## ■ 06c\_customer\_churn\_airflow\_execution.md

# **©Customer Churn using Serverless Spark through Airflow.**

Goal - Data Preparation and Model Training for Detecting Customer Churn Dataset.

Following are the lob modules:

- 1. Understanding Data
- 2. Solution Diagram
- 3. Uploading DAG files to DAGs folder
- 4. Execution of Airflow DAG
- 5. BQ output tables
- 6. Logging

## **©1.** Understanding Data

The dataset used for this project are <u>customer churn data</u> and <u>customer test data</u>. The dataset contains the following features:

- Churn Binary field which represents customers who left/were retained within the last month
- Services that each customer has signed up for phone, multiple lines, internet, online security, online backup, device protection, tech support, and streaming TV and movies
- Customer account information how long they've been a customer, contract, payment method, paperless billing, monthly charges, and total charges
- Demographic info about customers gender, age range, and if they have partners and dependents

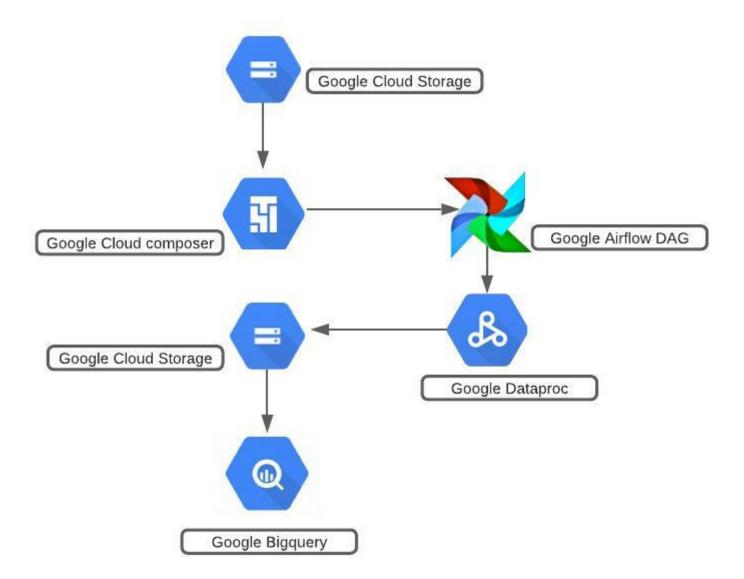
**Note:** The following features refer to these same-host connections.

- serror rate
- rerror\_rate
- same\_srv\_rate
- diff\_srv\_rate
- srv\_count

**Note:** The following features refer to these same-service connections.

- srv\_serror\_rate
- srv\_rerror\_rate
- srv\_diff\_host\_rate

# **2.** Solution Diagram



#### **Model Pipeline**

The model pipeline involves the following steps:

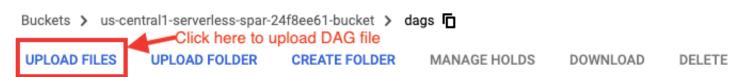
- Create buckets in GCS
- Create Dataproc and Persistent History Server Cluster
- Copy the raw data files, PySpark and notebook files into GCS
- Create a Cloud Composer environment and Airflow jobs to run the serverless spark job
- Creating Google BigQuery tables with summary of anomalous cell towers

# **3.** Uploading DAG files to DAGs folder

- From the code repository, download the file located at: customer\_churn>00-scripts>customer\_churn\_airflow.py
- Rename file to <your\_name\_here>-customer\_churn\_airflow.py
- Open the file and replace your name on row 21
- Navigate to Composer><composer\_environment>
- Next, navigate to Environment Configuration>DAGs folder URI
- Next, upload the DAG file to the GCS bucket corresponding to the DAGs folder URI

MONITORING	LOGS	DAGS PREVIEW	ENVIRONMENT CONFIGURATION			
Name		serverless-spark-composer				
Location		us-central1				
Service account		198454710197-compute@developer.gserviceaccount.com				
Image version		composer-2.0.5-airflow-2.2.3 UPGRADE				
		New version available. <u>Learn more</u>				
Python version		3				
DAGs folder		gs://us-central1-serverless-spar-24f8ee61-bucket/dags				

Location	Storage class	Public access	Protection		
us-central1 (lo	wa) Standard	Subject to object ACLs	None		
OBJECTS	CONFIGURATION	PERMISSIONS	PROTECTION	LIFECYCLE	



## **4.** Execution of Airflow DAG

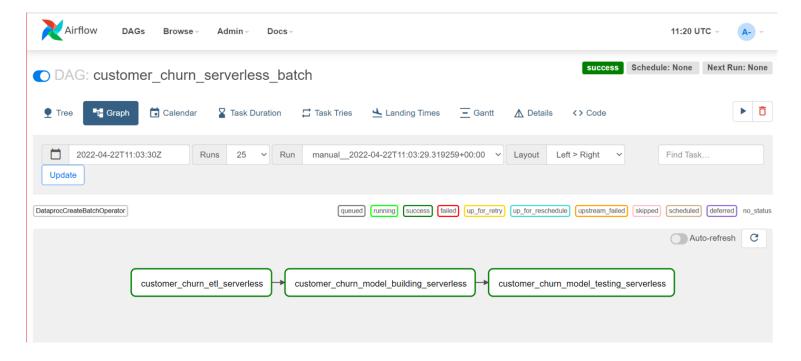
• Navigate to Composer><your\_environment>>Open Airflow UI



- Once the Airflow UI opens, navigate to **DAGs** and open your respective DAG
- Next, trigger your DAG by clicking on the **Trigger DAG** button



• Once the DAG is triggered, the DAG can be monitored directly through the Airflow UI as well as the Dataproc>Serverless>Batches window



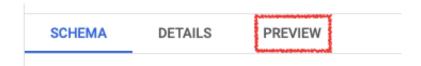
# **∞5. BQ** output tables

Navigate to BigQuery Console, and check the **customer-churn** dataset.

Once the Airflow DAG execution is completed, four new tables '<your\_name\_here>\_training\_data', '<your\_name\_here>\_test\_data', '<your\_name\_here>\_predictions\_data' and '<your\_name\_here>\_test\_output' are created.

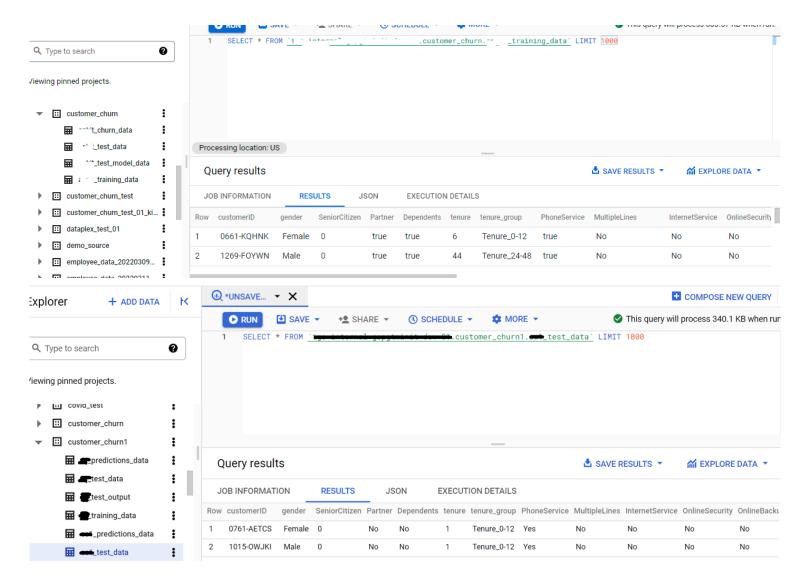
To view the data in these tables -

- Select the table from BigQuery Explorer by navigating 'project\_id' > 'dataset' > 'table\_name'
- Click on the **Preview** button to see the data in the table



**Note:** If the **Preview** button is not visible, run the below queries to view the data. However, these queries will be charged for the full table scan.

```
SELECT * FROM `'ct_name>.<dataset_name>.<your_name_here>_training_data` LIMIT 1000;
SELECT * FROM `ct_name>.<dataset_name>.<your_name_here>_test_data` LIMIT 1000;
SELECT * FROM `ct_name>.<dataset_name>.<your_name_here>_predictions_data` LIMIT 1000;
SELECT * FROM `ct_name>.<dataset_name>.<your_name_here>_test_output` LIMIT 1000;
```

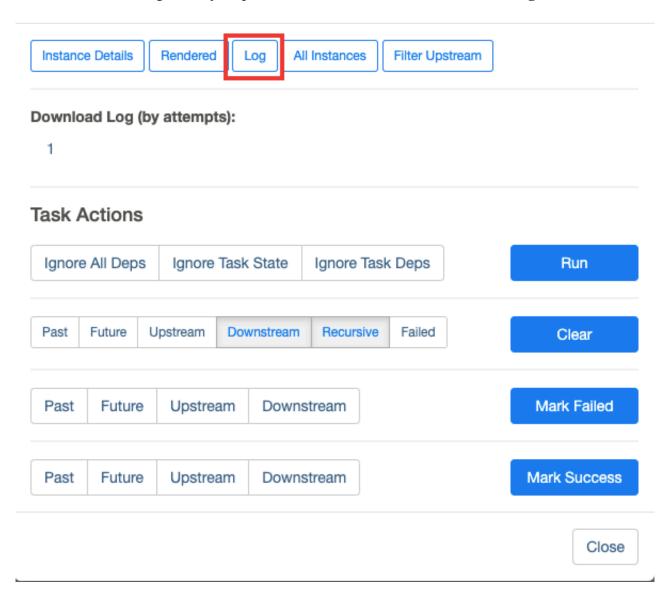


**Note:** Edit all occurrences of and <dataset\_name> to match the values of the variables PROJECT\_ID, and BQ\_DATASET\_NAME respectively

# **∞6.** Logging

## **℃6.1** Airflow logging

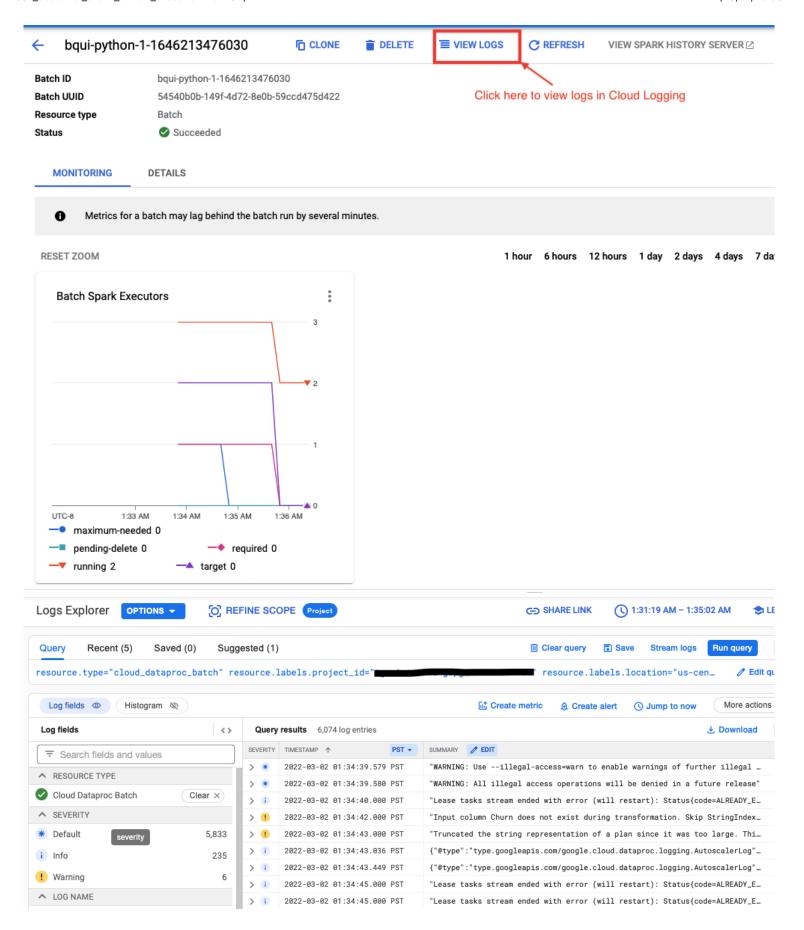
• To view the logs of any step of the DAG execution, click on the >Log button



## **%6.2** Serverless Batch logs

Logs associated with the application can be found in the logging console under **Dataproc > Serverless > Batches > <batch\_name>**.

You can also click on "View Logs" button on the Dataproc batches monitoring page to get to the logging page for the specific Spark job.



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### **%**6.3 Persistent History Server logs

To view the Persistent History server logs, click the 'View History Server' button on the Dataproc batches monitoring page and the logs will be shown as below:

