



Building Modern Data Stacks

Project Brief

Adventure Works

This project demonstrate how to build data a modern data stack for an e-commerce, implement machine learning Models, and develop business intelligence reporting solutions.

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Building Modern Data Stacks

First Things First !!!

Four Key Questions

- I. Where do we consolidate our data ? > [Storage](#)
- II. How will we get it there ? > [Ingestion](#)
- III. How will we clean it up? > [Transformation](#)
- IV. How will we analyze it? > [Reporting](#)



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The
Big Choice

Data Stack

Popular Options

Storage > [Snowflake](#), [BigQuery](#), [s3](#), Redshift

Ingestion > [Airbyte](#), [Airflow](#), Fivetran, dagster

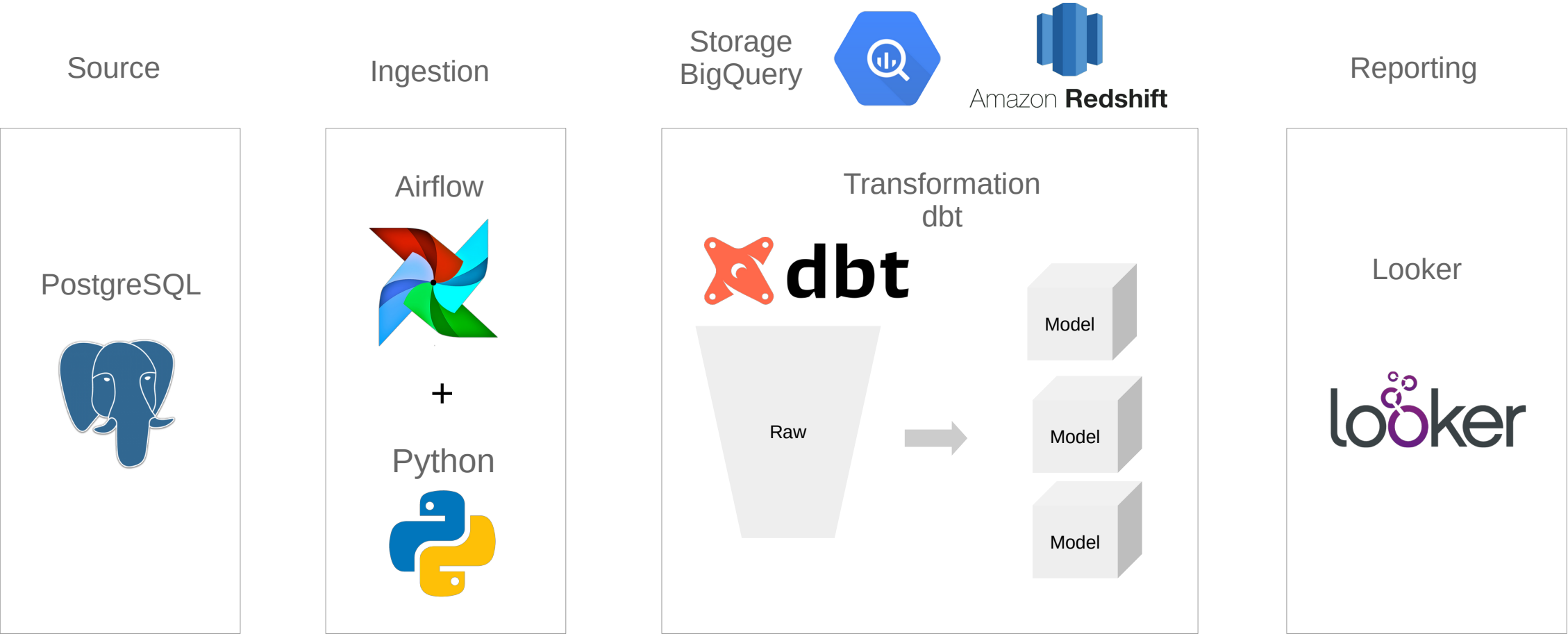
Transformation > [dbt](#)

Reporting > [Tableau](#), Power BI, [Looker](#), Superset

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Data Stack Architecture Design





End Goal

**Put data to use
- make decisions**



“Data is like garbage. You’d better know what
you are going to do with it before you collect it.”

~ Mark Twain

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PROJECTS

01

Storage/Database/Data Warehouse

Google [BigQuery](#)

[Snowflake](#)

[AWS Redshift](#)

02

Ingestion

[Apache Airflow](#)

[Airbyte](#)

[Dagster](#)

03

Transformation

Setting up [dbt](#)

Building Models

04

Reporting

[Looker](#)

[Tableau](#)

[Power BI](#)

02

INGESTION BUILDING ELT DATA PIPELINE - Apache Airflow

Setting up Apache Airflow - [Documentation](#)
[Python ELT \(Extract Load Transform\) script](#)

```
# importing libraries
```

```
from airflow.decorators import dag, task
from datetime import datetime, timedelta
import requests
from google.cloud import bigquery
import pandas as pd
import psycopg2
from io import StringIO
```

02

INGESTION

Setting up Apache Airflow

Defining a DAG - Directed Acyclic Graph

```
args{
  "owner": "gwayi",
  "retries": 1,
  "retry_delay": timedelta(minutes=5)
}

@dag(
  default_arguments = args
  schedule=timedelta(minutes=30),
  start_date=datetime(2024, 7, 29),
  catchup=False,
  tags=['Team B']
)
```


02

INGESTION

Setting up Apache Airflow

Extract Task Group – Source PostgreSQL Database

```
@task()
def extract():
    try:
        src_cursor.execute(sql)
        tables = cursor.fetchall()

        output = {}

        for table in tables:
            cursor.execute(f"SELECT *
                            FROM {table[0]}")
```

```
        rows = cursor.fetchall()

        output.update({table[0]: rows})
        return output

    except Exception as e:
        print("extract error:" +
              str(e))

    finally:
        connection.close()
    output = extract()
```

Incremental Load

Data Change Capture



Timestamp Approach – extracts all rows modified
Since the last execution date `{{ds}}`.

```
SELECT *  
FROM {table[0]}  
WHERE last_updated >= '{{ ds }}';
```

02

INGESTION

Setting up Apache Airflow

Load Task Group – Destination BigQuery

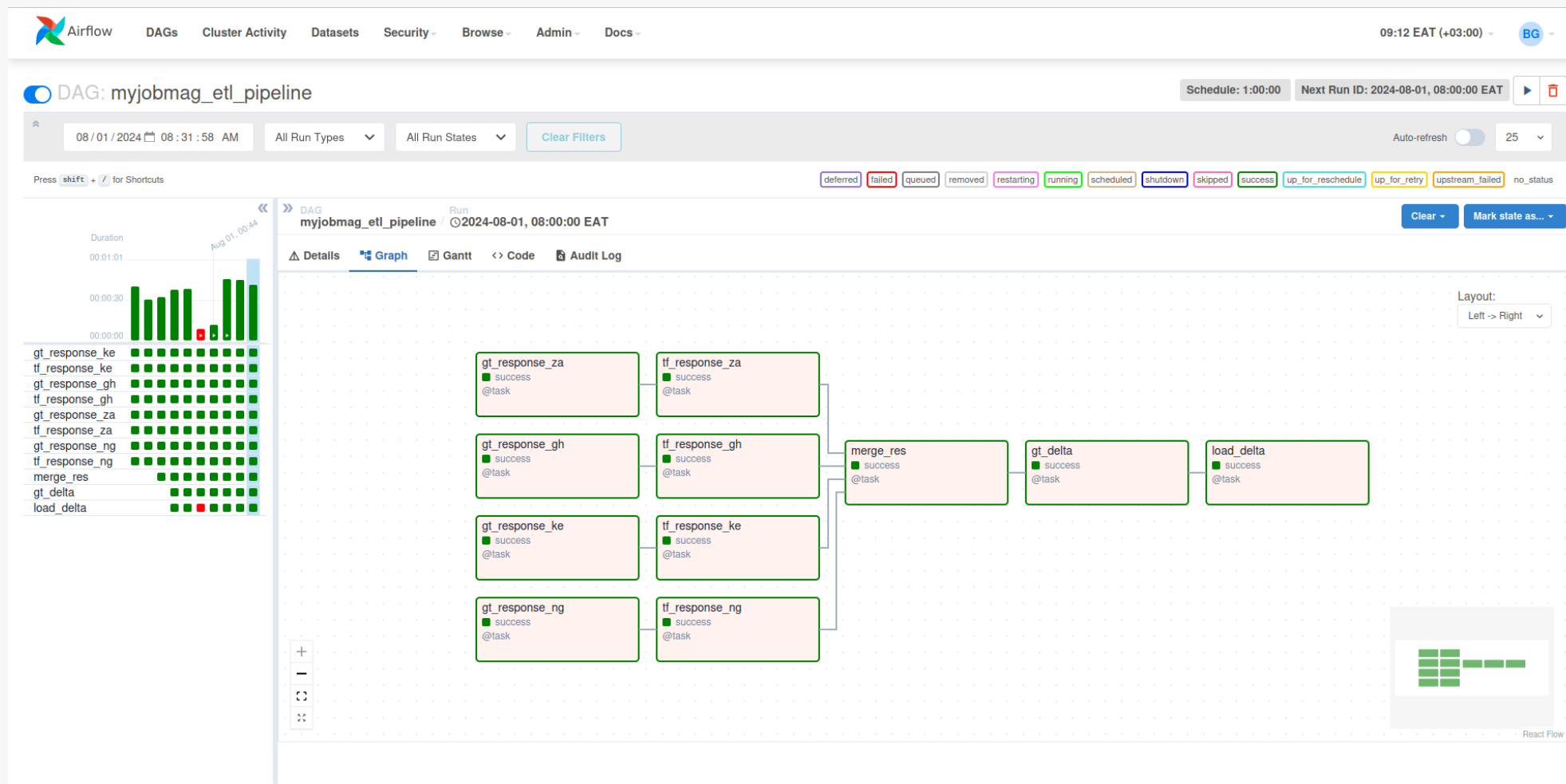
```
task()  
@def load(dict):  
    pandas_gbq.to_gbq(  
        df[{table}],  
        project_id=project_id,  
        if_exists=append,  
load(data)
```

Set dependencies

```
extract = extract()  
load = load(extract)
```

```
extract >> load
```

Orchestrating Data Pipeline - Airflow



Viewing resources.

[SHOW STARRED ONLY](#)

▼ **adventureworks-431609** ☆ ⋮

▶ 🔍 Queries ⋮

▶ 📓 Notebooks ⋮

▶ 🗂 Data canvases ⋮

▶ ≡ Data preparations ⋮

▶ 🔗 External connections ⋮

▼ 🗃 stg ☆ ⋮

🗃 customer ☆ ⋮

🗃 product ☆ ⋮

🗃 product_category ☆ ⋮

🗃 product_subcategory ☆ ⋮

🗃 returns ☆ ⋮

🗃 sales ☆ ⋮

🗃 territory ☆ ⋮

SCHEMA

DETAILS

PREVIEW

TABLE EXPLORER

PREVIEW

INSIGHTS

PREVIEW

LINEAGE

Row	customerid	firstname	lastname	fullname
1	1305	A.	Leonetti	A. Leonetti
2	1305	A.	Leonetti	A. Leonetti
3	829	Ed	Dudenhoefer	Ed Dudenhoefer
4	829	Ed	Dudenhoefer	Ed Dudenhoefer
5	1953	H.	Valentine	H. Valentine
6	1953	H.	Valentine	H. Valentine
7	539	Jo	Brown	Jo Brown
8	539	Jo	Brown	Jo Brown
9	1917	Abe	Tramel	Abe Tramel
10	1917	Abe	Tramel	Abe Tramel
11	323	Amy	Alberts	Amy Alberts
12	323	Amy	Alberts	Amy Alberts
13	735	Amy	Consentino	Amy Consentino
14	735	Amy	Consentino	Amy Consentino
15	1033	Ann	Hass	Ann Hass
16	1033	Ann	Hass	Ann Hass
17	437	Ann	Beebe	Ann Beebe



Transformation



Business Modeling

Using dbt build three models ;

ML Model – for Machine Learning Model

Production – for production

Reporting – for reporting



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Viewing resources.

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- ▼ adventureworks-431609 ☆ ⋮
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 - ▶ 📄 Notebooks ⋮
 - ▶ 🗂 Data canvases ⋮
 - ▶ ⚙ Data preparations ⋮
 - ▶ 🔌 External connections ⋮
 - ▶ 🗃 stg ☆ ⋮
 - ▶ 🗃 stg_ml ☆ ⋮
 - ▶ 🗃 stg_prod ☆ ⋮
 - ▶ 🗃 stg_reporting ☆ ⋮

SCHEMA		DETAILS	PREVIEW	TABLE EXPLORER	PREVIEW	INSIGHTS	PREVIEW	LINEAGE
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