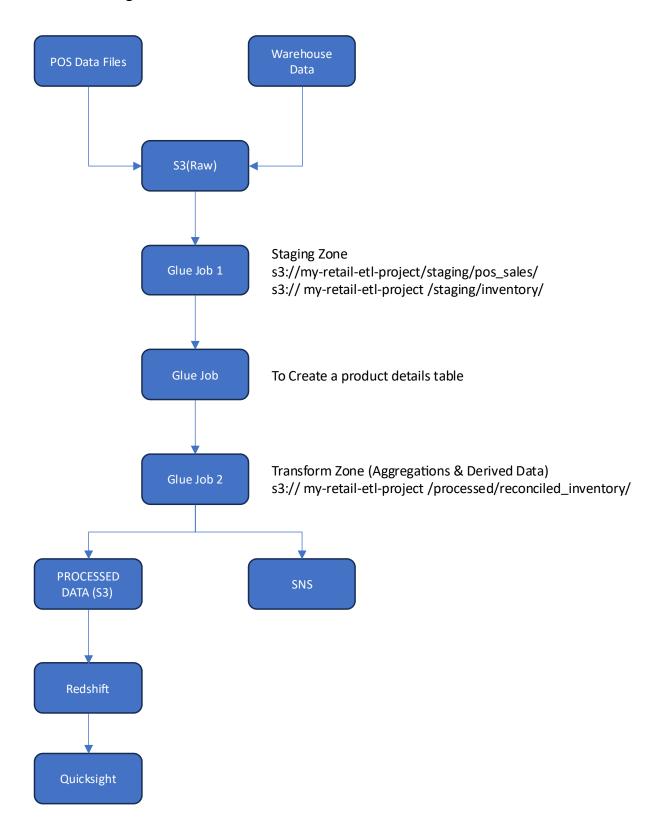
Retail Inventory Reconciliation (Batch ETL)

Summary:

- Purpose: This document outlines the architecture and data flow for the automated sales
 data pipeline. The primary goal of this pipeline is to ingest Point-of-Sale (POS) and
 supplementary warehouse data, process it, and load it into a centralized data warehouse
 to enable business intelligence (BI) and reporting.
- Business Goal: To provide the sales and marketing teams with daily updated dashboards in Amazon QuickSight for analyzing sales performance, product trends, and store efficiency.
- Technologies Used: AWS S3, AWS Glue, AWS SNS, Amazon Redshift, Amazon QuickSight, Airflow.

Architecture Diagram:

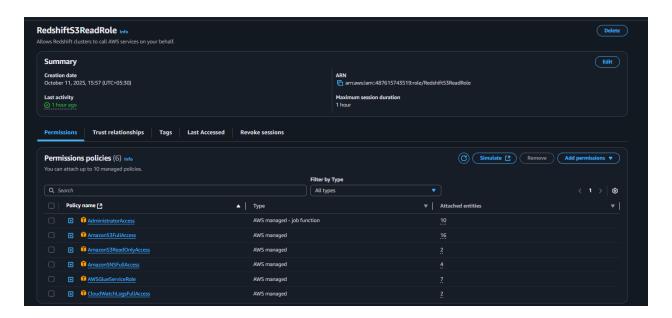


Detailed Implementation Steps & Configuration:

Step 1: Prerequisite: IAM Role Configuration

Before creating the pipeline, we need IAM roles with the correct permissions to allow AWS services to interact with each other.

IAM Role:



Step 2: Amazon S3 Bucket Setup

First, we'll create the main container for your project files. Remember, S3 bucket names must be **globally unique**.

- Navigate to S3: In the AWS Management Console, search for and select "S3".
- 2. Start Bucket Creation: Click the Create bucket button.
- 3. Configure Bucket Name and Region:
 - Bucket name: Choose a unique name. A good practice is [project-name]-[purpose]-[date].
 - Example: retail-etl-project-datalake-20251013
 - AWS Region: Select the region closest to you, for example, Asia Pacific (Mumbai)
 ap-south-1.
- 4. Object Ownership: Leave this as the recommended default, ACLs disabled.

5. Block Public Access (Security):

 Keep the box for Block all public access checked. This is the most secure setting and is correct for a data pipeline where access is programmatic, not public.

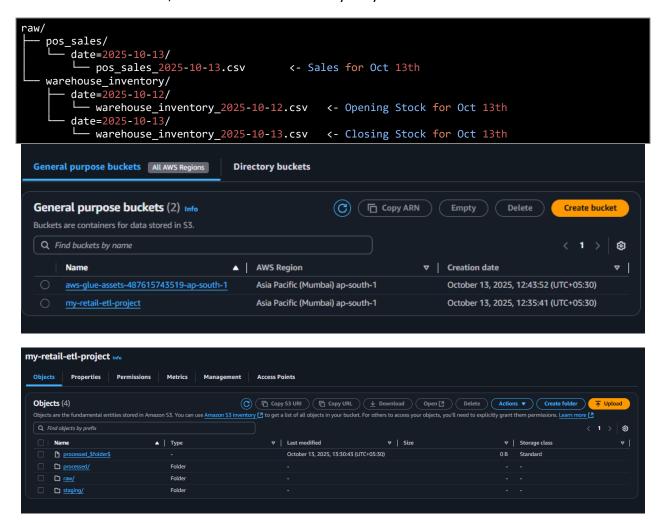
6. **Bucket Versioning (Data Protection):**

 Select Enable. This is a best practice that keeps a history of all your file versions, protecting you from accidental overwrites or deletions.

7. Create the Bucket:

Scroll to the bottom and click Create bucket.

You now have a secure, versioned S3 bucket ready for your data.





pos sales 2025-10-13:

```
transaction_id,store_id,sku,product_name,sale_date,quantity,price
T051,S001,SKU001,Wireless Mouse,2025-10-13,3,5
T052,S002,SKU008,Headphones,2025-10-13,5,1299
T053,S003,SKU016,Monitor 27Inch,2025-10-13,1,10999
T054,S004,SKU022,Microphone,2025-10-13,2,2999
T055,S001,SKU030,VR Headset,2025-10-13,1,9999
T056,S002,SKU045,Headphones,2025-10-13,4,1299
T057,S003,SKU012,Pen Drive 32GB,2025-10-13,8,499
T058,S004,SKU004,HDMI Cable,2025-10-13,12,299
T059,S001,SKU009,Mouse Pad,2025-10-13,10,199
T060,S002,SKU019,Portable SSD,2025-10-13,2,7999
T061,S003,SKU028,Stylus Pen,2025-10-13,5,599
T062,S004,SKU036,Wifi Adapter,2025-10-13,3,799
T063,S001,SKU050,Router,2025-10-13,1,2399
T064,S002,SKU041,HDMI Cable,2025-10-13,7,299
T065,S003,SKU033,Screen Cleaner Kit,2025-10-13,4,299
T066,S004,SKU025,Power Bank,2025-10-13,3,1299
T067,S001,SKU017,Smart Watch,2025-10-13,1,4999
T068,S002,SKU002,Keyboard,2025-10-13,4,899
T069,S003,SKU011,External HDD 1TB,2025-10-13,2,4599
T070,S004,SKU020,Wireless Charger,2025-10-13,6,999
T071,S001,SKU031,Smartphone Stand,2025-10-13,5,499
T072,S002,SKU048,External HDD 1TB,2025-10-13,1,4599
T073,S003,SKU039,Keyboard,2025-10-13,3,899
T074,S004,SKU049,Pen Drive 32GB,2025-10-13,10,499
T075,S001,SKU005,USB Hub,2025-10-13,2,499
T076,S002,SKU015,Laptop Bag,2025-10-13,2,1499
T077,S003,SKU024,Phone Case,2025-10-13,15,299
T078,S004,SKU032,Wireless Earbuds,2025-10-13,3,3499
T079,S001,SKU042,USB Hub,2025-10-13,4,499
T080,S002,SKU007,WebCam,2025-10-13,1,1999
T081,S003,SKU014,Printer Ink,2025-10-13,2,699
T082,S004,SKU021,HD Webcam,2025-10-13,2,2499
T083,S001,SKU035,Laptop Cooling Pad,2025-10-13,1,1499
T084,S002,SKU044,Webcam,2025-10-13,3,1999
T085,S003,SKU003,Monitor 24Inch,2025-10-13,1,7999
T086,S004,SKU013,Router,2025-10-13,1,2399
T087,S001,SKU023,Speaker Bluetooth,2025-10-13,2,1799
T088,S002,SKU034,Phone Tripod,2025-10-13,2,799
T089,S003,SKU043,Laptop Stand,2025-10-13,4,999
T090,S004,SKU047,Wireless Keyboard,2025-10-13,2,1299
T091,S001,SKU010,Wireless Keyboard,2025-10-13,3,1299
T092,S002,SKU026,Extension Board,2025-10-13,5,899
T093,S003,SKU037,Graphic Tablet,2025-10-13,1,8999
T094,S004,SKU046,Mouse Pad,2025-10-13,8,199
T095,S001,SKU006,Laptop Stand,2025-10-13,3,999
T096,S002,SKU018,Tablet 10Inch,2025-10-13,1,14999
T097,S003,SKU027,Monitor Stand,2025-10-13,1,1199
T098,S004,SKU029,HDMI Splitter,2025-10-13,2,699
T099,S001,SKU038,Wireless Mouse,2025-10-13,3,599
T100,S002,SKU040,Monitor 24Inch,2025-10-13,1,7999
```

warehouse inventory 2025-10-12:

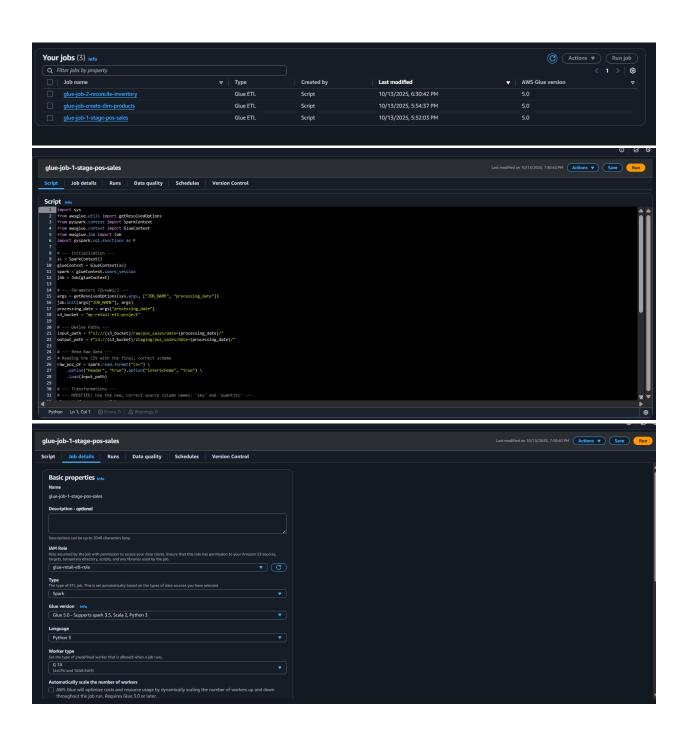
```
transaction_id,store_id,sku,product_name,sale_date,quantity,price
T051,S001,SKU001,Wireless Mouse,2025-10-13,3,599
T052,S002,SKU008,Headphones,2025-10-13,5,1299
T053,S003,SKU016,Monitor 27Inch,2025-10-13,1,10999
T054,S004,SKU022,Microphone,2025-10-13,2,2999
T055,S001,SKU030,VR Headset,2025-10-13,1,9999
T056,S002,SKU045,Headphones,2025-10-13,4,1299
T057,S003,SKU012,Pen Drive 32GB,2025-10-13,8,499
T058,S004,SKU004,HDMI Cable,2025-10-13,12,299
T059,S001,SKU009,Mouse Pad,2025-10-13,10,199
T060,S002,SKU019,Portable SSD,2025-10-13,2,7999
T061,S003,SKU028,Stylus Pen,2025-10-13,5,599
T062,S004,SKU036,Wifi Adapter,2025-10-13,3,799
T063,S001,SKU050,Router,2025-10-13,1,2399
T064,S002,SKU041,HDMI Cable,2025-10-13,7,299
T065,S003,SKU033,Screen Cleaner Kit,2025-10-13,4,299
T066,S004,SKU025,Power Bank,2025-10-13,3,1299
T067,S001,SKU017,Smart Watch,2025-10-13,1,4999
T068,S002,SKU002,Keyboard,2025-10-13,4,899
T069,S003,SKU011,External HDD 1TB,2025-10-13,2,4599
T070,S004,SKU020,Wireless Charger,2025-10-13,6,999
T071,S001,SKU031,Smartphone Stand,2025-10-13,5,499
T072,S002,SKU048,External HDD 1TB,2025-10-13,1,4599
T073,S003,SKU039,Keyboard,2025-10-13,3,899
T074,S004,SKU049,Pen Drive 32GB,2025-10-13,10,499
T075,S001,SKU005,USB Hub,2025-10-13,2,499
T076,S002,SKU015,Laptop Bag,2025-10-13,2,1499
T077,S003,SKU024,Phone Case,2025-10-13,15,299
T078,S004,SKU032,Wireless Earbuds,2025-10-13,3,3499
T079,S001,SKU042,USB Hub,2025-10-13,4,499
T080,S002,SKU007,WebCam,2025-10-13,1,1999
T081,S003,SKU014,Printer Ink,2025-10-13,2,699
T082,S004,SKU021,HD Webcam,2025-10-13,2,2499
T083,S001,SKU035,Laptop Cooling Pad,2025-10-13,1,1499
T084,S002,SKU044,Webcam,2025-10-13,3,1999
T085,S003,SKU003,Monitor 24Inch,2025-10-13,1,7999
T086,S004,SKU013,Router,2025-10-13,1,2399
T087,S001,SKU023,Speaker Bluetooth,2025-10-13,2,1799
T088,S002,SKU034,Phone Tripod,2025-10-13,2,799
T089,S003,SKU043,Laptop Stand,2025-10-13,4,999
T090,S004,SKU047,Wireless Keyboard,2025-10-13,2,1299
T091,S001,SKU010,Wireless Keyboard,2025-10-13,3,1299
T092,S002,SKU026,Extension Board,2025-10-13,5,899
T093,S003,SKU037,Graphic Tablet,2025-10-13,1,8999
T094,S004,SKU046,Mouse Pad,2025-10-13,8,199
T095,S001,SKU006,Laptop Stand,2025-10-13,3,999
T096,S002,SKU018,Tablet 10Inch,2025-10-13,1,14999
T097,S003,SKU027,Monitor Stand,2025-10-13,1,1199
T098,S004,SKU029,HDMI Splitter,2025-10-13,2,699
T099,S001,SKU038,Wireless Mouse,2025-10-13,3,599
T100,S002,SKU040,Monitor 24Inch,2025-10-13,1,7999
```

warehouse inventory 2025-10-13:

```
record_id,sku,product_name,category,warehouse_location,stock_on_hand,received_date
R051, SKU001, Wireless Mouse, Accessories, WH1, 111, 2025-10-13
R052, SKU002, Keyboard, Accessories, WH1, 73, 2025-10-13
R053, SKU003, Monitor 24Inch, Displays, WH2, 42, 2025-10-13
R054,SKU004,HDMI Cable,Cables,WH3,278,2025-10-13
R055,SKU005,USB Hub,Accessories,WH2,134,2025-10-13
R056,SKU006,Laptop Stand,Accessories,WH1,101,2025-10-13
R057, SKU007, WebCam, Electronics, WH4, 47, 2025-10-13
R058, SKU008, Headphones, Audio, WH2, 185, 2025-10-13
R059, SKU009, Mouse Pad, Accessories, WH3, 158, 2025-10-13
R060, SKU010, Wireless Keyboard, Accessories, WH1, 133, 2025-10-13
R061, SKU011, External HDD 1TB, Storage, WH2, 65, 2025-10-13
R062, SKU012, Pen Drive 32GB, Storage, WH3, 232, 2025-10-13
R063, SKU013, Router, Networking, WH4, 57, 2025-10-13
R064, SKU014, Printer Ink, Printers, WH1, 33, 2025-10-13
R065,SKU015,Laptop Bag,Accessories,WH2,94,2025-10-13
R066, SKU016, Monitor 27Inch, Displays, WH3, 27, 2025-10-13
R067,SKU017,Smart Watch,Wearables,WH4,22,2025-10-13
R068,SKU018,Tablet 10Inch,Tablets,WH1,18,2025-10-13
R069,SKU019,Portable SSD,Storage,WH2,35,2025-10-13
R070, SKU020, Wireless Charger, Accessories, WH3, 66, 2025-10-13
R071,SKU021,HD Webcam,Electronics,WH1,40,2025-10-13
R072, SKU022, Microphone, Audio, WH4, 31, 2025-10-13
R073, SKU023, Speaker Bluetooth, Audio, WH2, 104, 2025-10-13
R074, SKU024, Phone Case, Accessories, WH3, 127, 2025-10-13
R075, SKU025, Power Bank, Electronics, WH1, 71, 2025-10-13
R076, SKU026, Extension Board, Accessories, WH2, 60, 2025-10-13
R077, SKU027, Monitor Stand, Displays, WH4, 22, 2025-10-13
R078, SKU028, Stylus Pen, Accessories, WH1, 82, 2025-10-13
R079,SKU029,HDMI Splitter,Cables,WH3,37,2025-10-13
R080, SKU030, VR Headset, Electronics, WH2, 13, 2025-10-13
R081,SKU031,Smartphone Stand,Accessories,WH1,118,2025-10-13
R082, SKU032, Wireless Earbuds, Audio, WH2, 52, 2025-10-13
R083, SKU033, Screen Cleaner Kit, Accessories, WH3, 190, 2025-10-13
R084, SKU034, Phone Tripod, Accessories, WH4, 65, 2025-10-13
R085, SKU035, Laptop Cooling Pad, Accessories, WH1, 57, 2025-10-13
R086, SKU036, Wifi Adapter, Networking, WH2, 39, 2025-10-13
R087, SKU037, Graphic Tablet, Electronics, WH3, 18, 2025-10-13
R088,SKU038,Wireless Mouse,Accessories,WH4,93,2025-10-13
R089, SKU039, Keyboard, Accessories, WH1, 77, 2025-10-13
R090, SKU040, Monitor 24Inch, Displays, WH2, 48, 2025-10-13
R091,SKU041,HDMI Cable,Cables,WH3,294,2025-10-13
R092,SKU042,USB Hub,Accessories,WH4,71,2025-10-13
R093, SKU043, Laptop Stand, Accessories, WH1, 89, 2025-10-13
R094, SKU044, Webcam, Electronics, WH2, 50, 2025-10-13
R095, SKU045, Headphones, Audio, WH3, 162, 2025-10-13
R096, SKU046, Mouse Pad, Accessories, WH4, 147, 2025-10-13
R097, SKU047, Wireless Keyboard, Accessories, WH1, 69, 2025-10-13
R098, SKU048, External HDD 1TB, Storage, WH2, 57, 2025-10-13
R099, SKU049, Pen Drive 32GB, Storage, WH3, 212, 2025-10-13
R100,SKU050,Router,Networking,WH4,52,2025-10-13
```

Step 3: Glue Job 1

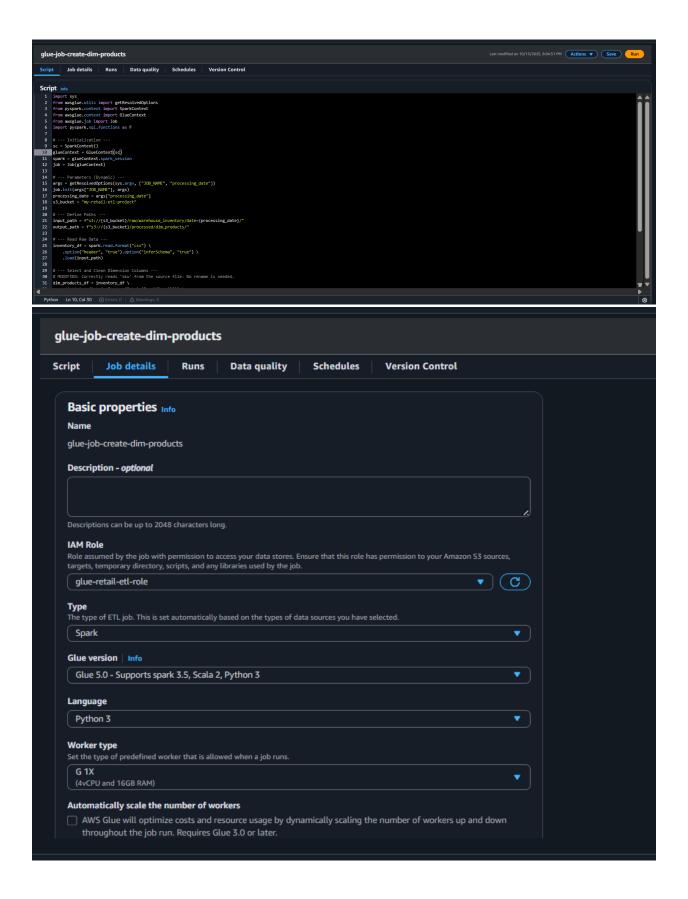
```
from awsglue.utils import getResolvedOptions
from pyspark.context import SparkContext
from awsglue.context import GlueContext
from awsglue.job import Job
import pyspark.sql.functions as F
# --- Initialization ---
sc = SparkContext()
glueContext = GlueContext(sc)
spark = glueContext.spark_session
job = Job(glueContext)
# --- Parameters (Dynamic) ---
args = getResolvedOptions(sys.argv, ["JOB_NAME", "processing_date"])
job.init(args["JOB_NAME"], args)
processing_date = args["processing_date"]
s3_bucket = "my-retail-etl-project
# --- Define Paths ---
input path = f"s3://{s3 bucket}/raw/pos sales/date={processing date}/"
output_path = f"s3://{s3_bucket}/staging/pos_sales/date={processing_date}/"
# --- Read Raw Data ---
# Reading the CSV with the final, correct schema
raw_pos_df = spark.read.format("csv") \
    .option("header", "true").option("inferSchema", "true") \
    .load(input_path)
# --- Transformations ---
# --- MODIFIED: Use the new, correct source column names: 'sku' and 'quantity' ---
cleaned_df = raw_pos_df \
    .withColumn("sku", F.upper(F.trim(F.col("sku")))) \
    .withColumn("quantity", F.col("quantity").cast("int"))
# --- MODIFIED: Group by 'sku'. We are no longer renaming this column. ---
# We also drop product_name from the grouping as it is descriptive data.
aggregated_sales_df = cleaned_df.groupBy("sku") \
     .agg(F.sum("quantity").alias("total_quantity_sold")) \
     .withColumn("date", F.lit(processing_date))
# --- Final Selection ---
# The 'sku' column name already matches our Redshift table, so no rename is needed. final_df = aggregated_sales_df.select(
    F.col("date").cast("date").alias("date_key"),
F.col("sku"),
    F.col("total_quantity_sold")
# --- Write to Staging Zone ---
final df.write \
    .mode("overwrite") \
    .parquet(output_path)
job.commit()
```



Step 4: Glue Job For dim-products table

glue-job-create-dim-products

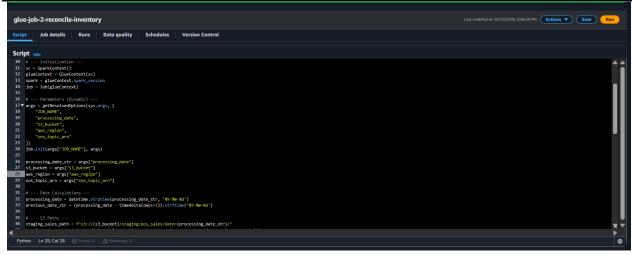
```
import sys
from awsglue.utils import getResolvedOptions
from pyspark.context import SparkContext
from awsglue.context import GlueContext
from awsglue.job import Job
import pyspark.sql.functions as F
# --- Initialization ---
sc = SparkContext()
glueContext = GlueContext(sc)
spark = glueContext.spark session
job = Job(glueContext)
# --- Parameters (Dynamic) ---
args = getResolvedOptions(sys.argv, ["JOB_NAME", "processing_date"])
job.init(args["JOB_NAME"], args)
processing_date = args["processing_date"]
s3 bucket = "my-retail-etl-project"
# --- Define Paths ---
input_path = f"s3://{s3_bucket}/raw/warehouse_inventory/date={processing_date}/"
output_path = f"s3://{s3_bucket}/processed/dim_products/'
# --- Read Raw Data ---
inventory_df = spark.read.format("csv") \
     .option("header", "true").option("inferSchema", "true") \
     .load(input_path)
# --- Select and Clean Dimension Columns ---
# MODIFIED: Correctly reads 'sku' from the source file. No rename is needed.
dim_products_df = inventory_df \
    .withColumn("sku", F.upper(F.trim(F.col("sku")))) \
    .withColumn("product_name", F.trim(F.col("product_name"))) \
.withColumn("category", F.trim(F.col("category"))) \
.select("sku", "product_name", "category") \
    .dropDuplicates(["sku"])
# --- Write to Processed Zone ---
dim_products_df.coalesce(1).write \
     .mode("overwrite") \
    .parquet(output_path)
job.commit()
```



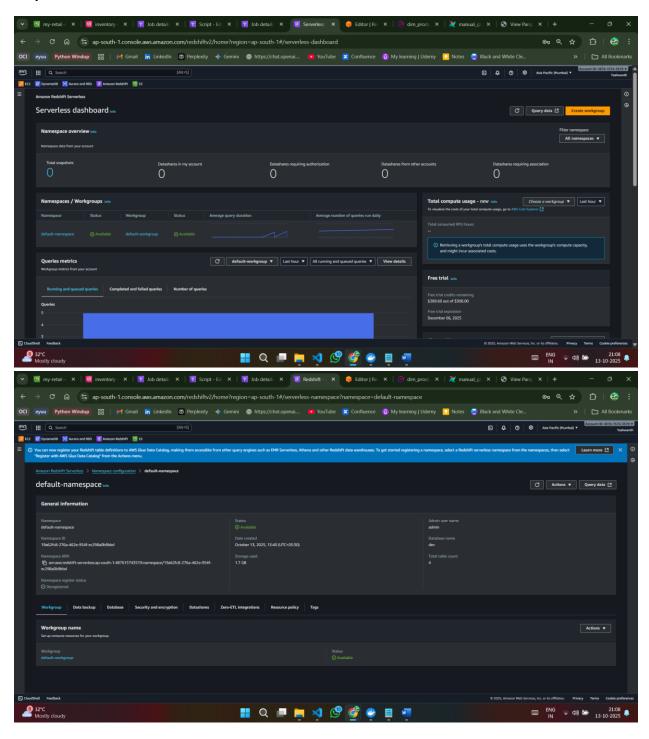
Step 5: Glue Job 2

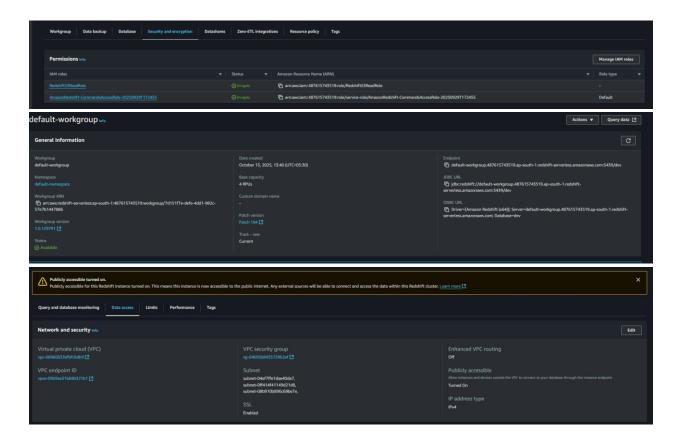
```
from awsglue.utils import getResolvedOptions
from pyspark.context import SparkContext
from awsglue.context import GlueContext
from awsglue.job import Job
import pyspark.sql.functions as F
from datetime import datetime, timedelta
import boto3
# --- Initialization ---
sc = SparkContext()
glueContext = GlueContext(sc)
spark = glueContext.spark session
job = Job(glueContext)
# --- Parameters (Dynamic) ---
args = getResolvedOptions(sys.argv, [
    "JOB_NAME",
"processing_date",
    "s3_bucket",
"aws_region",
    "sns_topic_arn"
job.init(args["JOB_NAME"], args)
processing_date_str = args["processing_date"]
s3_bucket = args["s3_bucket"]
aws_region = args["aws_region"]
sns_topic_arn = args["sns_topic_arn"]
# --- Date Calculations ---
processing_date = datetime.strptime(processing_date_str, '%Y-%m-%d')
previous_date_str = (processing_date - timedelta(days=1)).strftime('%Y-%m-%d')
staging_sales_path = f"s3://{s3_bucket}/staging/pos_sales/date={processing_date_str}/"
raw_inventory_path_today = f"s3://{s3_bucket}/raw/warehouse_inventory/date={processing_date_str}/"
raw_inventory_path_yesterday = f"s3://{s3_bucket}/raw/warehouse_inventory/date={previous_date_str}/'
dim products path = f"s3://{s3 bucket}/processed/dim products/'
processed_output_path =
f"s3://{s3_bucket}/processed/reconciled_inventory/date={processing date str}/"
# --- 1. Load DataFrames ---
daily sales_df = spark.read.parquet(staging_sales_path)
opening_stock_df = spark.read.format("csv").option("header", "true").option("inferSchema",
"true").load(raw_inventory_path_yesterday).select(F.upper(F.trim(F.col("sku"))).alias("sku"),F.col("
stock on hand").alias("opening stock"))
actual closing stock df = spark.read.format("csv").option("header", "true").option("inferSchema",
"true").load(raw_inventory_path_today).select(F.upper(F.trim(F.col("sku"))).alias("sku"),F.col("stoc
k_on_hand").alias("actual_closing_stock"))
dim products df = spark.read.parquet(dim products path)
# --- 2. Join and Reconcile ---
inventory_df = opening_stock_df.join(actual_closing_stock_df, "sku", "full_outer")
reconciliation_df = inventory_df.join(daily_sales_df, "sku", "left")
reconciliation with names df = reconciliation df.join(dim products df.select("sku", "product name"),
"sku", "left")
# --- 3. Calculate and Finalize ---
final_df = reconciliation_with_names_df.fillna(0, subset=['opening_stock', 'actual_closing_stock',
'total quantity sold']).withColumn("expected closing stock", (F.col("opening stock'
```

```
F.col("total_quantity_sold"))).withColumn("discrepancy", (F.col("actual_closing_stock") -
F.col("expected_closing_stock"))).withColumn("date", F.lit(processing_date_str))
# --- 4. Select and Rename Final Columns ---
final_df_selected =
final_df.select(F.col("date").cast("date").alias("date_key"),F.col("sku"),F.col("product_name"),F.col
l("opening_stock"),F.col("total_quantity_sold").alias("quantity_sold"),F.col("expected_closing_stock
"),F.col("actual_closing_stock"),F.col("discrepancy").alias("discrepancy_amount"))
# --- 5. Write to Processed Zone ---
final_df_selected.write.mode("overwrite").parquet(processed_output_path)
# --- 6. Check for Discrepancies and Send Notification ---
discrepancy_df = final_df_selected.filter(F.col("discrepancy_amount") != 0)
if discrepancy_df.count() > 0:
           print("Discrepancies found. Preparing to send SNS notification.")
           sns_client = boto3.client('sns', region_name=aws_region)
           discrepancy_examples = discrepancy_df.limit(5).collect()
           {\tt message = f"Inventory reconciliation for date \{processing\_date\_str\} found discrepancies. \verb|\n\| n = 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 1000 | 10
           message += f"Total items with discrepancies: {discrepancy df.count()}\n\n"
          message += "Example Discrepancies:\n"
           for row in discrepancy_examples:
                     product_name = row['product_name'] if row['product_name'] else "N/A"
message += f"- SKU: {row['sku']}, Product: {product_name}, Discrepancy:
 {row['discrepancy_amount']}\n"
           message += f"\nFull report available at: {processed_output_path}"
          subject = f"Alert: Inventory Discrepancy Found for {processing_date_str}"
           response = sns_client.publish(TopicArn=sns_topic_arn, Message=message, Subject=subject)
           print(f"SNS notification sent! Message ID: {response['MessageId']}")
           print("No discrepancies found. No notification sent.")
job.commit()
```

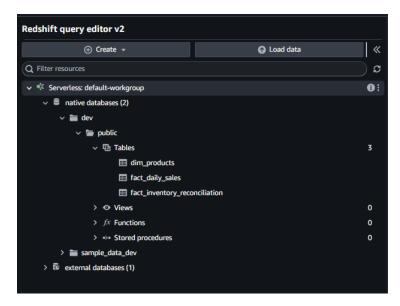


Step 6: Create a Redshift Serverless

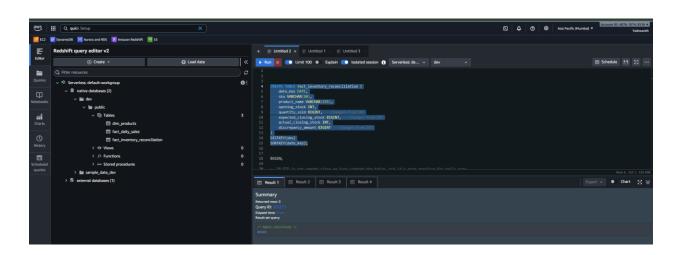




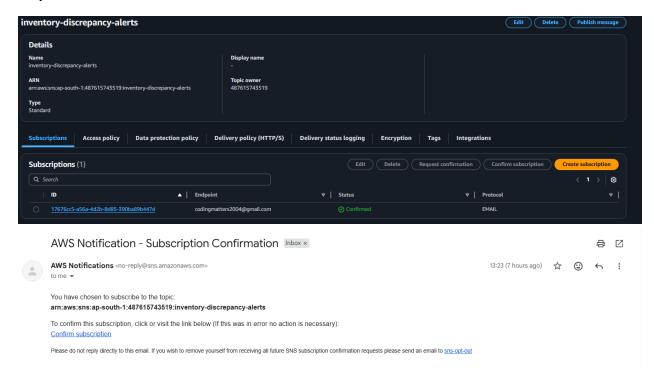
Step 7: Create Tables in Redshift



```
CREATE TABLE fact_inventory_reconciliation (
    date_key DATE,
sku VARCHAR(50),
product_name VARCHAR(255),
    opening_stock INT,
    quantity_sold BIGINT, -- Changed from INT expected_closing_stock BIGINT, -- Changed from INT
    actual_closing_stock INT,
discrepancy_amount BIGINT -- Changed from INT
DISTKEY(sku)
SORTKEY(date_key);
CREATE TABLE IF NOT EXISTS dim_products (
    sku VARCHAR(50) NOT NULL,
    product_name VARCHAR(255),
    category VARCHAR(100)
DISTSTYLE ALL
                     -- Use DISTSTYLE ALL because it's a small table that joins to large fact tables.
SORTKEY(sku);
CREATE TABLE IF NOT EXISTS fact_daily_sales (
    date key DATE,
    sku VARCHAR(50),
    product_name VARCHAR(255),
    total_quantity_sold BIGINT -- Using BIGINT to match the Spark sum() output
SORTKEY(date_key);
```



Step 8: Create Sns and Subscribe to email



Project Resource Summary

Here is a breakdown of the AWS services utilized in this data pipeline and the rationale for their use.

1. IAM (Identity and Access Management)

- Resource: An IAM Role.
- Purpose: Created to grant the necessary permissions for AWS services to communicate
 with each other securely. For example, this role allows the AWS Glue jobs to read data
 from the S3 bucket and write the processed data to the Amazon Redshift warehouse
 without needing to manage static credentials.

2. S3 (Simple Storage Service)

- Resource: An S3 Bucket.
- Purpose: Serves as the central data lake for this project. It is used to store raw source
 data (like daily sales files) and potentially intermediate, processed data before it's loaded
 into the data warehouse. Its scalability and cost-effectiveness make it ideal for storing
 large volumes of data.

3. AWS Glue

- Resource: Three distinct AWS Glue ETL jobs.
- **Purpose:** Used as the serverless Extract, Transform, and Load (ETL) service to process the raw data from S3.
 - Glue Job (dim_products): This job is responsible for processing the productrelated data to create and populate the dim_products dimension table. It cleans, transforms, and standardizes product information.
 - Glue Job 1 (fact_daily_sales): This job processes the daily transactional sales data, aggregating and structuring it to be loaded into the fact_daily_sales table.
 - Glue Job 2 (reconciliation): This job performs a reconciliation process, likely comparing inventory or sales data from different sources to ensure accuracy and consistency. The output is loaded into the fact_inventory_reconciliation table.

4. Amazon Redshift Serverless

- Resource: A Redshift Serverless data warehouse.
- Purpose: Acts as the project's central data warehouse. It stores the final, structured, and
 transformed data in a columnar format, which is highly optimized for fast analytical
 queries and business intelligence (BI) reporting. The serverless option was chosen to
 automatically scale compute resources based on workload, simplifying management and
 optimizing costs.

5. Redshift Tables

- Resource: Three tables within the Redshift warehouse.
- **Purpose:** To store the cleaned and processed data in a structured, queryable format.
 - dim_products: A dimension table that stores descriptive attributes about the products (e.g., name, category, brand).
 - fact_daily_sales: A fact table that stores quantitative sales metrics (e.g., units sold, revenue) on a daily basis.
 - fact_inventory_reconciliation: A fact table containing the results of the data reconciliation process, highlighting discrepancies or confirming consistency.

6. SNS (Simple Notification Service)

- **Resource:** An **SNS Topic** with an email subscription.
- Purpose: Implemented as an alerting and monitoring mechanism. It sends out
 automated email notifications to subscribers upon the success or failure of the Glue jobs
 or other pipeline events. This ensures that stakeholders are immediately aware of the
 pipeline's status.

Apache Airflow?

Airflow is used for **workflow orchestration**. While we have created individual jobs (like your Glue jobs), a real-world project requires you to run them in a specific order, on a schedule, and handle any failures gracefully.

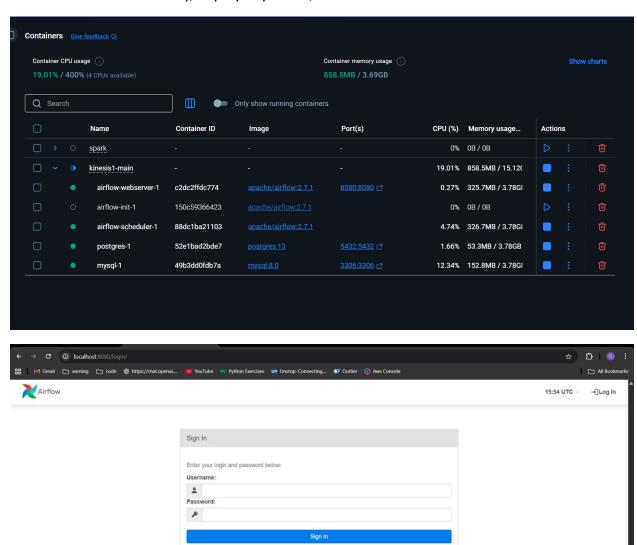
Airflow allows you to:

- **Schedule & Automate:** Automatically run your entire data pipeline at a specific time (e.g., daily at 2 AM) without manual intervention.
- Manage Dependencies: Define the exact order of operations. For example, you can
 ensure your fact_daily_sales and dim_products jobs complete successfully before the
 reconciliation job begins.
- Monitor & Alert: Provides a powerful user interface to visualize your pipeline's progress, check logs for each step, and send alerts when tasks fail.
- **Retry & Recover:** Automatically retries a failed task a set number of times, which can resolve temporary issues without waking you up in the middle of the night.
- **Scalability:** Manages complex workflows with many steps and dependencies, which is difficult to do with simple scripts or cron jobs.

Essentially, Airflow acts as the "brain" or "conductor" of your data pipeline, ensuring every part runs correctly, in the right order, and at the right time.

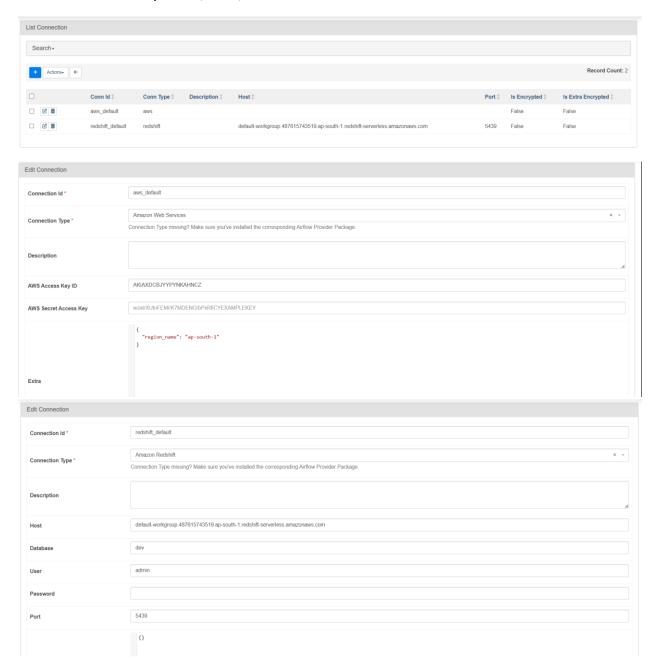
Set Up an Airflow Environment:

Docker Airflow: Build locally, deploy anywhere, and orchestrate with confidence.



Define Connections:

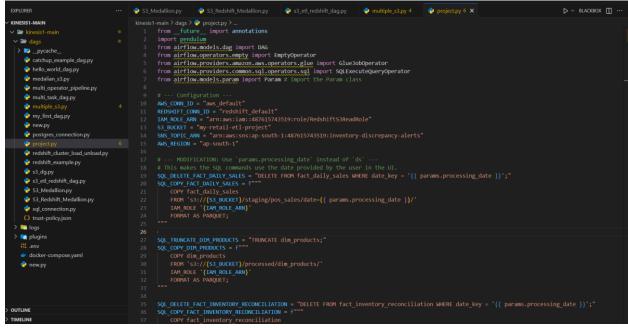
• In the Airflow UI, you configure connections to your external systems. You would create an **AWS Connection** that gives Airflow the permissions (via an IAM AccessKeys) to interact with your S3, Glue, and Redshift services.

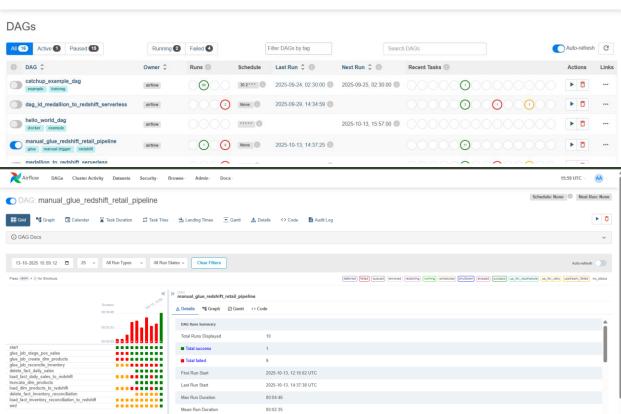


Airflow Dag:

```
future
                import annotations
import pendulum
from airflow.models.dag import DAG
from airflow.operators.empty import EmptyOperator
from airflow.providers.amazon.aws.operators.glue import GlueJobOperator
from airflow.providers.common.sql.operators.sql import SQLExecuteQueryOperator
from airflow.models.param import Param # Import the Param class
# --- Configuration ---
AWS CONN ID = "aws default"
REDSHIFT CONN ID = "redshift default"
IAM ROLE ARN = "arn:aws:iam::487615743519:role/RedshiftS3ReadRole"
S3_BUCKET = "my-retail-etl-project"
SNS_TOPIC_ARN = "arn:aws:sns:ap-south-1:487615743519:inventory-discrepancy-alerts"
AWS REGION = "ap-south-1"
# --- MODIFICATION: Use `params.processing_date` instead of `ds` ---
# This makes the SQL commands use the date provided by the user in the UI.
SQL_DELETE_FACT_DAILY_SALES = "DELETE FROM fact_daily_sales WHERE date_key = '{{
params.processing_date }}';
SQL_COPY_FACT_DAILY_SALES = f"""
    COPY fact_daily_sales
    FROM 's3://{S3_BUCKET}/staging/pos_sales/date={{ params.processing_date }}/'
IAM_ROLE '{IAM_ROLE_ARN}'
    FORMAT AS PARQUET;
SQL TRUNCATE DIM PRODUCTS = "TRUNCATE dim products;"
SQL_COPY_DIM_PRODUCTS = f"""
    COPY dim_products
    FROM 's3://{S3_BUCKET}/processed/dim_products/'
IAM_ROLE '{IAM_ROLE_ARN}'
    FORMAT AS PARQUET;
SQL_DELETE_FACT_INVENTORY_RECONCILIATION = "DELETE FROM fact_inventory_reconciliation WHERE date_key
  '{{ params.processing date }}';
SQL_COPY_FACT_INVENTORY_RECONCILIATION = f"""
    COPY fact_inventory_reconciliation
    FROM 's3://{S3_BUCKET}/processed/reconciled_inventory/date={{ params.processing_date }}/'
IAM_ROLE '{IAM_ROLE_ARN}'
    FORMAT AS PARQUET;
with DAG(
    dag_id="manual_glue_redshift_retail_pipeline",
    start_date=pendulum.datetime(2025, 1, 1, tz="UTC"),
    schedule=None,
    catchup=False,
    doc_md=""
    ## Manual Retail Data Pipeline
    Orchestrates AWS Glue jobs and Redshift loads.
    Click the play button and enter a `processing_date` to run.
    # --- MODIFICATION: Define a parameter for the user to enter ---
    params={
        "processing_date": Param(
            "2025-10-12", # This is the default value that will appear in the box
            type="string
```

```
title="Processing Date",
            description="The date for which to run the pipeline (format: YYYY-MM-DD).",
        )
    tags=["glue", "redshift", "manual-trigger"],
) as dag:
    start = EmptyOperator(task_id="start")
    # --- MODIFICATION: Use `params.processing date` in script args ---
   glue_job_stage_pos_sales = GlueJobOperator(task_id="glue_job_stage_pos_sales", job_name="glue-
job-1-stage-pos-sales", aws_conn_id=AWS_CONN_ID, script_args={"--processing_date": "{{
params.processing_date }}"})
    glue_job_create_dim_products = GlueJobOperator(task_id="glue_job_create_dim_products",
job name="glue-job-create-dim-products", aws conn id=AWS CONN ID, script args={"--processing date":
 {{ params.processing date }}"})
    glue_job_reconcile_inventory = GlueJobOperator(
        task id="glue_job_reconcile_inventory",
        job_name="glue-job-2-reconcile-inventory",
        aws conn id=AWS CONN ID,
        script_args={
             --processing_date": "{{ params.processing_date }}",
            "--s3_bucket": S3_BUCKET,
            "--aws region": AWS_REGION,
            "--sns_topic_arn": SNS_TOPIC_ARN
        }
    )
    # Redshift tasks now use the single-statement SQL variables
   delete_fact_daily_sales = SQLExecuteQueryOperator(task_id="delete_fact_daily_sales",
conn_id=REDSHIFT_CONN_ID, sql=SQL_DELETE_FACT_DAILY_SALES)
    load_fact_daily_sales = SQLExecuteQueryOperator(task_id="load_fact_daily_sales_to_redshift",
conn_id=REDSHIFT_CONN_ID, sql=SQL_COPY_FACT_DAILY_SALES)
    truncate dim products = SQLExecuteQueryOperator(task id="truncate dim products",
conn id=REDSHIFT CONN ID, sql=SQL TRUNCATE DIM PRODUCTS)
    load dim products = SQLExecuteQueryOperator(task id="load dim products to redshift",
conn_id=REDSHIFT_CONN_ID, sql=SQL_COPY_DIM_PRODUCTS)
delete_fact_inventory_reconciliation =
SQLExecuteQueryOperator(task_id="delete_fact_inventory_reconciliation", conn_id=REDSHIFT_CONN_ID,
sql=SQL DELETE FACT INVENTORY RECONCILIATION)
    load_fact_inventory_reconciliation =
SQLExecuteQueryOperator(task id="load fact inventory reconciliation to redshift",
conn id=REDSHIFT CONN ID, sql=SQL COPY FACT INVENTORY RECONCILIATION)
    end = EmptyOperator(task id="end")
    # --- Dependencies (Unchanged) ---
    start >> [glue_job_stage_pos_sales, glue_job_create_dim_products]
    glue_job_stage_pos_sales >> delete_fact_daily_sales >> load_fact_daily_sales
    glue_job_create_dim_products >> truncate_dim_products >> load_dim_products
    [glue_job_stage_pos_sales, glue_job_create_dim_products] >> glue_job_reconcile_inventory
    glue_job_reconcile_inventory >> delete_fact_inventory_reconciliation >>
load_fact_inventory_reconciliation
    [load_fact_daily_sales, load_dim_products, load_fact_inventory_reconciliation] >> end
```



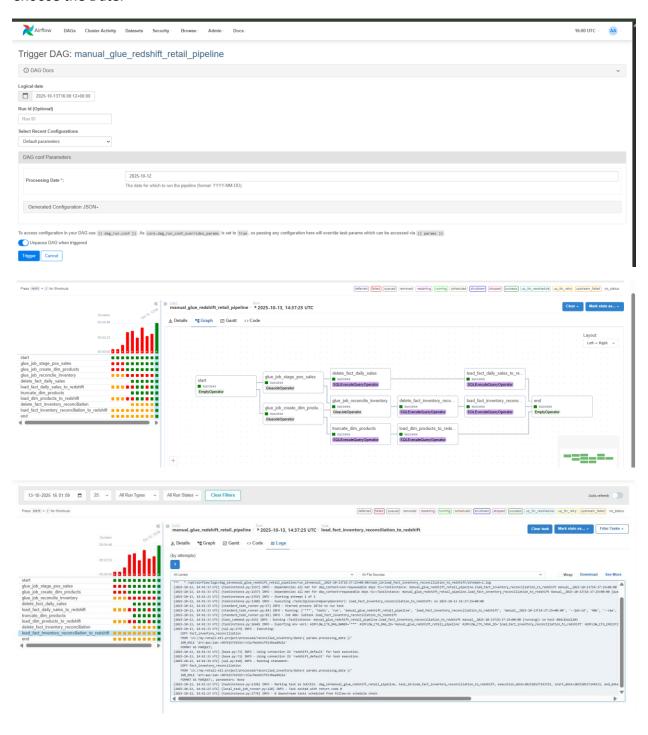


Min Run Duration

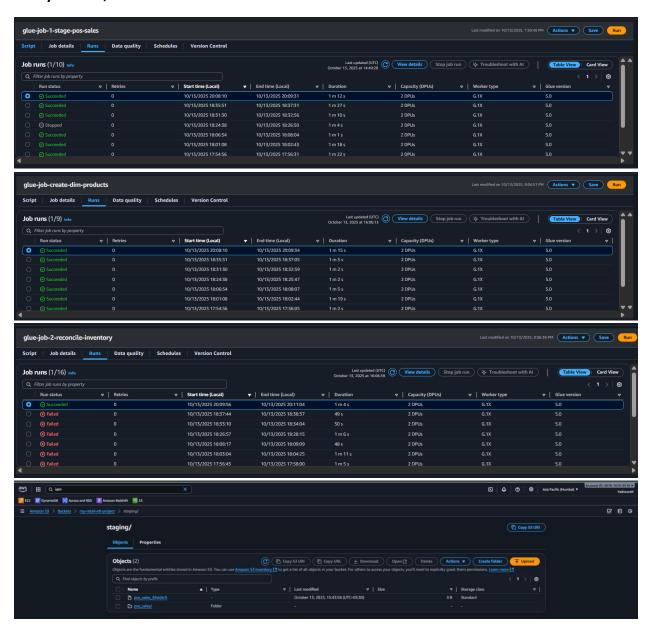
00:00:08

Run Dag:

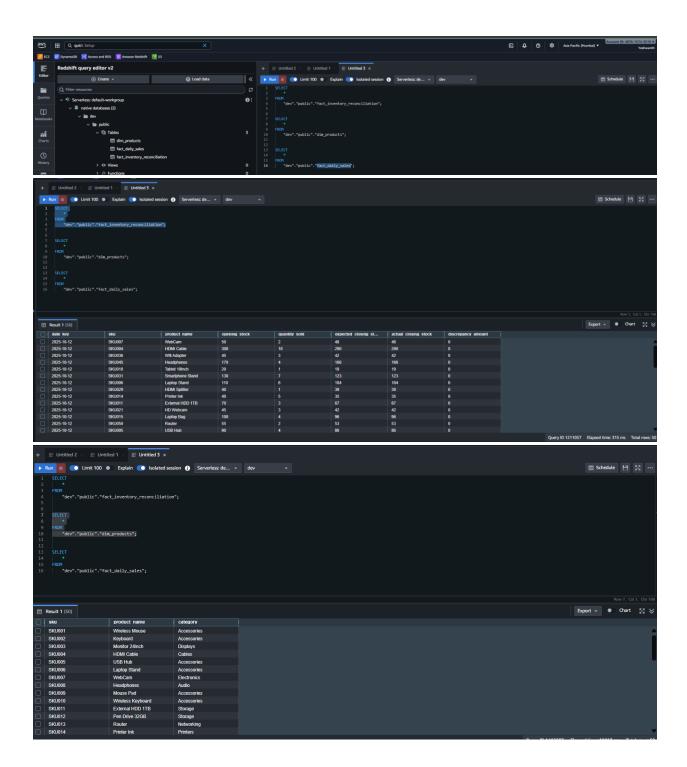
Choose the Date.

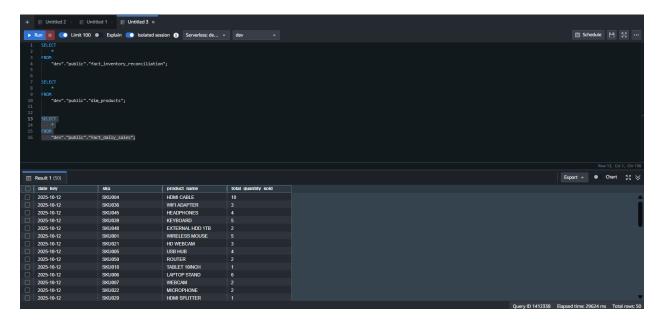


Verify in Glue, S3 and Redshift:

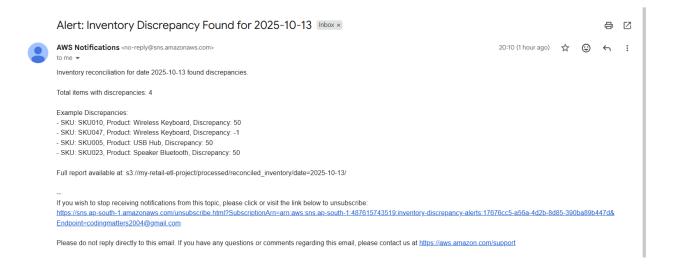








Check Email:



QuickSight:

