



Building Modern Data Stacks

Project Brief

A d v e n t u r e W o r k s

Adventure works is a bicycle manufacturing company. This project demonstrated how to build data pipelines for an e-commerce, implement machine learning models, and develop business intelligence reporting solutions.



BRIAN GWAYI





Building Modern Data Stacks

Foundation 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](#)



BRIAN GWAYI





The
Big Choice

Data Stack



Popular Options

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

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

Transformation > [dbt](#)

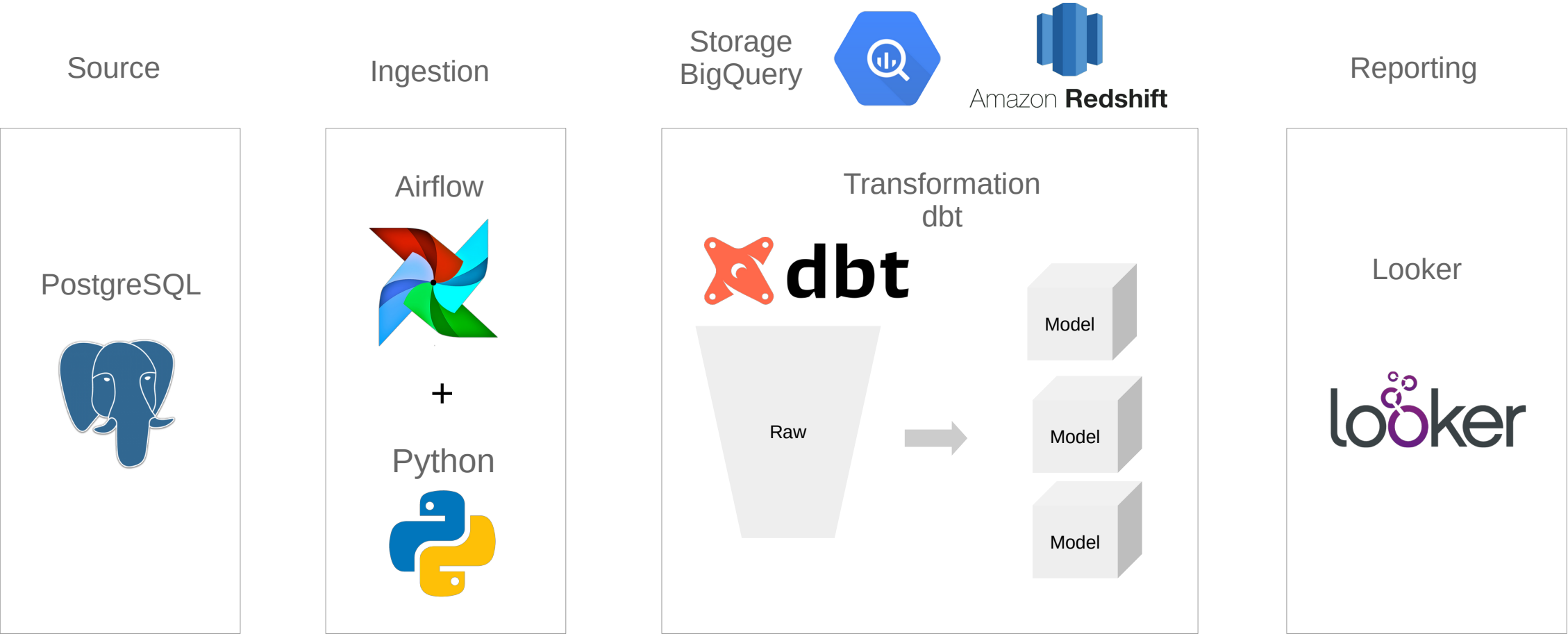
Reporting > [Tableau](#), Power BI, [Looker](#), Superset

N/B This is not an exhaustive list.

BRIAN GWAYI



Data Stack Architecture Design





End Goal

Put data to use



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

~ Mark Twain

BRIAN GWAYI



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

Setting up Apache Airflow - [Documentation](#)

[Phase I: Development](#)

[-Writing python scripts](#)

```
# 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()
```

02

INGESTION

Setting up Apache Airflow

Load Task Group – Destination BigQuery

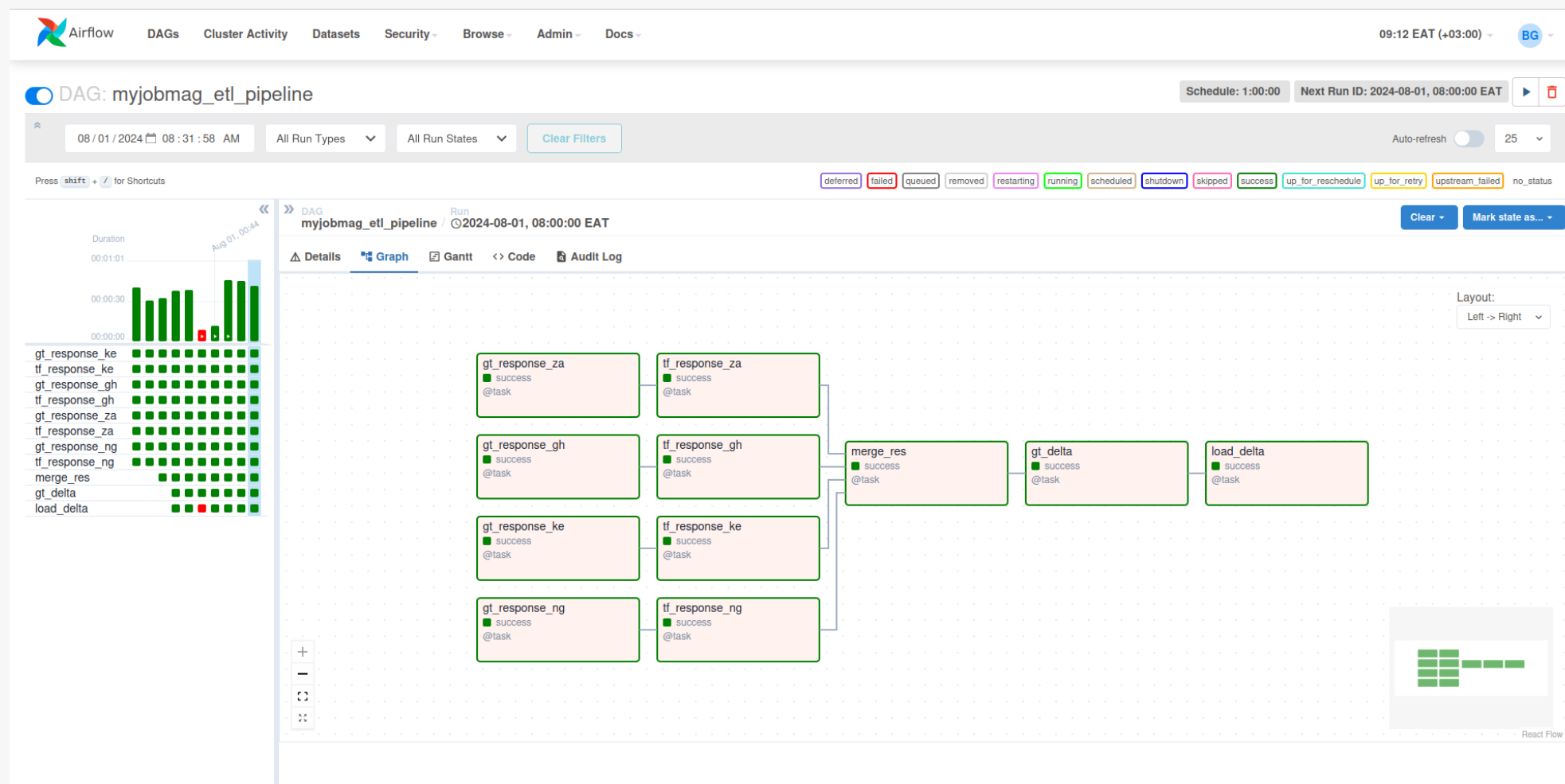
```
task()  
@def load(dict):  
    pandas_gbq.to_gbq(  
        df,  
  
        project_id=project_id,  
        if_exists=append,  
load(data)
```

Set dependencies

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

```
extract >> load
```

Orchestrating Data Pipeline - Airflow



Google BigQuery

Viewing resources.

[SHOW STARRED ONLY](#)

- ▼ adventureworks-431609
 - ▶ Queries
 - ▶ Notebooks
 - ▶ Data canvases
 - ▶ External connections
 - ▼ adw_dwh
 - customer
 - customers
 - employees
 - my_first_dbt_model
 - my_second_dbt_model
 - orders
 - product
 - productcategory
 - ▶ prod

SUMMARY

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		