Cross-Platform Data Lineage with OpenLineage

Tracing lineage in Spark and Airflow

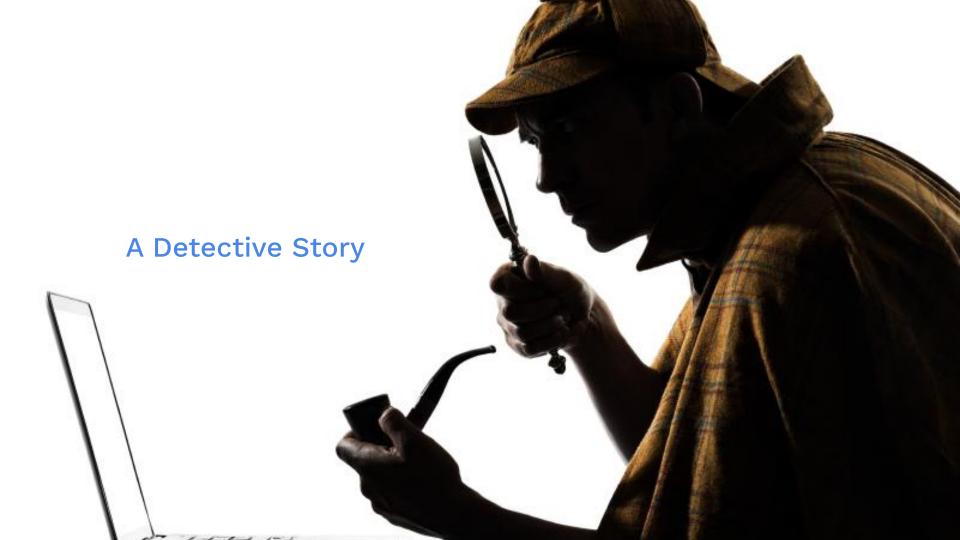


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Who am I?

- Staff Software Engineer at Astronomer
- Previously data infrastructure at Amazon and Cruise
- Clickstream data ingestion and analysis
- A/B test data analysis
- Data catalogs
- Query tools and libraries











Us Cooks Recommendations

Orders near you

Your neighborhood's latest orders



Kaffa Ethiopian Restaurant

2 items • \$23.98 • 1 min ago





Veggie Grill - Downtown Seattle 3 items • \$30.00 • 1 min ago 450



Sharetea - Renton

1 item • \$8.25 • 1 min ago



Today's offers



Aloha Plates \$1.99 Delivery Fee • 50-60 min





Huckleberry Square Restaurant \$0.49 Delivery Fee • 35-45 min



Taqueria los Potrillos - Seattle, ... \$2.99 Delivery Fee • 40-50 mii 2 2 2 2 1178

Us Cooks Recommendations

Metrics used to determine recommendations include

- Star ratings
- # of reviews
- Delivery time
- Repeat customers
- Regular and sale price

Menu Item Normalization

- Same menu items, different names
 - Pâté of roasted indigenous legumes, paired with a compote of seasonal berries, served on hearty sprouted wheat bread
 - Pan-roasted pastry rolls, layered with an herbed tomato puree, a creamy blend of artisanal cheeses, and tender bites of aged salami
 - Minced beef shoulder and caramelized onions in a sweet tomato and vinegar reduction served on a fresh brioche roll
- Restaurant rebranding
- Same chef, different theme







Goodhart's Law

"Any observed statistical regularity will tend to collapse once pressure is placed upon it for control purposes."

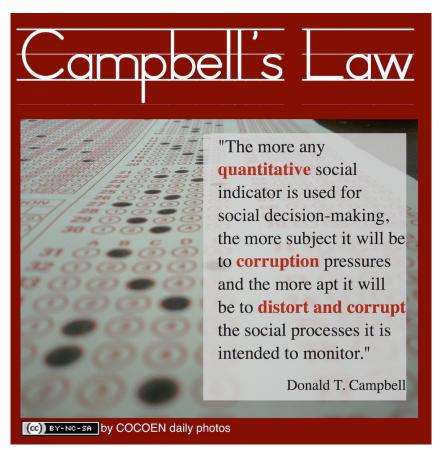
- Charles Goodhart

That is:

When a measure becomes a target, it ceases to be a good measure.



Campbell's Law



Data Manipulation in Us Cooks

Algorithmically driving traffic to:

- New restaurants
- Highly reviewed restaurants
- Up and coming restaurants
- Popular orders
- Recommended orders

Leads to:

- Restaurants generating fake orders
- Restaurants leaving fake reviews
- Fake tweets and retweets
- Wholesalers generating fake reviews
- Fake review company services

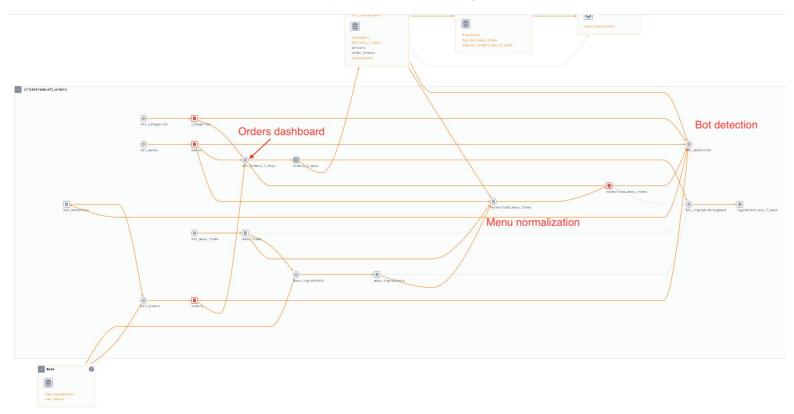
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Data Manipulation in Us Cooks

Counterattack:

- De-fancification
- Bot detection
- Fake order detection
- Paid review detection
- Fake repeat customer detection

Data Dependency Graph

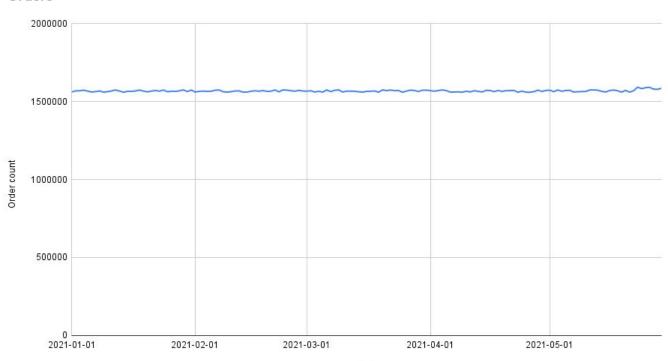


The culprit

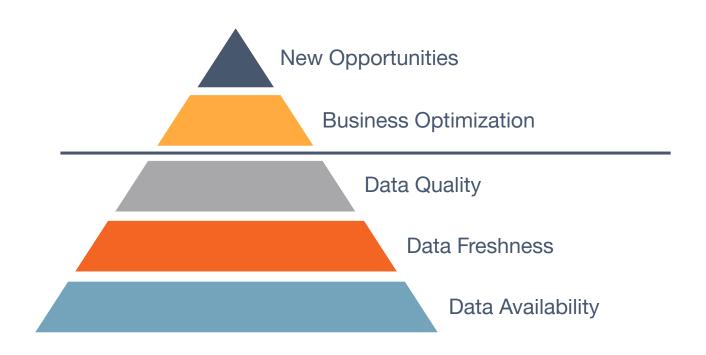


The Fixed Chart

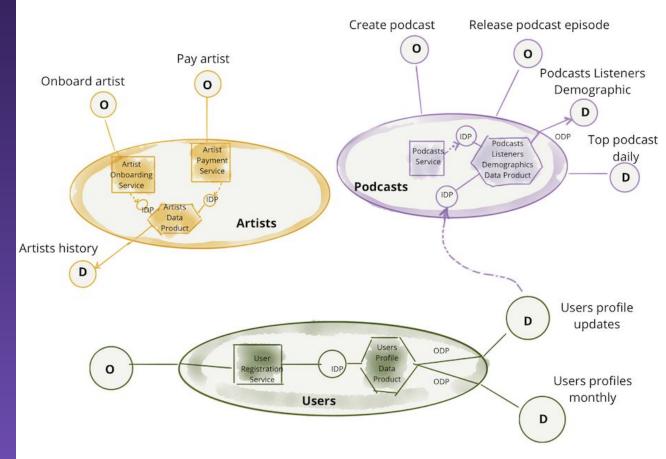




Maslow's Data hierarchy of needs



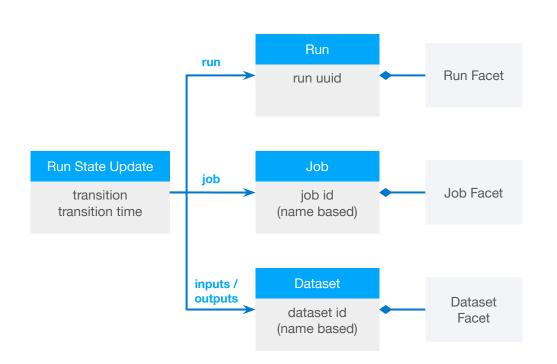
Lineage problems
scale with the number
of teams and
interdependent
datasets, NOT with the
size of the data



OpenLineage Primer



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)

Jobs Hierarchy

Jobs are hierarchical

- Parent jobs may have no inputs or outputs
- Child jobs inherit the namespace of their parents

Examples:

- Scheduler -> Task
- Single application -> multiple operations
- DAG -> Task -> External job

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Facet Examples

Dataset:

- Input/Output statistics
- Schema
- Version

Job:

- Parent job
- Source code location
- Query plan
- Source control

Run:

- Scheduled time
- Batch ID
- Parameters
- Cluster properties

Collecting Lineage

OpenLineage in Airflow and Spark



Airflow support

Airflow 1.10.x

```
- from airflow import DAG
+ from openlineage.airflow import DAG
```

Airflow 2.x

```
AIRFLOW__LINEAGE__BACKEND=openlineage.lineage_backend.OpenLineageBackend
```

• Airflow 2.3+

"airflow.plugins": ["OpenLineagePlugin = openlineage.airflow.plugin:OpenLineagePlugin]"

Airflow Operator Support

- Relies on lineage data exposed by the operators
 - BigQueryOperator
 - SnowflakeOperator
 - PostgresOperator
 - MySQL
 - o dbt
 - GreatExpectations
- Custom Extractors
- PythonOperator and BashOperator unsupported

Custom Extractor

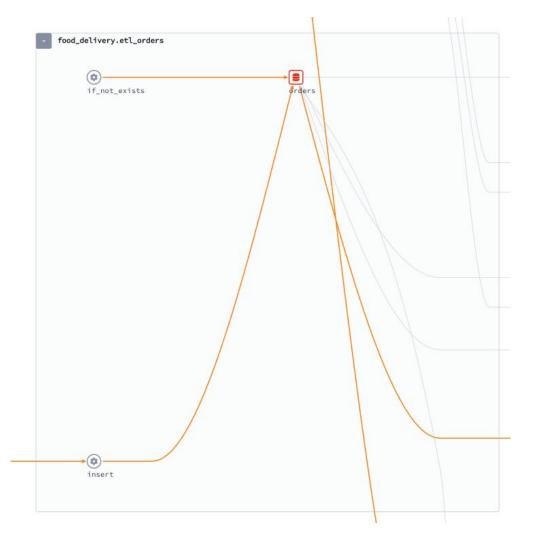
```
class CustomExtractor(BaseExtractor):
   @classmethod
   def get_operator_classnames(cls) -> List[str]:
       return ['CustomOperator']
   def extract(self) -> Union[Optional[TaskMetadata], List[TaskMetadata]]:
        return TaskMetadata(
            "test",
            inputs=[
                Dataset(
                    namespace="test",
                    name="dataset",
                    facets={}
```

Example DAG

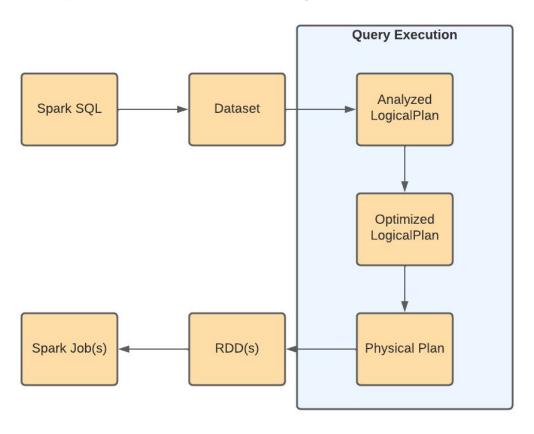
- DAG becomes a parent job (no inputs/outputs)
- Tasks are (typically) jobs with inputs/outputs
- Task job names are prefixed with DAG name

```
dag = DAG(
    'etl_orders',
    schedule_interval='@hourly',
   catchup=False,
   default_args=default_args,
    description='Loads newly placed orders daily.'
t1 = BigQueryOperator(
    task_id='if_not_exists',
   sal='''
   CREATE TABLE IF NOT EXISTS food_delivery.orders (
     id
                  INT64,
     placed on
                  TIME.
     menu_item_id INT64,
     quantity
                  INT64,
     discount_id INT64,
                  STRING
      comment
   use_legacy_sql=False,
   dag=dag
t2 = BigQueryOperator(
   task_id='insert',
   sal='''
   SELECT id, placed_on, menu_item_id, quantity, discount_id, comment
     FROM food_delivery.tmp_orders;
```

Example DAG



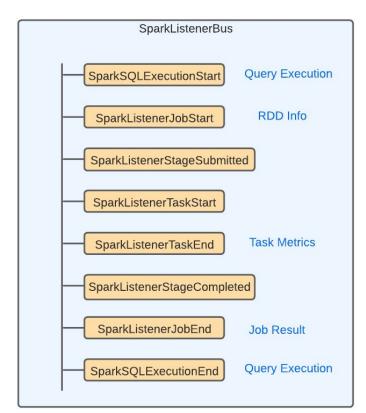
Spark SQL Query Execution



Spark ListenerBus

We implement the SparkListener interface and capture information from the Spark plan

- Serialized Optimized LogicalPlan
- Task input/output metrics
 - Bytes
 - Row counts
- Dataset Metadata
 - Location
 - Schema
 - Version (when available)
- Environment metadata
 - Databricks cluster properties



Configuring Spark

```
SparkSession.builder \
    .config('spark.jars.packages', 'io.openlineage:openlineage-spark:0.6.+') \
    .config('spark.extraListeners', 'io.openlineage.spark.agent.OpenLineageSparkListener') \
    .config('spark.openlineage.host', 'https://api.demo.datakin.com') \
    .config('spark.openlineage.apiKey', 'your datakin api key') \
    .config('spark.openlineage.namespace', '<NAMESPACE_NAME>') \
    .getOrCreate()
```

Example Spark Job

```
spark = SparkSession\
  .builder \
  .appName("recommended_menu_items") \
  .getOrCreate()
GCS sources
orders = spark.read.parquet("gs://bq-airflow-spark/data/orders")
deliveries = spark.read.parquet("gs://bq-airflow-spark/data/top_delivery_times")
nexthour = datetime.now() + timedelta(hours=1)
(orders.join(deliveries, orders.id == deliveries.order_id, how='inner')
 .withColumn('hour', date_format(date_trunc('hour', from_unixtime('order_placed_on')), 'HH'))
  .filter(f"hour == '{nexthour.hour}'")
  .select(deliveries.menu_item_id, 'quantity')
  .groupBy('menu_item_id')
  .agg(sum('quantity').alias('quantity'))
  .sort(col('quantity').desc())
  .limit(20)
  .write.mode(saveMode='overwrite') GCS out
  .parquet(gs://bq-airflow-spark/data/uproming_order_items'))
                                                    and back in
(spark.read.parquet('gs://bq-airflow-spark/data/upcoming_order_items')
 .write.format('jdbc')
 .mode(saveMode='overwrite') Postgres output
 .option("url", / jdbc:postgresql://localhost:5432/spark
 .option('driver', 'org.postgresql.Driver')
 .option("dbtable", "public.upcoming_order_items"
 .option("user", "spark_user")
 .option("password", "sparkpassword")
 .save())
```

Spark Jobs to OpenLineage



Spark Datasets supported

- Hadoop FS
 - o HDFS
 - o GCS
 - o **S3**
 - o ABFS
- JDBC (postgres, redshift, etc.)
- Hive
- BigQuery
- Snowflake
- Iceberg
- Kafka

Extensible API

- Custom LogicalPlan visitors
 - Extract dataset information from custom datasources
 - Extract custom dataset facets from plan nodes
- Custom dataset and facet builders
 - Receive SparkListener events and return custom datasets and facets
 - Generate custom job facets or run facets from application environment
- Java ServiceLoader mechanics
- Contribute to open source or keep them in house

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Custom Datasets and Facets

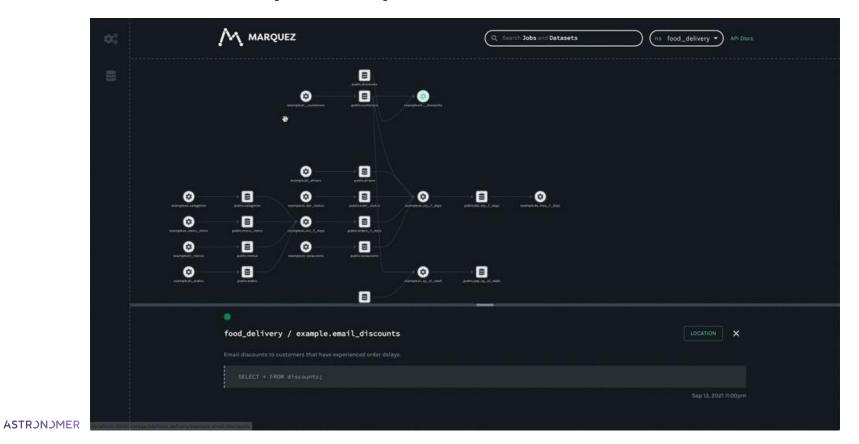
- QueryPlanVisitor
- AbstractInputDatasetBuilder
- AbstractOutputDatasetBuilder
- CustomFacetBuilder

```
1/**
* {@link CustomFacetBuilder} that generates a {@link EnvironmentFacet} when
 * Databricks.
*/
@Slf4i
public class DatabricksEnvironmentFacetBuilder
    extends CustomFacetBuilder<SparkListenerJobStart, EnvironmentFacet> {
 private HashMap<String, Object> dbProperties;
 private final OpenLineageContext openLineageContext;
 private Class dbutilsClass;
 private DbfsUtils dbutils;
  public DatabricksEnvironmentFacetBuilder(OpenLineageContext openLineage(
    this.openLineageContext = openLineageContext;
 public static boolean isDatabricksRuntime() {
   return System.getenv().containsKey("DATABRICKS_RUNTIME_VERSION");
 @Override
 protected void build(
      SparkListenerJobStart event, BiConsumer<String, ? super Environment
    consumer.accept(
        t: "environment-properties",
        new EnvironmentFacet(getDatabricksEnvironmentalAttributes(event))]
 private HashMan<String, Object> getDatabricksEnvironmentalAttributes(
```

Marquez: open source metadata



Marquez: open source metadata



OpenLineage Contributors













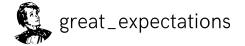












Resources

- https://openlineage.io/
- https://github.com/OpenLineage/OpenLineage
- https://openlineage.slack.com/
- https://twitter.com/OpenLineage
- https://datakin.com/blog/
- https://marquezproject.github.io/marquez/

- in michael-collado-80351753
- @PeladoCollado

Q & A

