# Week 8: Homework 2: Project: Movie Recommendation with MLlib - Collaborative Filtering (implementation 3)

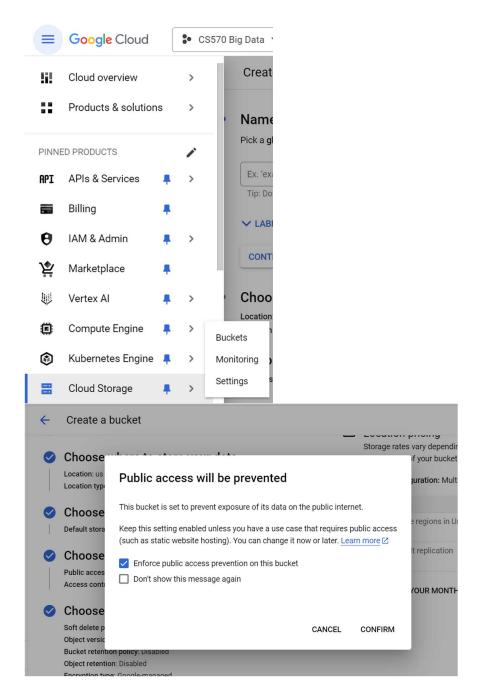
Step-by-Step Guide for Deployment on GCP with Correct File Paths

1. Upload Data and Scripts to GCS

Ensure that movies.csv, ratings.csv, and your PySpark script (e.g., Recommendation\_Engine\_MovieLens.py) are already uploaded to your GCS bucket.

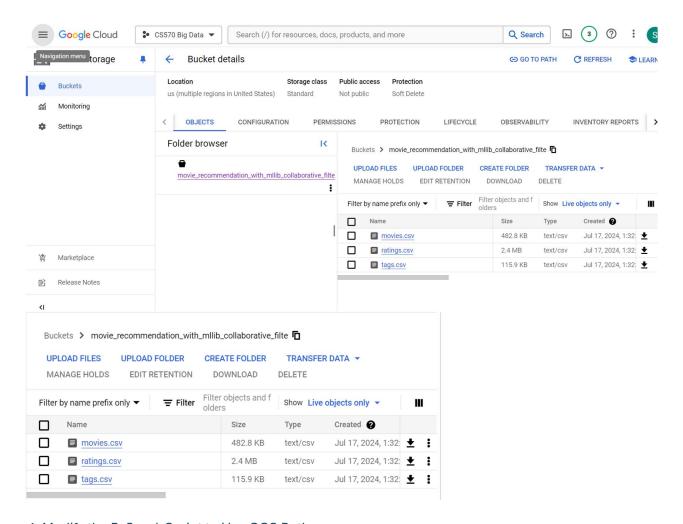
2. Create a Google Cloud Storage (GCS) Bucket

Create a bucket in GCS to store your scripts and data:



# 3. Upload Data and Scripts to GCS

Upload the movies.csv, ratings.csv, and your PySpark script to your GCS bucket:



# 4. Modify the PySpark Script to Use GCS Paths

Update your PySpark script to read the files from GCS. Use command-line arguments to pass the paths of the CSV files, enhancing the script's flexibility. Upload the modified script to the bucket:

```
shagos90499@cloudshell:~ (cs570-big-data-424809)$ vim Recommendation_Engine_MovieLens.py shagos90499@cloudshell:~ (cs570-big-data-424809)$
```

# Here is the script:

```
from pyspark.sql import SparkSession
from pyspark.sql.functions import col, explode
from pyspark.ml.evaluation import RegressionEvaluator
from pyspark.ml.recommendation import ALS
from pyspark.ml.tuning import ParamGridBuilder, CrossValidator
import argparse

# Parse command-line arguments
parser = argparse.ArgumentParser()
parser.add argument('--input path movies', required=True)
```

```
parser.add argument('--input path ratings', required=True)
args = parser.parse args()
# Initialize Spark session
spark = SparkSession.builder.appName('Recommendations').getOrCreate()
# Load data from GCS
movies = spark.read.csv(args.input path movies, header=True)
ratings = spark.read.csv(args.input path ratings, header=True)
# Preprocess data
ratings = ratings \
    .withColumn('userId', col('userId').cast('integer')) \
    .withColumn('movieId', col('movieId').cast('integer')) \
    .withColumn('rating', col('rating').cast('float')) \
    .drop('timestamp')
# Split data into training and testing sets
(train, test) = ratings.randomSplit([0.8, 0.2], seed=1234)
# Build ALS model
als = ALS(userCol="userId", itemCol="movieId", ratingCol="rating",
nonnegative=True, implicitPrefs=False, coldStartStrategy="drop")
param grid = ParamGridBuilder() \
    .addGrid(als.rank, [10, 50, 100, 150]) \
    .addGrid(als.regParam, [.01, .05, .1, .15]) \
evaluator = RegressionEvaluator(metricName="rmse", labelCol="rating",
predictionCol="prediction")
cv = CrossValidator(estimator=als, estimatorParamMaps=param grid,
evaluator=evaluator, numFolds=5)
# Train model
model = cv.fit(train)
best model = model.bestModel
# Evaluate model
test predictions = best model.transform(test)
RMSE = evaluator.evaluate(test predictions)
print(f"Root-mean-square error = {RMSE}")
# Generate recommendations
nrecommendations = best model.recommendForAllUsers(10)
nrecommendations = nrecommendations \
    .withColumn("rec_exp", explode("recommendations")) \
    .select('userId', col("rec exp.movieId"), col("rec exp.rating"))
nrecommendations.show()
# Join with movie titles for better interpretability
nrecommendations.join(movies, on='movieId').filter('userId = 100').show()
ratings.join(movies, on='movieId').filter('userId = 100').sort('rating',
ascending=False).limit(10).show()
# Stop Spark session
spark.stop()
```

```
# Build ALS model
  als = ALS(userCol="userId", itemCol="movieId", ratingCol="rating", nonnegative=True, implicitPrefs=False,
  param_grid = ParamGridBuilder() \
    (train, test) = ratings.randomSplit([0.8, 0.2], seed=1234)
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  param_grid = ParamGridBuilder() \
           .addGrid(als.rank, [10, 50, 100, 150]) \
.addGrid(als.regParam, [.01, .05, .1, .1
              .build()
  evaluator = RegressionEvaluator(metricName="xmse", labelCol="rating", predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionCol="predictionC
  cv = CrossValidator(estimator=als, estimatorParamMaps=param grid, evaluator=evaluator, numFolds=5)
  model = cv.fit(train)
  best_model = model.bestModel
  test_predictions = best_model.transform(test)
  RMSE = evaluator.evaluate(test_predictions)
  print(f"
  nrecommendations = best model.recommendForAllUsers(10)
  nrecommendations = nrecommendations
          .withColumn("rec_exp", explode("recommendations'
.select('userId', col("rec_exp.mcvieId"), col(":
ecommendations.show()
  nrecommendations.show()
  # Join with movie titles for better interpretability
nrecommendations.join(movies, on='movieta').filter(')
                                                                                                                                                              lserId = 100').show()
100').sort('reting', ascending=False).limit(10).show()
  ratings.join(movies, on= movieid').filter(
Upload the file:
```

```
gsutil cp Recommendation Engine MovieLens.py
gs://movie recommendation with mllib collaborative filte
```

```
shagos90499@cloudshell:~ (cs570-big-data-424809)$ gsutil cp Recommendation_Engine_MovieLens.py gs://movie_recommendation_with_mllib_collaborative_filte
Copying file://Recommendation_Engine_MovieLens.py [Content-Type=text/x-python]...
/ [1 files] [ 2.2 KiB/ 2.2 KiB]
Operation completed over 1 objects/2.2 KiB.
shagos90499@cloudshell:~ (cs570-big-data-424809)$
```

## 5. Create the Cluster with the Desired Configuration

## Create a Dataproc cluster:

```
gcloud dataproc clusters create spark-cluster \
    --region us-west1 \
    --zone us-west1-a \
    --master-machine-type n1-standard-4 \
    --worker-machine-type n1-standard-4 \
    --num-workers 2
```

```
shagos90499@cloudshell:~ (cs570-big-data-424809)$ gcloud dataproc clusters create spark-cluster \
--region us-west1 - \
--zone us-west1 - \
--master-machine-type nl-standard-4 \
--morker-machine-type nl-standard-4 \
--num-workers 2
Waiting on operation [projects/cs570-big-data-424809/regions/us-westl/operations/65ebb5fa-f4e4-3975-958f-82a592bed5cf].
Waiting for cluster creation operation...
WARNING: No image specified. Using the default image version. It is recommended to select a specific image version in production, as the default image version may change at any time.
WARNING: Consider using Auto Zone rather than selecting a zone manually. See https://cloud.google.com/dataproc/docs/concepts/configuring-clusters/auto-zone
WARNING: Failed to validate permissions required for default service account: '72008339659-compute@developer.gserviceaccount.com'. Cluster creation could still
be successful if required permissions have been granted to the respective service accounts as mentioned in the document https://cloud.google.com/dataproc/docs/co
ncepts/configuring-clusters/service-accounts#dataproc_service_accounts_2. This could be due to Cloud Resource Manager API hasn't been enabled in your recommended to update by visiting 'https://console.developers.google.com/apis/api/cloudresourcemanager.googleapis.com/overviev?project-
720083396599' before or it is disabled. Enable it by visiting 'https://console.developers.google.com/apis/apis/cloudresourcemanager.googleapis.com/overviev?project-
720083396599'.
WARNING: The firewall rules for specified network or subnetwork would allow ingress traffic from 0.0.0.0/0, which could be a security risk.
WARNING: The firewall rules for specified network or subnetwork would allow ingress traffic from 0.0.0.0/0, which could be a security risk.
WARNING: The firewall rules for specified network or subnetwork would allow ingress traffic from 0.0.0.0/0, which could be a security risk.
WARNING: The specified custom staging bucket 'dataproc-staging-us-west1-72008339699-ncaxf7jl' is not usin
```

### 6. Submit the PySpark Job with GCS Paths

• Submit your PySpark job to the Dataproc cluster, specifying the GCS paths for the input files:

Replace your-bucket-name with the actual name of your GCS bucket. By following these steps, your PySpark script will correctly read the files from GCS when running on GCP Dataproc.

Test result: root mean square error is calculated.

```
Root-mean-square error = 0.8685666272031658
+----+
|userId|movieId| rating|
    471|
         3379| 4.822564|
         8477|4.6659493|
    471|
    471| 33649|4.5504856|
    471 | 102217 | 4.5333 |
    471 | 92494 |
                 4.5333|
    471 | 33779 | 4.5333 |
    471 | 171495 | 4.527984 |
         7096|4.4821672|
    471|
    471| 84273|4.4345856|
    471 | 117531 | 4.4345856 |
     31| 33649|5.0889573|
     31 I
         3379|4.9877176|
     31 I
         6086| 4.85124|
     31| 3200| 4.813297|
     31| 171495| 4.79994|
     31 | 93988 | 4.786241 |
     31 | 184245 | 4.7817674 |
     31 | 84273 | 4.7817674 |
     31 | 26073 | 4.7817674 |
          7071|4.7817674|
only showing top 20 rows
```

```
|movieId|userId| rating|
                                         title|
           100|5.1201425|Strictly Sexual (...|Comedy|Drama|Romance|
  676181
           100| 5.064743| On the Beach (1959)|
  33791
           100| 5.042285| Glory Road (2006)|
  427301
           100| 5.021657| Saving Face (2004)|Comedy|Drama|Romance| 100|4.9267745| Watermark (2014)| Documentary|
  336491
 117531|
           100|4.9267745|Woman Under the I...|
   70711
                                                         Dramal
 184245|
           100|4.9267745|De platte jungle ...|
                                                        Documentary|
  260731
           100|4.9267745|Human Condition I...|
                                                        Drama|War|
 179135|
           100|4.9267745|Blue Planet II (2...|
                                                        Documentary|
           100|4.9267745|Zeitgeist: Moving...|
  842731
                                                        Documentary|
|movieId|userId|rating|
                                    title|
                                               genres|
                   5.0| Top Gun (1986)| Action|Romance|
   1101|
            100|
                                                 Comedy|Drama|
   1958|
           100|
                  5.0|Terms of Endearme...|
   24231
           100|
                  5.0|Christmas Vacatio...|
                                                          Comedy
                                                 Drama|Romance|
                  5.0|Officer and a Gen...|
   4041|
           100|
                   5.0|Sweet Home Alabam...| Comedy|Romance|
   56201
           100|
                         Maverick (1994) | Adventure | Comedy | . . . |
    3681
            100|
                   4.5|
    9341
            1001
                   4.5|Father of the Bri...| Comedy|
    5391
            100|
                   4.5|Sleepless in Seat...|Comedy|Drama|Romance|
           1001
                          Casino (1995) | Crime|Drama|
     161
                   4.51
                   4.5|
    5531
           100|
                           Tombstone (1993) | Action | Drama | Western |
```

```
d24303a13656f17a345a38/driveroutput
jobUuid: 8494b30d-d291-35e9-a8b3-547accce96de
placement:
 clusterName: spark-cluster
clusterUuid: a61a1a11-d3e2-46ad-8b23-ecfcaf287da7
pysparkJob:
  args:
  --input_path_movies=gs://movie_recommendation_with_mllib_collaborative_filte/movies.csv
---input_path_ratings=gs://movie_recommendation_with_mllib_collaborative_filte/ratings.csv
mainPythonFileUri: gs://movie_recommendation_with_mllib_collaborative_filte/Recommendation_Engine_MovieLens.py
reference:
  jobId: e4151b89e5d24303a13656f17a345a38
projectId: cs570-big-data-424809
status:
  state: DONE
  stateStartTime: '2024-07-17T09:27:50.761112Z'
statusHistory:
- state: PENDING
  stateStartTime: '2024-07-17T09:00:46.828315Z'
- state: SETUP_DONE
  stateStartTime: '2024-07-17T09:00:46.862237Z'
- details: Agent reported job success
  state: RUNNING
  stateStartTime: '2024-07-17T09:00:47.054125Z'
yarnApplications:
  name: Recommendations
  progress: 1.0
  state: FINISHED
  trackingUrl: http://spark-cluster-m:8088/proxy/application_1721206458042_0002/
```