Building A Simple P Data Thi a m Stack

A Project Brief

This project demonstrate how to implement a modern data stack, build data pipelines, machine learning and reporting capabilities using a variety of solutions.

> BRIAN GWAYI Independent Data Lead & Engineer



Five Key Questions

I. Where is our data? Source

II. Where do we consolidate our data? Storage

III. How will we get it there? <u>Ingestion</u>

IV. How will we clean it up? <u>Transformation</u>

V. How will we analyze it? Reporting

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Data Stack Architecture Design

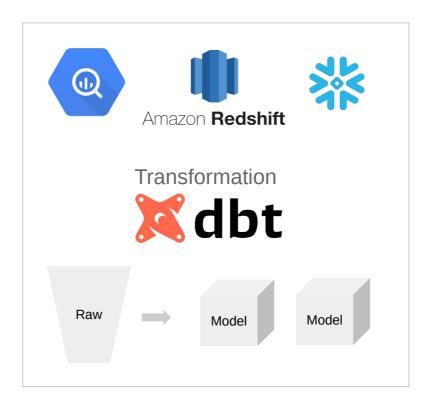
Where is our data?



How will we get it there?



Where do we consolidate our data?



How do we analyze it?



Ultimate End Data

Goal

Data + Insights + Action = Actionable Insights

What happened/ will happen?

Insight Action

Why did it happened/ will thappened/ will it happen?

Where

is Our Data?

Source : PostgresSQL

Schema : Public

Database Name: adw_db

Tables Count : 7
Tables : [customer,

product,

product_category,

return, sales,

territory,

product_subcategory]

>	Aa FTS Parsers					
>	FTS Templates					
>	📋 Foreign Tables					
>	(Functions					
>	Materialized Views					
>	> 🖺 Operators					
> (() Procedures						
> 1.3 Sequences						
→ Image: Tables (7)						
	> 🗎 customer					
	> == product					
	> 🖽 product_category					
	> 🖽 product_subcategory					
	> == returns					
	> 🖽 sales					
	> 🖽 territory					

> (a) Trigger Functions

√ □ Types

→ □ Views

Data Output Messages Notifications									
=+ 1	<u> </u>		♣ ✓ SQL						
	orderdate date	stockdate date	ordernumber character varying (255)	productkey integer	customerkey integer				
1	2022-01-01	2021-12-13	S061285	529	23791				
2	2022-01-01	2021-09-24	S061285	214	23791				
3	2022-01-01	2021-09-04	S061285	540	23791				
4	2022-01-01	2021-09-28	S061301	529	16747				
5	2022-01-01	2021-10-21	S061301	377	16747				
6	2022-01-01	2021-10-23	S061301	540	16747				
7	2022-01-01	2021-09-04	S061269	215	11792				
8	2022-01-01	2021-10-21	S061269	229	11792				
9	2022-01-01	2021-10-24	S061286	528	11530				
10	2022-01-01	2021-09-27	S061286	536	11530				
11	2022-01-01	2021-10-23	S061298	530	18155				
12	2022-01-01	2021-12-02	S061298	214	18155				
13	2022-01-01	2021-12-15	SO61298	223	18155				
14	2022-01-01	2021-10-01	S061310	538	13541				
15	2022-01-01	2021-11-08	S061310	584	13541				

HOW do we ingest Our Data?

Ingestion : python script
Orchestration : Apache Airflow

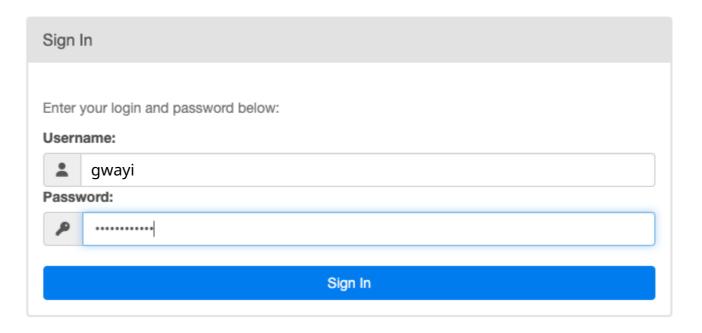
Apache Airflow Setup

Terminal

- \$ python3 -m venv airflow-env
- \$ source airflow-env/bin/activate
- \$ export AIRFLOW_HOME=~/airflow
- \$ pip install apache-airflow
- \$ airflow db init
- \$ airflow webserver -p 8080
- \$ airflow sheduler

Apache Airflow Webserver UI





HOW do we ingest Our Data?

```
Ingestion : python script
Orchestration : Apache Airflow
```

ELT Python Script

```
pip install google-cloud-bigquery
pip install --upgrade snowflake-connector-python
# importing libraries
from airflow.decorators import dag, task
from datetime import datetime, timedelta
from google.cloud import bigquery
import pandas as pd
import psycopg2
```

Insatiate a DAG

```
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']
    )
```

HOW do we ingest Our Data?

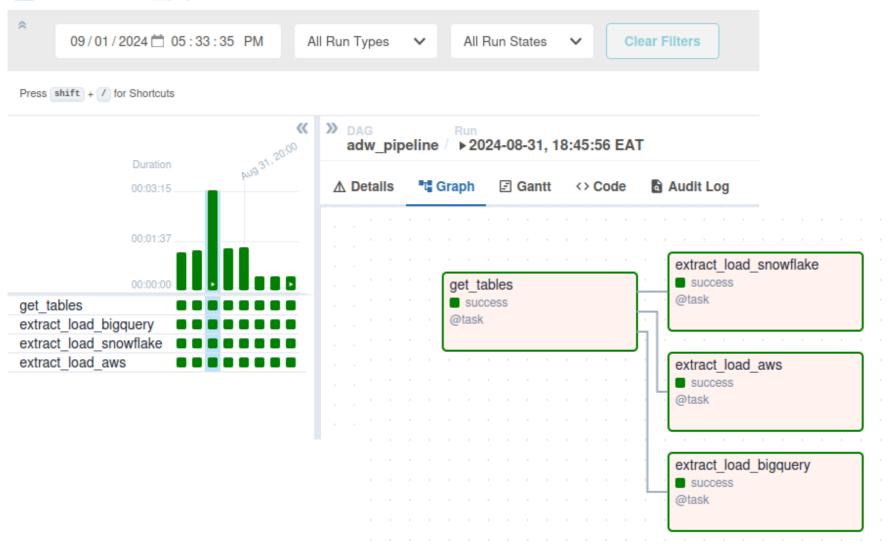
Ingestion: Python Script

```
Orchestration: Apache Airflow
Task I (Get Tables)
@task()
def get tables():
"""extract list of tables in public schema"""
 try:
    cursor.execute(
        f"""SELECT table name
         FROM information schema.tables
        WHERE table schema = 'public'""
 tbls = [x[0]] for x in cursor.fetchall()]
```

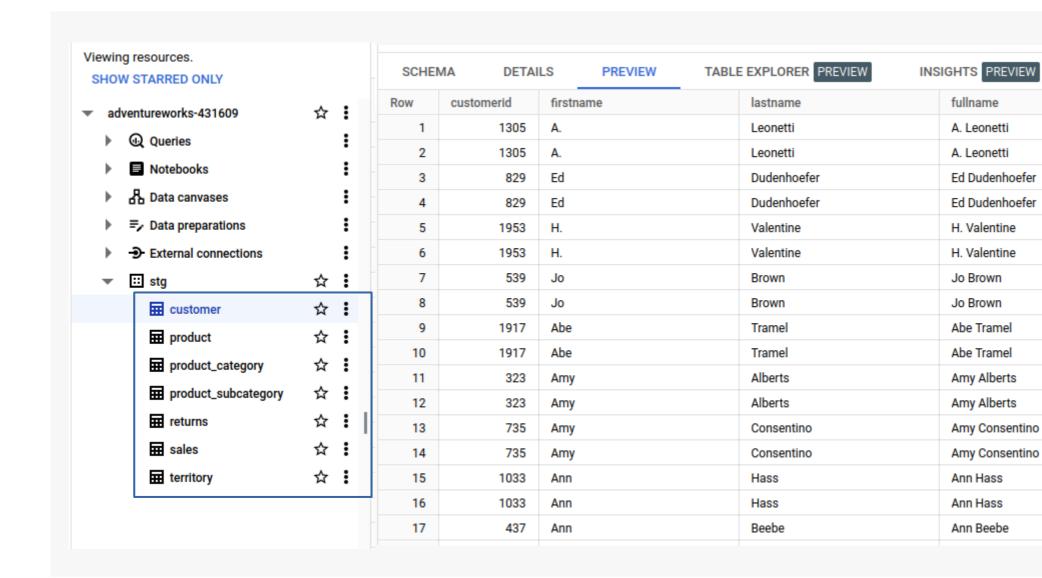
Task II (Extract_Load)

```
@task()
      def extract|load_bigguery(tbls, conn):
"""loop through tbls then extract & load"""
     client = bigguery.Client()
     job_config = bigquery.LoadJobConfig(
     write disposition="WRITE TRUNCATE")
      for tbl in tbls:
      table id = f"adventureworks-431609.stg.{tbl}"
      sql = f"SELECT * FROM {tbl} WHERE
      updated_at >= {ds}'"
      df = pd.read sql(sql, conn)
      job = client.load table from dataframe(
      df, table_id, job_config=job_config)
      iob.result()
    get_tables = get_tables()
    extract load = extract load(get tables)
```

DAG: adw_pipeline



Tables Loaded: BigQuery



LINEAGE

How do we

transform Our Data?

Insatiate a DAG

args

Transformation : dbt

Orchestration : Apache Airflow

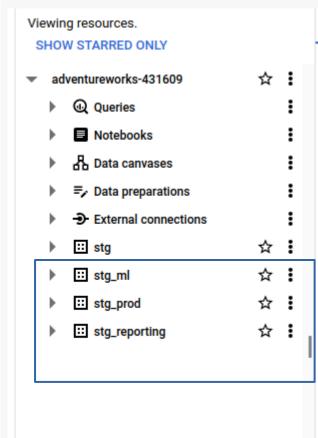
Models 3 : [production,

machine learning,

Reporting]

dbt Setup

```
pip install google-cloud-bigquery
pip install --upgrade snowflake-connector-python
```



SCHE	MA DETAI	LS 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	

PROJECTS

Storage/Data Warehouse 01 Implementing Data Warehouse Solutions Google BigQuery | Snowflake | AWS Redshift | Oracle ADW Ingestion Developing Data Pipelines Python | Airflow | Airbyte | dagster | Prefect 03 **Transformation** Setting up dbt **Building Models** Reporting Looker | Tableau | Power BI **Machine Learning** 05 Building ML Models