06b_customer_churn_console_execution.md

Customer Churn using Serverless Spark through Google Cloud Console

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

Following are the lab modules:

- 1. Understanding Data
- 2. Solution Architecture
- 3. Parameter Requirements for the lab
- 4. Data Preparation
- 5. Model Training and Testing
- 6. Model Evaluation
- 7. Logging

1. Understanding Data

The dataset used for this project is customer churn data and customer test data...

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

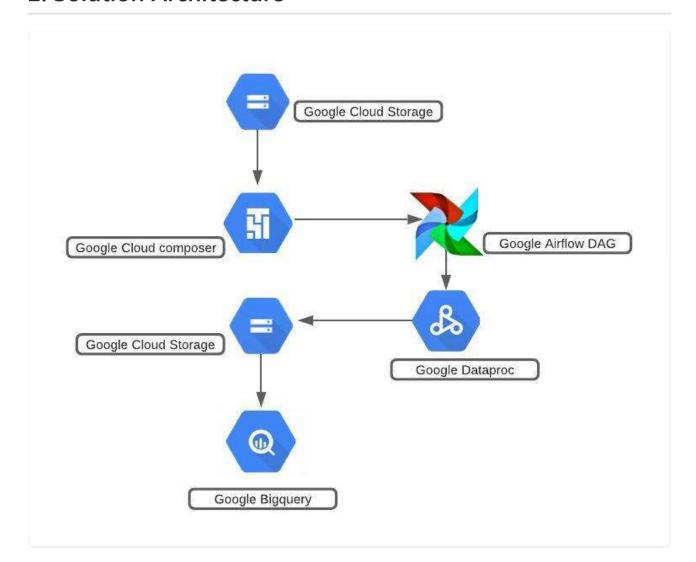
localhost:6419 1/14

- 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 Architecture



Model Pipeline

The model pipeline involves the following steps:

localhost:6419 2/14

- Data cleanup and preparation
- Building and training a Machine Learning Model (Random Forest Classifier) before saving it into a GCS bucket
- Using the model built in above step to evaluate test data

3. Parameters required for the lab

Keep the following details handy for configuring the serverless batch jobs:

```
#current GCP project where we ar
PROJECT ID=
REGION=
                                                     #GCP region where all our resour
BQ DATASET NAME=
                                                     #BigQuery dataset where all the
BUCKET CODE=
                                                     #GCP bucket where our code, data
                                                    #name of the history server which
HISTORY SERVER NAME=spark-phs
                                                    # Primary VPC containing the subn
VPC NAME=
SUBNET=
                                                     #subnet which has private google
UMSA=serverless-spark
                                                     #name of the user managed servic
SERVICE ACCOUNT=$UMSA@$PROJECT ID.iam.gserviceaccount.com
NAME=
                                                      #Your unique identifier
```

Note: The values for all the above parameters will be provided by the admin team.

4. Data Preparation

Based on EDA, the data preparation script has been created. Among the 21 columns, relevant features have been selected and stored in BQ for the next step of model training.

4.1. Create a new batch

Navigate to Dataproc > Serverless > Batches and click on +CREATE



localhost:6419 3/14

4.2. Provide the details for the batch

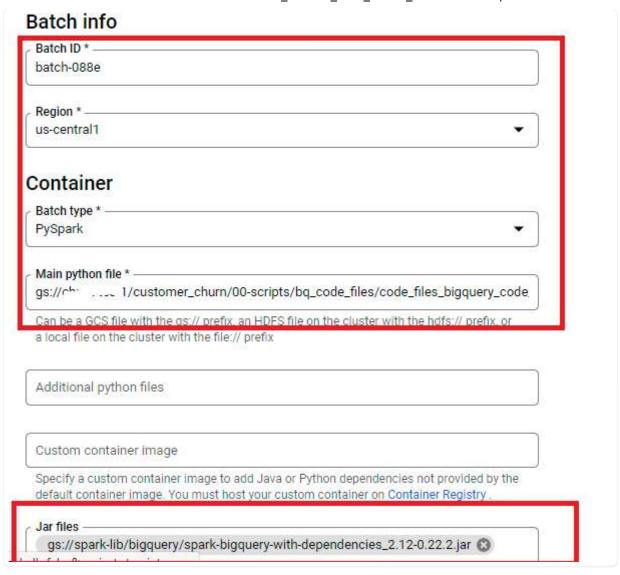
Next, fill in the following values in the batch creation window as shown in the images below:

- Batch ID A unique identifier for your batch
- Region The region name provided by the Admin team
- Batch Type PySpark
- Main Python File gs://<your_code_bucket_name>/customer_churn/00-scripts/customer_churn_data_prep.py
- JAR Files gs://spark-lib/bigguery/spark-bigguery-with-dependencies_2.12-0.22.2.jar
- Arguments -

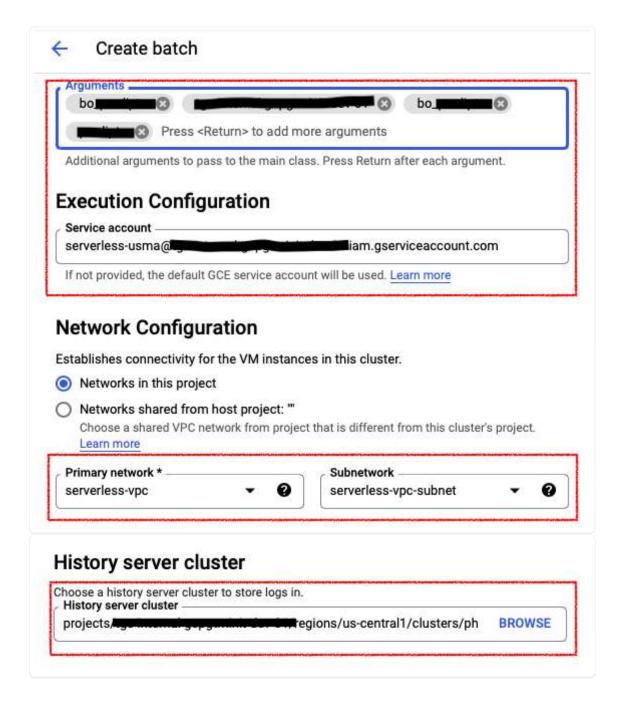
Four Arguments needs to be provided.

- <your_project_id>
- < <your_dataset_name>
- <your_code_bucket_name>
- <your_name> Note: Press RETURN after each argument
- Service Account <UMSA_NAME>@<PROJECT_ID>.iam.gserviceaccount.com
- **Network Configuration** select the network and subnetwork with Private Google Access Enabled Run PySpark Serverless Batch for Data Preparation
- History Server Cluster <your_phs_cluster_name>

localhost:6419 4/14



localhost:6419 5/14



4.3. Submit the Serverless batch

Once all the details are in, you can submit the batch. As the batch starts, you can see the execution details and logs on the console.

4.4. Check the output table in BQ

Navigate to BigQuery Console, and check the **customer_churn_lab** dataset.

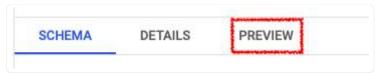
Once the data preparation batch is completed, new tables

'<your_name_here>_training_data' and '<your_name_here>_test_data' will be created.

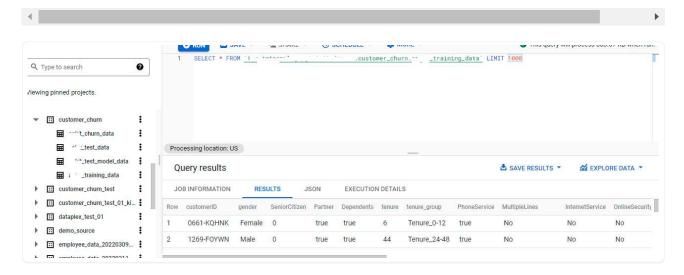
To view the data in these tables -

localhost:6419 6/14

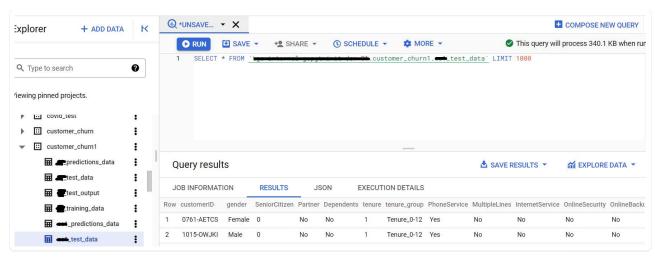
- Select the table from BigQuery Explorer by navigating 'project_id' > 'dataset' > 'table name'
- Click on the **Preview** button to see the data in the table



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.



<your_name_here>: _test_data table



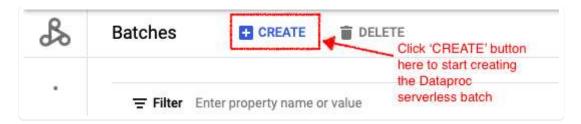
localhost:6419 7/14

5. Model Training and Testing

Repeat the same steps as above to submit another batch for model training.

5.1. Create a new batch

Navigate to Dataproc > Serverless > Batches and click on +CREATE



5.2. Provide the details for the batch

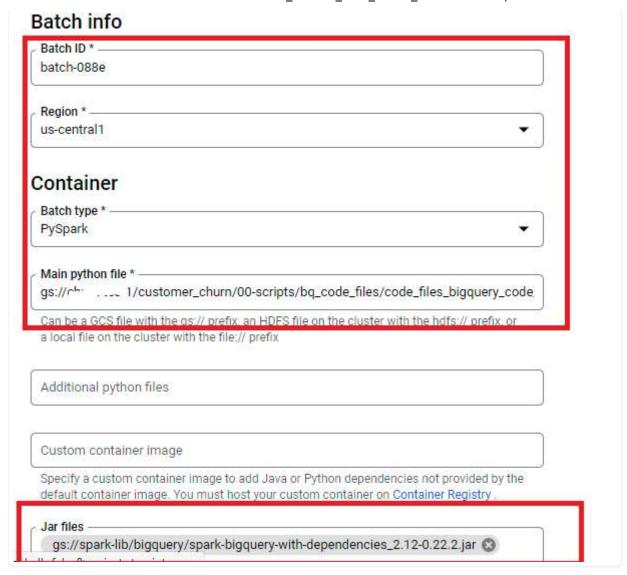
Next, fill in the following values in the batch creation window as shown in the images below:

- Batch ID A unique identifier for your batch
- Region The region name provided by the Admin team
- Batch Type PySpark
- Main Python File gs://<your_code_bucket_name>/customer_churn/00-scripts/customer_churn_model_building.py
- JAR Files gs://spark-lib/bigquery/spark-bigquery-with-dependencies_2.12-0.22.2.jar
- Arguments -

Four Arguments needs to be provided.

- <your_project_id>
- <your_dataset_name>
- <your_code_bucket_name>
- <your_name> Note: Press RETURN after each argument
- **Service Account** <UMSA_NAME>@<PROJECT_ID>.iam.gserviceaccount.com
- **Network Configuration** select the network and subnetwork with Private Google Access Enabled Run PySpark Serverless Batch for Data Preparation
- History Server Cluster <your_phs_cluster_name>

localhost:6419 8/14



5.3. Query the model_test results BQ table

Navigate to BigQuery Console, and check the **customer_churn_lab** dataset.

Once the modeling batch is completed, a new table '<your_name_here>_predictions_data' will be created.

To view the data in this table -

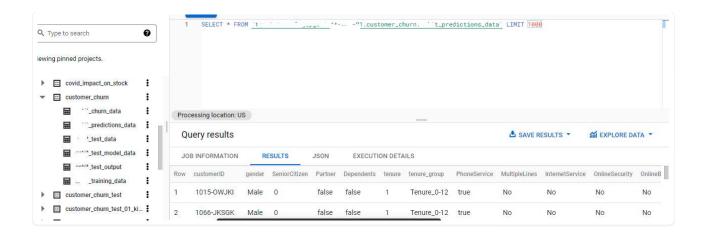
- Select the table from BigQuery Explorer by navigating 'project_id' > 'dataset' > 'table_name'
- Click on the **Preview** button to see the data in the table



localhost:6419 9/14

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 `roject_name>.<dataset_name>.<your_name_here>_predictions_data` LI



6. Model Evaluation

Repeat the same steps as above to submit another batch for model training.

6.1. Create a new batch

Navigate to Dataproc > Serverless > Batches and click on +CREATE



6.2. Provide the details for the batch

Next, fill in the following values in the batch creation window as shown in the images below:

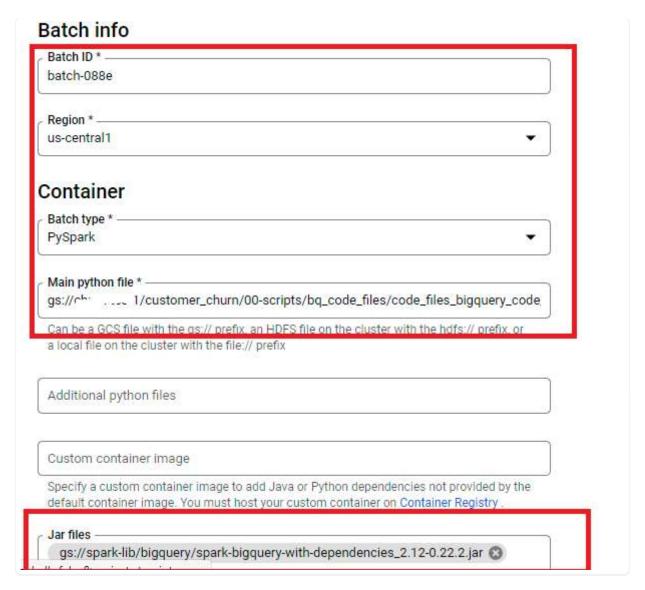
- Batch ID A unique identifier for your batch
- Region The region name provided by the Admin team

localhost:6419 10/14

- Batch Type PySpark
- Main Python File gs://<your_code_bucket_name>/customer_churn/00-scripts/customer_churn_model_testing.py
- JAR Files gs://spark-lib/bigguery/spark-bigguery-with-dependencies_2.12-0.22.2.jar
- Arguments -

Four Arguments needs to be provided.

- <your_project_id>
- < <your_dataset_name>
- <your_code_bucket_name>
- <your_name> Note: Press RETURN after each argument
- **Service Account <UMSA_NAME>@<PROJECT_ID>.iam.gserviceaccount.com**
- Network Configuration select the network and subnetwork with Private Google Access Enabled Run PySpark Serverless Batch for Data Preparation
- History Server Cluster <your_phs_cluster_name>



localhost:6419 11/14

6.3. Query the model_test results BQ table

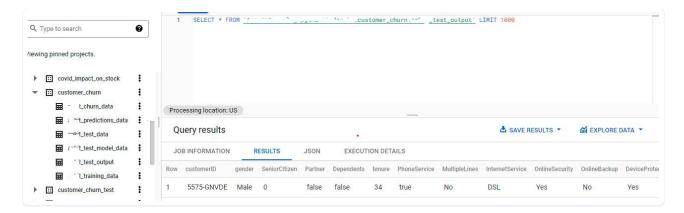
Navigate to BigQuery Console, and check the **customer_churn_lab** dataset. Once the model_testing batch is completed, a new table '<your_name_here>_test_output' will be created.

To view the data in this table -

- 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.



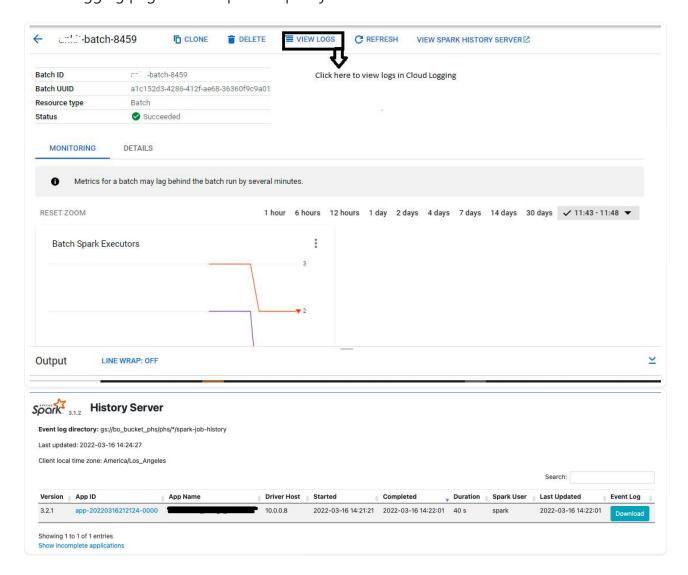
7. Logging

7.1 Serverless Batch logs

localhost:6419 12/14

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

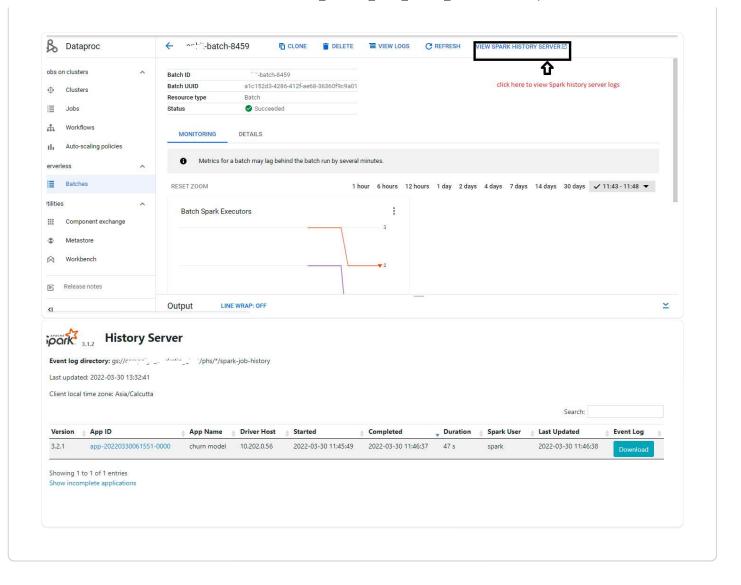
You can also click on "View Logs" button on the Dataproc batches monitoring page to get to the logging page for the specific Spark job.



7.2 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:

localhost:6419 13/14



localhost:6419 14/14