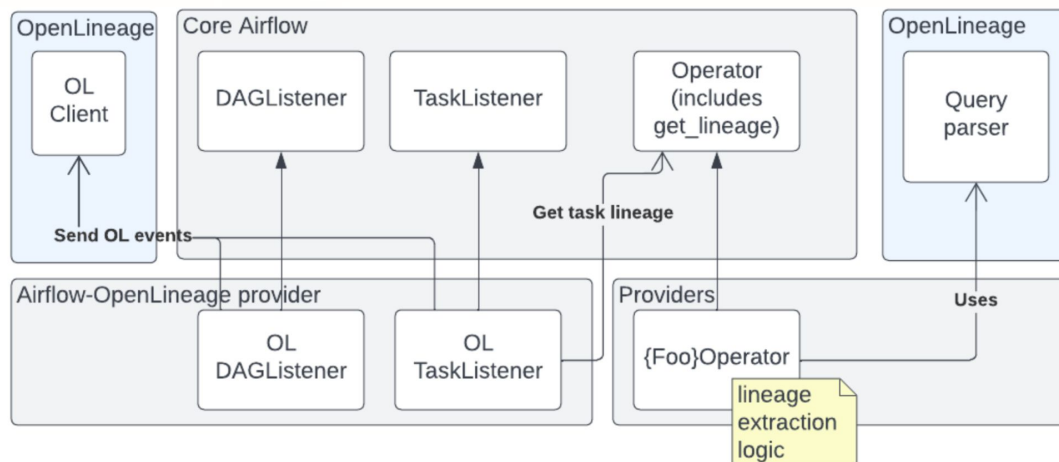
Synergy

- Building on Airflow - using Connections, Hooks, Assets, Object Storage gives you much more than using Airflow as a “dumb” scheduler



AIP-72 Task Execution Interface

- Decoupling access to database from Worker requires us to rework approach for integration
- Below diagram won't be accurate: worker after completion will send metadata using AIP-72 interface, and OL will process it asynchronously



AIP-72 Task Execution Interface

- Gains:
 - Async event emission: does not block a worker slot
 - Isolated execution: won't affect worker if some bug happen
 - Possible to enrich the data more due to asynchronous execution
- However, it's possible that contributing OpenLineage Operator implementations will get more complex



What AIP-72 enables

Running Airflow Tasks Anywhere, in any Language

Speaker(s):




[Ash Berlin-Taylor](#)



[Vikram Koka](#)

Sep-10 12:00-12:45 in Elizabethan A+B

 [Add to Calendar](#)

Imagine a world where writing Airflow tasks in languages like Go, R, Julia, or maybe even Rust is not just a dream but a native capability. Say goodbye to BashOperators; welcome to the future of Airflow task execution.

Here's what you can expect to learn from this session:

- **Multilingual Tasks:** Explore how we empower DAG authors to write tasks in any language while retaining seamless access to Airflow Variables and Connections.
- **Simplified Development and Testing:** Discover how a standardized interface for task execution promises to streamline development efforts and elevate code maintainability.
- **Enhanced Scalability and Remote Workers:** Learn how enabling tasks to run on remote workers opens up possibilities for seamless deployment on diverse platforms, including Windows and remote Spark or Ray clusters. Experience the convenience of effortless deployments as we unlock new avenues for Airflow usage.

Join us as we embark on an exploratory journey to shape the future of Airflow task execution. Your insights and contributions are invaluable as we refine this vision together. Let's chart a course towards a more versatile, efficient, and accessible Airflow ecosystem.



Related event: OpenLineage Meetup

Open Lineage

Get more details and the signup link at <https://openlineage.io/blog>

Where

Astronomer
8 California St.

When

Thursday, 9/12
6-9 pm

**Join us for in-depth
talks & discussion over
dinner!**

Agenda

- **Unlocking Data Products with OpenLineage at Astronomer:** Julian LaNeve and Jason Ma, Astronomer.
- **OpenLineage: From Operators to Hooks** by Maciej Obuchowski, Astronomer+GetInData/Xebia.
- **Activating Operational Metadata with Airflow, Atlan and Openlineage** by Kacper Muda, GetInData/Xebia.
- **Hamilton, a Scaffold for all Your Python Platform Concerns (and a New OpenLineage Producer)** by Stefan Krawczyk
- **Lightning Talk on New Marquez Features and the Marquez Project Roadmap** by Willy Lulciuc, Marquez Lead, and Peter Hicks, Marquez Committer.



Thank you!
Any questions?



Bonus slides



Consistent API introduction

- Adding new APIs is easy, right?
 - Add new method to Airflow Core
 - Make providers use it
 - Release it all
 - Then it breaks on older versions of Airflow

Consistent API introduction

- Adding new APIs is easy, right?
 - Add new method to Airflow Core
 - Make providers use it
 - Release it all
 - Then it breaks on older versions of Airflow
- Make provider check Airflow version, and perform an action only if version matches the requirement
 - Requires potentially a lot of duplicated work across providers

Consistent API introduction

- Adding new APIs is easy, right?
 - Add new method to Airflow Core
 - Make providers use it
 - Release it all
 - Then it breaks on older versions of Airflow
- Make provider check Airflow version, and perform an action only if version matches the requirement
 - Requires potentially a lot of duplicated work across providers
- Airflow's solution: common.compat provider
 - Introduce a method within compat provider
 - Method should perform version check, and have fallback for older version