

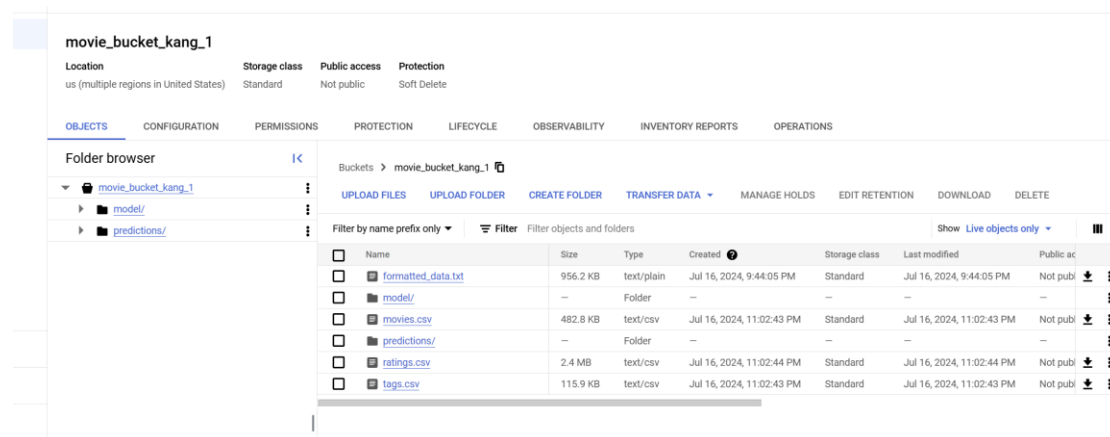
Project: Movie Recommendation with MLlib - Collaborative Filtering (implementaiton 3)

[TianzeKang2000/Movie-Recommendation-System \(github.com\)](https://github.com/TianzeKang2000/Movie-Recommendation-System)

Using the GCP Console to Upload the File:

Navigate to Cloud Storage, create a new bucket.

Click the "Upload Files" button and select C:\Users\KANG\Downloads\formatted_data.txt to upload.



Create a new Python script named recommendation.py:

Import Libraries

```
from pyspark.sql import SparkSession
```

```
from pyspark.mllib.recommendation import ALS, Rating
```

- **SparkSession**: Entry point to programming Spark with the DataFrame and SQL API.
- **ALS**: Alternating Least Squares, a collaborative filtering algorithm for recommender systems.
- **Rating**: Class used to store user, product, and rating information.

Initialize Spark Session

```
spark = SparkSession.builder.appName("MovieRecommendation").getOrCreate()
```

```
sc = spark.sparkContext
```

- Initializes a Spark session named "MovieRecommendation".
- sc is the SparkContext object, which is the entry point for using Spark functionality.

Load and Parse the Data

```
data = sc.textFile("gs://movie_bucket_kang_1/formatted_data.txt")
```

```
ratings = data.map(lambda l: l.split(',')) \
    .map(lambda l: Rating(int(l[0]), int(l[1]), float(l[2])))
```

- Loads the data from Google Cloud Storage (GCS) as a text file.
- Splits each line of the file by commas and maps it to a Rating object containing user, product, and rating.

Train the Recommendation Model Using ALS

```
rank = 10
```

```
numIterations = 10
```

```
model = ALS.train(ratings, rank, numIterations)
```

- **rank:** Number of latent factors in the model.
- **numIterations:** Number of iterations to run the ALS algorithm.
- Trains the ALS model using the provided ratings data.

Generate Predictions

```
users_products = ratings.map(lambda r: (r.user, r.product))
```

```
predictions = model.predictAll(users_products).map(lambda r: (r.user, r.product, r.rating))
```

- Creates a list of (user, product) pairs from the ratings data.
- Uses the trained model to predict ratings for all user-product pairs.
- Maps the predictions to include user, product, and predicted rating.

Convert Predictions to DataFrame

```
predictions_df = predictions.toDF(["user", "product", "rating"])
```

- Converts the predictions RDD to a DataFrame with columns "user", "product", and "rating".

Save Predictions as a Single CSV File to GCS

```
predictions_df.coalesce(1).write.mode("overwrite").option("header",
```

```
"true").csv("gs://movie_bucket_kang_1/predictions/predictions.csv")
```

- **coalesce(1):** Ensures the DataFrame is written to a single CSV file instead of multiple part files.
- **mode("overwrite"):** If the file already exists, it will be overwritten.
- **option("header", "true"):** Writes the DataFrame with a header row.
- **csv("gs://movie_bucket_kang_1/predictions/predictions.csv"):** Specifies the GCS path where the CSV file will be saved.



```
1 from pyspark.sql import SparkSession
2 from pyspark.mllib.recommendation import ALS, Rating
3
4 # Initialize Spark session
5 spark = SparkSession.builder.appName("MovieRecommendation").getOrCreate()
6 sc = spark.sparkContext
7
8 # Load and parse the data
9 data = sc.textFile("gs://movie_bucket_kang_1/formatted_data.txt")
10 ratings = data.map(lambda l: l.split(',')) \
11     .map(lambda l: Rating(int(l[0]), int(l[1]), float(l[2])))
12
13 # Train the recommendation model using ALS
14 rank = 10
15 numIterations = 10
16 model = ALS.train(ratings, rank, numIterations)
17
18 # Generate predictions
19 users_products = ratings.map(lambda r: (r.user, r.product))
20 predictions = model.predictAll(users_products).map(lambda r: (r.user, r.product, r.rating))
21
22 # Convert predictions to DataFrame
23 predictions_df = predictions.toDF(["user", "product", "rating"])
24
25 # Save predictions as a single CSV file to GCS
26 predictions_df.coalesce(1).write.mode("overwrite").option("header", "true").csv("gs://movie_bucket_kang_1/predictions/predictions.csv")
27
28 spark.stop()
29
```

Submit the PySpark Job

In Cloud Shell, submit the PySpark job to your Dataproc cluster using the following command:

```
gcloud dataproc jobs submit pyspark recommendation_example.py --cluster=cluster-a9c6 -  
-region=us-central1
```

```

Job [d3d2a26a1d9848cfaddc97ef209f33a8] submitted.
Waiting for job output...
24/07/17 06:15:50 INFO SparkEnv: Registering MapOutputTracker
24/07/17 06:15:50 INFO SparkEnv: Registering BlockManagerMaster
24/07/17 06:15:51 INFO SparkEnv: Registering BlockManagerMasterHeartbeat
24/07/17 06:15:51 INFO SparkEnv: Registering OutputCommitCoordinator
24/07/17 06:15:52 INFO DefaultMetricsWrapper: Connecting to ResourceManager at cluster-a9c6-m-us-central-f-c.enhanced-mote-424120-m9.internal/10.128.0.14:8032
24/07/17 06:15:52 INFO AHSPxy: Connecting to Application History server at cluster-a9c6-m-us-central-f-c.enhanced-mote-424120-m9.internal/10.128.0.14:10200
24/07/17 06:15:53 INFO Configuration: resource-types.xml not found
24/07/17 06:15:53 INFO ResourceConf: Unable to find 'resource-types.xml'.
24/07/17 06:15:54 INFO YarnClientImpl: Submitted application application_1721188828903_0005
24/07/17 06:15:55 INFO DefaultMetricsWrapper: Connecting to ResourceManager at cluster-a9c6-m-us-central-f-c.enhanced-mote-424120-m9.internal/10.128.0.14:8030
24/07/17 06:15:57 INFO MetricsConfig: loaded properties from hadoop-metrics2.properties
24/07/17 06:15:57 INFO MetricsSystemImpl: Scheduled metric snapshot period at 10 second(s).
24/07/17 06:15:57 INFO MetricsSystemImpl: google-hadoop-file-system-metric system started
24/07/17 06:15:58 INFO GoogleCloudStorageImpl: Ignoring exception of type GoogleJsonResponsesException verified object already exists with desired state.
24/07/17 06:15:59 INFO GoogleCloudStorageImpl: W-egp due to rate limit [RateLimitError(status=429)]: readers will not yet see flushed data for gs://dataproc-temp-us-central-1-cb7c60-2a91-4ffa-ab1b-534d70ee32e/spark-job-history/application_1721188828903_0005.inprogress [CONTEXT ratelimit_period=1 MINUTES ]
24/07/17 06:16:00 INFO FileInputFormat: Total input files to process is 1
24/07/17 06:16:02 INFO FileOutputCommitFactory: No output committer factory defined, defaulting to FileOutputCommitFactory
24/07/17 06:16:13 INFO GoogleCloudStorageFilesystemImpl: Successfully repaired 'gs://movie_bucket_kang_1/predictions/predictions.csv/' directory.
Job [d3d2a26a1d9848cfaddc97ef209f33a8] finished successfully.
done: true
driverControlFilesDir: gs://dataproc-staging-us-central-425505028363-7ewrrdz/google-cloud-dataproc-metainfo/0cb7c60-2a91-4ffa-ab1b-534d70ee32e/jobs/d3d2a26a1d9848cfaddc97ef209f33a8/
driverOutputResourceUrl: gs://dataproc-staging-us-central-425505028363-7ewrrdz/google-cloud-dataproc-metainfo/0cb7c60-2a91-4ffa-ab1b-534d70ee32e/jobs/d3d2a26a1d9848cfaddc97ef209f33a8/driver
jobId: 2113610c-9f8f-3415-93e4-10989a279255
placement:
  clusterName: cluster-a9c6
  clusterUuid: 0cb7c60-2a91-4ffa-ab1b-534d70ee32e
  pySparkJob:
    mainPythonFileDir: gs://dataproc-staging-us-central-425505028363-7ewrrdz/google-cloud-dataproc-metainfo/0cb7c60-2a91-4ffa-ab1b-534d70ee32e/jobs/d3d2a26a1d9848cfaddc97ef209f33a8/staging/f
  y
  references:
    jobId: d3d2a26a1d9848cfaddc97ef209f33a8
    projectId: enhanced-mote-424120-m9
  status:
    state: DONE
    stateStartTime: '2024-07-17T06:16:38.390601Z'
    stateHistory:
      - state: PENDING
        stateStartTime: '2024-07-17T06:15:43.503474Z'
      - state: SETUP_DONE
        stateStartTime: '2024-07-17T06:15:43.529512Z'
      - details: Agent reported job success
      - state: RUNNING
        stateStartTime: '2024-07-17T06:15:43.782305Z'
    yarnApplications:
      - name: MovieRecommender
        progress: 1.0
        state: FINISHED
    trackingUrl: http://cluster-a9c6-m-us-central-f-c.enhanced-mote-424120-m9.internal/8088/proxy/application_1721188828903_0005/

```

Output

```
24/07/17 06:19:52 INFO FileInputFormat: Total input files to process : 1
Top 20 Predictions:
User: 676, Product: 169, Rating: 6.172566597659339
User: 38, Product: 451, Rating: 6.153389792590913
User: 887, Product: 763, Rating: 6.140956443979752
User: 462, Product: 313, Rating: 6.10545380688081
User: 580, Product: 250, Rating: 6.086721429903589
User: 306, Product: 19, Rating: 5.978829222553157
User: 137, Product: 96, Rating: 5.959157362579283
User: 887, Product: 410, Rating: 5.878580775382083
User: 628, Product: 333, Rating: 5.86916227710036
User: 239, Product: 179, Rating: 5.856604903940478
User: 887, Product: 90, Rating: 5.8431491140332295
User: 264, Product: 173, Rating: 5.842799062589959
User: 562, Product: 143, Rating: 5.830419157452209
User: 366, Product: 53, Rating: 5.827840310697544
User: 180, Product: 694, Rating: 5.821170233887927
User: 337, Product: 520, Rating: 5.80567877970879
User: 42, Product: 496, Rating: 5.796843500314828
User: 532, Product: 313, Rating: 5.777972486175805
User: 59, Product: 127, Rating: 5.763390137322121
User: 642, Product: 173, Rating: 5.759447658367496
```