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1 AWS Data Lake Query Notebook

This notebook sets up the environment, configures Athena, and executes various queries to analyze the ingested data from the AWS S3 Data Lake.

```
[111]: import boto3
       import sagemaker
       import pandas as pd
       sess = sagemaker.Session()
       bucket = sess.default_bucket()
       role = sagemaker.get_execution_role()
       region = boto3.Session().region_name
       account_id = boto3.client("sts").get_caller_identity().get("Account")
       sm = boto3.Session().client(service_name="sagemaker", region_name=region)
[112]: s3_private_path_tsv = "s3://{}/amazon-reviews-pds/tsv".format(bucket)
       print(s3_private_path_tsv)
      s3://sagemaker-us-east-1-672518276407/amazon-reviews-pds/tsv
[113]: | !aws s3 cp "dataset.csv" $s3_private_path_tsv/
      upload: ./dataset.csv to s3://sagemaker-us-east-1-672518276407/amazon-reviews-
      pds/tsv/dataset.csv
[114]: print(s3_private_path_tsv)
      s3://sagemaker-us-east-1-672518276407/amazon-reviews-pds/tsv
[95]: from pyathena import connect
[119]: database_name = "dsoawsv5"
       s3_staging_dir = "s3://sagemaker-us-east-1-672518276407/amazon-reviews-pds/tsv"
       conn = connect(region_name=region, s3_staging_dir=s3_staging_dir)
       statement = "CREATE DATABASE IF NOT EXISTS {}".format(database_name)
       print(statement)
```

```
[120]: import pandas as pd
    pd.read_sql(statement, conn)

/tmp/ipykernel_3285/3803073958.py:3: UserWarning: pandas only supports
    SQLAlchemy connectable (engine/connection) or database string URI or sqlite3
    DBAPI2 connection. Other DBAPI2 objects are not tested. Please consider using
    SQLAlchemy.
    pd.read_sql(statement, conn)

[120]: Empty DataFrame
    Columns: []
    Index: []

[121]: statement = "SHOW DATABASES"

    df_show = pd.read_sql(statement, conn)
```

/tmp/ipykernel_3285/3999478089.py:3: UserWarning: pandas only supports SQLAlchemy connectable (engine/connection) or database string URI or sqlite3 DBAPI2 connection. Other DBAPI2 objects are not tested. Please consider using SQLAlchemy.

df_show = pd.read_sql(statement, conn)

```
[121]: database_name
0 default
1 dsoaws
2 dsoawsv2
3 dsoawsv3
4 dsoawsv4
```

df_show.head(10)

1.1 Fix the issue with dataset.csv file

```
[135]: import pandas as pd

# Path to the uploaded CSV file
csv_path = "./dataset.csv"

try:
    # Attempt to read the CSV file with common delimiters and encodings
    df_comma = pd.read_csv(csv_path)
    df_tab = pd.read_csv(csv_path, delimiter='\t')
    df_pipe = pd.read_csv(csv_path, delimiter='|')

# Display the first few rows and column names for each delimiter
```

```
print("Comma Delimiter:")
   print("Columns:", df_comma.columns.tolist())
   print(df_comma.head(), "\n")
   print("Tab Delimiter:")
   print("Columns:", df_tab.columns.tolist())
   print(df_tab.head(), "\n")
   print("Pipe Delimiter:")
   print("Columns:", df_pipe.columns.tolist())
   print(df_pipe.head(), "\n")
    # Check encoding issues
   with open(csv_path, 'rb') as file:
       raw_data = file.read(1000) # Read a sample of the file
       print("Sample Encoding Inspection:")
       print(raw_data[:200]) # Print the first 200 bytes
except Exception as e:
   print(f"Error occurred while inspecting CSV: {e}")
```

Error occurred while inspecting CSV: Error tokenizing data. C error: Expected 1 fields in line 715, saw 2

```
[136]: import pandas as pd

# Path to the uploaded CSV file
csv_path = "./dataset.csv"

try:
    # Open the file and inspect the problematic lines manually
    with open(csv_path, 'r', encoding='utf-8') as file:
        lines = file.readlines()

# Display lines around the problematic line (715)
    print("Lines around the problematic area:")
    for i in range(710, 720):
        print(f"Line {i+1}: {lines[i].strip()}")
except Exception as e:
    print(f"Error occurred while reading CSV lines: {e}")
```

Lines around the problematic area:
Line 711: 709,1qHiDbTJI7GB62W3BBFigx,A Great Big World;Futuristic,When the
Morning Comes,Hold Each Other (feat. Futuristic),47,216040,False,0.61,0.796,10,5.431,1,0.0656,0.0365,0.0,0.274,0.51,159.944,4,acoustic
Line 712: 710,5msvkAkQ8o2GhNxOu3YW5D,Hanare Gumi, , ,28,274133,False,0.63,
0.367,1,-11.054,1,0.0232,0.052,0.000202,0.103,0.547,98.995,4,acoustic
Line 713: 711,3YllcvA3PW1DUwjckCVjIw,Ray LaMontagne,MONOVISION,I Was Born To

```
Love You, 60, 251933, False, 0.659, 0.236, 9, -12.56, 1, 0.0304, 0.84, 6.32e-
05,0.206,0.216,78.088,4,acoustic
Line 714: 712,2cTaSKEc80ZdF6Tpg2QQsS,Aimyon,Heard that there's good pasta,Naked
Heart, 47, 296653, False, 0.559, 0.448, 8, -4.11, 1, 0.0275, 0.537, 1.85e-
05,0.0915,0.354,148.02,4,acoustic
Line 715: 713,1z2fSrYZqrO5tMqzULn90D,Tyrone Wells,"The ""Hits"" | Acoustic",Days
I Will Remember, 30, 201506, False, 0.804, 0.457, 6, -
7.845,1,0.197,0.6,0.0,0.15,0.552,98.035,4,acoustic
Line 716: 714,5QRt4Qovc0NoIVzwnlkU7z,Kaiak,River Love,Smell the Coffee,49,201861
False, 0.637, 0.46, 3, -9.07, 1, 0.0384, 0.855, 4.8e-05, 0.114, 0.345, 93.955, 4, acoustic
Line 717: 715,7Igc6JB0xPDcn5yRkkVGQX, Jason Mraz, Know., Unlonely, 51,231266, False, 0
.795, 0.667, 5, -4.831, 0, 0.0392, 0.379, 0.0, 0.221, 0.754, 97.987, 4, acoustic
Line 718: 716,3XfMyT4Xf5LegDhvbrFEjp,Boyce Avenue;Bea Miller, "Cover Sessions,
Vol. 4", See You Again, 50, 239802, False, 0.57, 0.413, 9, -
7.034,1,0.0345,0.535,0.0,0.076,0.278,152.032,4,acoustic
Line 719: 717,3IvEnBPF1egP9y1q3C2Fg4,Sakura Fujiwara,SUPERMARKET,
                                                                       ,28,247666
,False,0.565,0.566,6,-
7.258,1,0.0272,0.472,0.000426,0.148,0.551,92.017,4,acoustic
Line 720: 718,4XBkxG1jC5RBQJt65gma8v,Filip Nordin,Acustic Covers of Popular
Songs, More Than You Know - Acoustic Covers of Popular Songs, 51, 155510, False, 0.77
6,0.319,8,-7.496,1,0.031,0.791,0.0,0.159,0.501,98.016,4,acoustic
```

[140]: import pandas as pd # Path to the uploaded CSV file csv_path = "./dataset.csv" try: # Correctly load the CSV file by handling embedded quotes and commas df = pd.read_csv(csv_path, quotechar='"', escapechar='\\', delimiter=',') print("CSV loaded successfully with proper quote handling.") print(df.head(10)) except Exception as e: print(f"Error occurred while loading CSV: {e}")

CSV loaded successfully with proper quote handling.

artists	track_id	0	Unnamed:	
Gen Hoshino	5SuOikwiRyPMVoIQDJUgSV	0		0
Ben Woodward	4qPNDBW1i3p13qLCt0Ki3A	1		1
Ingrid Michaelson;ZAYN	1iJBSr7s7jYXzM8EGcbK5b	2		2
Kina Grannis	6lfxq3CG4xtTiEg7opyCyx	3		3
Chord Overstreet	5vjLSffimiIP26QG5WcN2K	4		4
Tyrone Wells	01MV019KtVTNfFiBU9I7dc	5		5
A Great Big World; Christina Aguilera	6Vc5wAMmXdKIAM7WUoEb7N	6		6
Jason Mraz	1EzrEOXmMH3G43AXT1y7pA	7		7
Jason Mraz;Colbie Caillat	0IktbUcnAGrvD03AWnz3Q8	8		8
Ross Copperman	7k9GuJYLp2AzqokyEdwEw2	9		9
	Gen Hoshino Ben Woodward Ingrid Michaelson; ZAYN Kina Grannis Chord Overstreet Tyrone Wells A Great Big World; Christina Aguilera Jason Mraz Jason Mraz; Colbie Caillat	5SuOikwiRyPMVoIQDJUgSV 4qPNDBW1i3p13qLCtOKi3A Ben Woodward 1iJBSr7s7jYXzM8EGcbK5b 6lfxq3CG4xtTiEg7opyCyx 5vjLSffimiIP26QG5WcN2K 01MVO19KtVTNfFiBU9I7dc 6Vc5wAMmXdKIAM7WUoEb7N 1EzrEOXmMH3G43AXT1y7pA 0IktbUcnAGrvD03AWnz3Q8 Gen Hoshino Ben Woodward Ingrid Michaelson; ZAYN Kina Grannis Chord Overstreet A Great Big World; Christina Aguilera	O 5SuOikwiRyPMVoIQDJUgSV Gen Hoshino 1 4qPNDBW1i3p13qLCt0Ki3A Ben Woodward 2 1iJBSr7s7jYXzM8EGcbK5b Ingrid Michaelson;ZAYN 3 61fxq3CG4xtTiEg7opyCyx Kina Grannis 4 5vjLSffimiIP26QG5WcN2K Chord Overstreet 5 01MV019KtVTNfFiBU9I7dc Tyrone Wells 6 6Vc5wAMmXdKIAM7WUoEb7N A Great Big World;Christina Aguilera 7 1EzrEOXmMH3G43AXT1y7pA Jason Mraz 8 01ktbUcnAGrvD03AWnz3Q8 Jason Mraz;Colbie Caillat	0 5SuOikwiRyPMVoIQDJUgSV Gen Hoshino 1 4qPNDBW1i3p13qLCt0Ki3A Ben Woodward 2 1iJBSr7s7jYXzM8EGcbK5b Ingrid Michaelson;ZAYN 3 6lfxq3CG4xtTiEg7opyCyx Kina Grannis 4 5vjLSffimiIP26QG5WcN2K Chord Overstreet 5 01MV0l9KtVTNfFiBU9I7dc Tyrone Wells 6 6Vc5wAMmXdKIAM7WUoEb7N A Great Big World;Christina Aguilera 7 1EzrEOXmMH3G43AXT1y7pA Jason Mraz 8 0IktbUcnAGrvD03AWnz3Q8 Jason Mraz;Colbie Caillat

```
album_name
0
                                                  Comedy
                                       Ghost (Acoustic)
1
2
                                         To Begin Again
3
   Crazy Rich Asians (Original Motion Picture Sou...
4
                                                 Hold On
5
                                   Days I Will Remember
                           Is There Anybody Out There?
6
7
                  We Sing. We Dance. We Steal Things.
8
                  We Sing. We Dance. We Steal Things.
9
                                                  Hunger
                    track_name
                                  popularity
                                               duration_ms
                                                             explicit
                                           73
0
                         Comedy
                                                     230666
                                                                 False
1
              Ghost - Acoustic
                                           55
                                                                 False
                                                     149610
2
                To Begin Again
                                           57
                                                                 False
                                                     210826
3
   Can't Help Falling In Love
                                           71
                                                     201933
                                                                 False
4
                        Hold On
                                           82
                                                     198853
                                                                False
5
         Days I Will Remember
                                           58
                                                                False
                                                     214240
6
                 Say Something
                                          74
                                                     229400
                                                                False
7
                      I'm Yours
                                                                False
                                           80
                                                     242946
8
                                           74
                                                                False
                          Lucky
                                                     189613
9
                         Hunger
                                           56
                                                     205594
                                                                False
   danceability
                 energy
                              loudness
                                         mode
                                                speechiness
                                                              acousticness
0
           0.676
                  0.4610
                                 -6.746
                                             0
                                                      0.1430
                                                                     0.0322
1
           0.420
                  0.1660
                               -17.235
                                             1
                                                      0.0763
                                                                     0.9240
2
                                             1
           0.438
                  0.3590
                                 -9.734
                                                      0.0557
                                                                     0.2100
3
           0.266
                  0.0596
                               -18.515
                                                                     0.9050
                                                      0.0363
4
           0.618
                  0.4430
                                 -9.681
                                             1
                                                      0.0526
                                                                     0.4690
5
                                                                     0.2890
           0.688
                  0.4810
                                 -8.807
                                             1
                                                      0.1050
6
           0.407
                  0.1470
                                 -8.822
                                             1
                                                      0.0355
                                                                     0.8570
7
                  0.4440
           0.703
                                 -9.331
                                             1
                                                      0.0417
                                                                     0.5590
                                 -8.700
8
           0.625
                  0.4140
                                             1
                                                      0.0369
                                                                     0.2940
9
           0.442
                  0.6320
                                 -6.770
                                             1
                                                      0.0295
                                                                     0.4260
   instrumentalness
                       liveness
                                  valence
                                              tempo
                                                      time_signature
                                                                       track genre
0
            0.00001
                         0.3580
                                   0.7150
                                             87.917
                                                                    4
                                                                           acoustic
1
            0.000006
                         0.1010
                                   0.2670
                                             77.489
                                                                    4
                                                                           acoustic
2
            0.00000
                         0.1170
                                   0.1200
                                             76.332
                                                                    4
                                                                           acoustic
3
            0.000071
                         0.1320
                                   0.1430
                                            181.740
                                                                    3
                                                                           acoustic
4
                                                                    4
            0.00000
                         0.0829
                                   0.1670
                                            119.949
                                                                           acoustic
5
            0.00000
                         0.1890
                                   0.6660
                                             98.017
                                                                    4
                                                                           acoustic
6
                                                                    3
            0.000003
                         0.0913
                                   0.0765
                                            141.284
                                                                           acoustic
7
                                                                    4
            0.00000
                         0.0973
                                   0.7120
                                            150.960
                                                                           acoustic
8
            0.00000
                         0.1510
                                   0.6690
                                            130.088
                                                                    4
                                                                           acoustic
9
            0.004190
                         0.0735
                                   0.1960
                                             78.899
                                                                    4
                                                                           acoustic
```

```
[10 rows x 21 columns]
```

1.2 Use the cleaned csv file for creating datalake on AWS Athena

```
[142]: csv_output_path = "./cleaned_dataset.csv"
       df.to_csv(csv_output_path, index=False, quotechar='"', escapechar='\\')
       print(f"DataFrame saved to {csv_output_path}")
      DataFrame saved to ./cleaned_dataset.csv
[147]: s3_staging_dir = "s3://sagemaker-us-east-1-672518276407/amazon-reviews-pds-1/
        ⇔tsv"
       !aws s3 cp "cleaned_dataset.csv" $s3_staging_dir/
      upload: ./cleaned_dataset.csv to s3://sagemaker-us-east-1-672518276407/amazon-
      reviews-pds-1/tsv/cleaned_dataset.csv
[144]: print(s3_private_path_tsv)
      s3://sagemaker-us-east-1-672518276407/amazon-reviews-pds/tsv
[158]: database name = "dsoaws new"
       conn = connect(region_name=region, s3_staging_dir=s3_staging_dir)
       statement = "CREATE DATABASE IF NOT EXISTS {}".format(database name)
       print(statement)
       pd.read_sql(statement, conn)
      CREATE DATABASE IF NOT EXISTS dsoaws_new
      /tmp/ipykernel_3285/750785309.py:5: UserWarning: pandas only supports SQLAlchemy
      connectable (engine/connection) or database string URI or sqlite3 DBAPI2
      connection. Other DBAPI2 objects are not tested. Please consider using
      SQLAlchemy.
        pd.read_sql(statement, conn)
[158]: Empty DataFrame
       Columns: []
       Index: []
[159]: statement = "SHOW DATABASES"
       df_show = pd.read_sql(statement, conn)
       df show.head(20)
      /tmp/ipykernel_3285/1371328689.py:2: UserWarning: pandas only supports
      SQLAlchemy connectable (engine/connection) or database string URI or sqlite3
      DBAPI2 connection. Other DBAPI2 objects are not tested. Please consider using
      SQLAlchemy.
```

df show = pd.read sql(statement, conn)

```
[159]:
          database_name
                default
       1
                 dsoaws
       2
              dsoaws001
       3
                dsoaws2
       4
                dsoaws3
       5
                dsoaws4
       6
             dsoaws_001
       7
             dsoaws_new
       8
               dsoawsv2
       9
               dsoawsv3
       10
               dsoawsv4
               dsoawsv6
       11
       12
               dsoawsv8
[73]: import boto3
       import pandas as pd
       from io import StringIO
       # S3 bucket and file path
       bucket_name = "sagemaker-us-east-1-672518276407"
       file_key = "amazon-reviews-pds/tsv/dataset.csv"
       # Create an S3 client
       s3 = boto3.client("s3")
       try:
           # Fetch the CSV file from S3
           response = s3.get_object(Bucket=bucket_name, Key=file_key)
           # Read the content of the file
           content = response["Body"].read().decode("utf-8")
           # Load the content into a DataFrame
           data = pd.read_csv(StringIO(content))
           # Display the first few rows
           print(data.head())
       except Exception as e:
           print(f"Error occurred: {e}")
```

```
Unnamed: 0
                             track_id
                                                      artists \
            O 5SuOikwiRyPMVoIQDJUgSV
0
                                                  Gen Hoshino
1
            1 4qPNDBW1i3p13qLCt0Ki3A
                                                 Ben Woodward
            2 1iJBSr7s7jYXzM8EGcbK5b
2
                                      Ingrid Michaelson; ZAYN
3
            3 6lfxq3CG4xtTiEg7opyCyx
                                                 Kina Grannis
4
            4 5vjLSffimiIP26QG5WcN2K
                                             Chord Overstreet
```

```
0
                                                    Comedy
     1
                                          Ghost (Acoustic)
     2
                                            To Begin Again
     3
        Crazy Rich Asians (Original Motion Picture Sou...
     4
                                                   Hold On
                        track name popularity
                                                duration ms
                                                              explicit \
                                                      230666
                                                                 False
     0
                            Comedy
                                             73
     1
                  Ghost - Acoustic
                                             55
                                                      149610
                                                                 False
     2
                    To Begin Again
                                             57
                                                                 False
                                                      210826
        Can't Help Falling In Love
                                             71
                                                                 False
     3
                                                      201933
     4
                           Hold On
                                             82
                                                                 False
                                                      198853
        danceability energy
                                 loudness mode
                                                  speechiness acousticness
     0
               0.676 0.4610
                                    -6.746
                                               0
                                                       0.1430
                                                                     0.0322
               0.420 0.1660 ...
                                  -17.235
                                                       0.0763
                                                                     0.9240
     1
                                               1
     2
               0.438 0.3590 ...
                                   -9.734
                                               1
                                                       0.0557
                                                                     0.2100
     3
               0.266 0.0596 ...
                                  -18.515
                                               1
                                                       0.0363
                                                                     0.9050
                                    -9.681
                                                       0.0526
     4
               0.618 0.4430 ...
                                               1
                                                                     0.4690
        instrumentalness liveness valence
                                               tempo
                                                       time_signature track_genre
     0
                0.000001
                            0.3580
                                       0.715
                                               87.917
                                                                          acoustic
     1
                0.000006
                            0.1010
                                       0.267
                                               77.489
                                                                    4
                                                                          acoustic
     2
                0.000000
                            0.1170
                                       0.120
                                              76.332
                                                                    4
                                                                          acoustic
     3
                0.000071
                            0.1320
                                       0.143 181.740
                                                                    3
                                                                          acoustic
     4
                0.000000
                            0.0829
                                       0.167
                                              119.949
                                                                    4
                                                                          acoustic
     [5 rows x 21 columns]
[74]: # Filter songs with popularity greater than or equal to 99
      df high popularity = data[data['popularity'] >= 99]
      df_high_popularity = df_high_popularity[['artists', 'track_name', 'popularity']]
      print("Songs with popularity 99:")
      print(df_high_popularity)
     Songs with popularity
                         artists
                                                              track_name popularity
     20001 Sam Smith; Kim Petras
                                               Unholy (feat. Kim Petras)
                                                                                  100
     51664
                Bizarrap; Quevedo : Bzrp Music Sessions, Vol. 52
                                                                                  99
     81051 Sam Smith; Kim Petras
                                               Unholy (feat. Kim Petras)
                                                                                  100
[75]: # Group by artist and calculate the average popularity
      df_avg_popularity = data.groupby('artists')['popularity'].mean().reset_index()
      # Filter artists with an average popularity of 92
```

album_name \

```
df_avg_popularity_92 = df_avg_popularity[df_avg_popularity['popularity'] == 92]
       print("Artists with an average popularity of 92:")
       print(df_avg_popularity_92)
      Artists with an average popularity of 92:
                       artists popularity
      11491
                  Harry Styles
                                      92.0
      22845 Rema; Selena Gomez
                                       92.0
[76]: # Group by genre and calculate the average energy
       df_avg_energy = data.groupby('track_genre')['energy'].mean().reset_index()
       # Get the top 10 genres by average energy
       df_top10_genres = df_avg_energy.sort_values(by='energy', ascending=False).
        \hookrightarrowhead(10)
       print("Top 10 genres with the highest average energy:")
       print(df_top10_genres)
      Top 10 genres with the highest average energy:
            track_genre
                           energy
      22
            death-metal 0.931470
              grindcore 0.924201
      42
      72
              metalcore 0.914485
      46
                  happy 0.910971
      49
              hardstyle 0.901246
      27 drum-and-bass 0.876635
            black-metal 0.874897
      6
      50
            heavy-metal 0.874003
      78
                  party 0.871237
      61
                 j-idol 0.868677
[192]: # Filter tracks by the artist 'Bad Bunny'
       bad_bunny_count = data[data['artists'].str.contains('Bad Bunny', na=False)].
        ⇒shape[0]
       print(f"Number of tracks featuring Bad Bunny: {bad_bunny_count}")
      Number of tracks featuring Bad Bunny: 416
[78]: # Group by genre and find the maximum popularity within each genre
       df_genre_popularity = data.groupby('track_genre')['popularity'].max().
        →reset index()
       # Sort by popularity and get the top 10 genres
       df_top10_popular_genres = df_genre_popularity.sort_values(by='popularity',__
        ⇒ascending=False).head(10)
       print("Top 10 genres by their most popular track:")
       print(df_top10_popular_genres)
```

```
Top 10 genres by their most popular track:
        track_genre popularity
     80
                            100
                pop
     20
              dance
                            100
                             99
     51
            hip-hop
             latino
                             98
     68
     89
          reggaeton
                             98
     30
                edm
                             98
     67
              latin
                             98
     88
             reggae
                             98
     79
              piano
                             96
     90
               rock
                             96
[79]: statement = "SHOW TABLES in {}".format('dsoaws')
      df_show = pd.read_sql(statement, conn)
      df_show.head(5)
     /tmp/ipykernel 3285/2205842015.py:3: UserWarning: pandas only supports
     SQLAlchemy connectable (engine/connection) or database string URI or sqlite3
     DBAPI2 connection. Other DBAPI2 objects are not tested. Please consider using
     SQLAlchemy.
       df_show = pd.read_sql(statement, conn)
[79]:
                   tab name
      0 amazon_reviews_tsv
[80]: # SageMaker session setup
      sess = sagemaker.Session()
      bucket = sess.default_bucket()
      role = sagemaker.get_execution_role()
      region = boto3.Session().region_name
      # Athena connection setup
      database_name = "dsoaws"
      table_name_tsv = "amazon_reviews_tsv"
      conn = connect(region_name=region, s3_staging_dir=s3_staging_dir)
      print("Setup completed.")
     Setup completed.
[81]: # Athena connection setup
      conn = connect(region_name=region, s3_staging_dir=s3_staging_dir)
      # Improved execute_query function
      def execute_query(query, description="Query Result"):
```

```
Executes a Pandas query or calculation and prints the result.
Parameters:
    query (pd.DataFrame or int): The result of the query.
    description (str): A brief description of the query result.
try:
    print(f"\n{description}")
    print("=" * len(description))
    # Check if the query returned a DataFrame
    if isinstance(query, pd.DataFrame):
        if not query.empty:
            display(query.head(100)) # Display the first 100 rows
            print(f"Total Records: {len(query)}")
        else:
            print("No matching records found.")
    # Check if the query returned a single value (like a count)
    elif isinstance(query, int):
        print(f"Result: {query}")
    else:
        print("Unexpected query result type.")
except Exception as e:
    print(f"Error executing query: {e}")
```

1.3 List artist, track_name, and popularity for songs with popularity 99

```
[82]: import pandasql as psql

# SQL-like query on the DataFrame
query = """
SELECT artists, track_name, popularity
FROM data
WHERE popularity >= 99
"""
result = psql.sqldf(query, locals())
execute_query(result, "Songs with Popularity 99 (SQL-like)")
```

```
Songs with Popularity 99 (SQL-like)
```

```
artists track_name popularity

O Sam Smith; Kim Petras Unholy (feat. Kim Petras) 100

Bizarrap; Quevedo Quevedo: Bzrp Music Sessions, Vol. 52 99

2 Sam Smith; Kim Petras Unholy (feat. Kim Petras) 100
```

Total Records: 3

1.4 List artists with an average popularity of 92

```
[83]: query = """
    SELECT artists, AVG(popularity) AS avg_popularity
    FROM data
    GROUP BY artists
    HAVING AVG(popularity) = 92
    """
    result = psql.sqldf(query, locals())
    execute_query(result, "List artists with an average popularity of 92")
```

List artists with an average popularity of 92

artists avg_popularity
0 Harry Styles 92.0
1 Rema; Selena Gomez 92.0

Total Records: 2

1.5 List the Top 10 genres with the highest average energy

List the Top 10 genres with the highest average energy

```
track_genre avg_energy
0
    death-metal
                  0.931470
      grindcore 0.924201
1
      metalcore 0.914485
2
          happy 0.910971
3
4
      hardstyle 0.901246
5
 drum-and-bass 0.876635
6
    black-metal 0.874897
7
    heavy-metal
                 0.874003
8
          party
                  0.871237
         j-idol
                  0.868677
```

Total Records: 10

1.6 How many tracks is Bad Bunny on?

```
[193]: query = """
    SELECT COUNT(*)
    FROM data
    WHERE artists like '%Bad Bunny%'
    """
    result = psql.sqldf(query, locals())
    execute_query(result, "How many tracks is Bad Bunny on?")
```

```
How many tracks is Bad Bunny on?

COUNT(*)

416

Total Records: 1
```

1.7 Show the top 10 genres in terms of popularity, sorted by their most popular track

Show the top 10 genres in terms of popularity, sorted by their most popular track

track_genre max_popularity 0 100 pop 1 dance 100 2 99 hip-hop 3 reggaeton 98 4 98 reggae 5 latino 98 6 98 latin 7 98 edm8 rock 96

```
9    piano     96
Total Records: 10

[87]: sess = sagemaker.Session()
    bucket = sess.default_bucket()
    role = sagemaker.get_execution_role()
    region = boto3.Session().region_name

sm = boto3.Session().client(service_name="sagemaker", region_name=region)

[178]: import awswrangler as wr

# Database name
database_name = "dsoaws_parquet"
```

```
[178]: import awswrangler as wr

# Database name
database_name = "dsoaws_parquet"

# Create the Glue database if it doesn't exist
try:
    wr.catalog.create_database(name=database_name)
    print(f"Database '{database_name}' created successfully or already exists.")
except Exception as e:
    print(f"Error creating database: {e}")
```

Database 'dsoaws_parquet' created successfully or already exists.

```
[179]: import pandas as pd
       import awswrangler as wr
       # Load the DataFrame from the CSV file
       csv_path = "./cleaned_dataset.csv"
       df = pd.read_csv(csv_path)
       # S3 bucket and file path for Parquet
       bucket_name = "sagemaker-us-east-1-672518276407"
       parquet_path = f"s3://{bucket_name}/amazon-reviews-pds-1/parquet/
        \negcleaned_dataset.parquet"
       # Save DataFrame as Parquet to S3
       try:
           wr.s3.to_parquet(
               df=df,
               path=parquet_path,
               dataset=True,
               mode="overwrite",
               database="dsoaws_parquet",
               table="amazon_reviews_parquet"
           print(f"Data uploaded successfully to {parquet_path} as Parquet.")
```

```
except Exception as e:
    print(f"Error uploading DataFrame to S3: {e}")
```

Data uploaded successfully to s3://sagemaker-us-east-1-672518276407/amazon-reviews-pds-1/parquet/cleaned_dataset.parquet as Parquet.

```
[182]: # List all tables in the specified database
import awswrangler as wr

# Database name
database_name = "dsoaws_parquet"

# List all tables in the specified database
try:
    tables = list(wr.catalog.get_tables(database=database_name))
    print("Tables in the database:")
    for table in tables:
        print(f"- {table['Name']}")
except Exception as e:
    print(f"Error retrieving tables: {e}")
```

Tables in the database:

- amazon_reviews_parquet

Sample data from the Athena table:

```
unnamed 0
                            track id
                                                     artists \
          0 5SuOikwiRyPMVoIQDJUgSV
                                                 Gen Hoshino
0
          1 4qPNDBW1i3p13qLCt0Ki3A
1
                                                Ben Woodward
2
           2 1iJBSr7s7jYXzM8EGcbK5b Ingrid Michaelson;ZAYN
3
           3 6lfxq3CG4xtTiEg7opyCyx
                                                Kina Grannis
          4 5vjLSffimiIP26QG5WcN2K
                                            Chord Overstreet
                                          album_name \
                                              Comedy
0
1
                                    Ghost (Acoustic)
2
                                      To Begin Again
3
  Crazy Rich Asians (Original Motion Picture Sou...
```

Hold On

```
track_name popularity
                                                          explicit \
                                            duration_ms
0
                                                  230666
                                                             False
                        Comedy
                                        73
1
             Ghost - Acoustic
                                        55
                                                  149610
                                                             False
2
               To Begin Again
                                        57
                                                             False
                                                  210826
  Can't Help Falling In Love
                                                             False
3
                                        71
                                                  201933
4
                      Hold On
                                        82
                                                             False
                                                  198853
   danceability energy
                            loudness
                                       mode
                                              speechiness acousticness
0
          0.676 0.4610
                               -6.746
                                          0
                                                   0.1430
                                                                 0.0322
          0.420 0.1660
                              -17.235
                                                   0.0763
                                                                 0.9240
1
                                          1
2
          0.438 0.3590 ...
                               -9.734
                                                   0.0557
                                                                 0.2100
                                          1
3
          0.266 0.0596 ...
                              -18.515
                                          1
                                                   0.0363
                                                                 0.9050
4
                               -9.681
          0.618 0.4430 ...
                                          1
                                                   0.0526
                                                                 0.4690
   instrumentalness liveness valence
                                           tempo
                                                  time_signature
                                                                   track_genre
0
           0.000001
                       0.3580
                                  0.715
                                          87.917
                                                                       acoustic
1
           0.000006
                       0.1010
                                  0.267
                                          77.489
                                                                4
                                                                       acoustic
2
                                          76.332
           0.000000
                       0.1170
                                  0.120
                                                                4
                                                                      acoustic
                                  0.143 181.740
3
           0.000071
                       0.1320
                                                                3
                                                                       acoustic
           0.000000
4
                       0.0829
                                  0.167 119.949
                                                                4
                                                                       acoustic
```

[5 rows x 21 columns]

1.8 List artist, track_name, and popularity for songs that have a popularity greater than or equal to 99

Songs with popularity >= 99:

```
artists track_name popularity

0 Sam Smith; Kim Petras Unholy (feat. Kim Petras) 100

1 Bizarrap; Quevedo Quevedo: Bzrp Music Sessions, Vol. 52 99

2 Sam Smith; Kim Petras Unholy (feat. Kim Petras) 100
```

1.9 List artists with an average popularity of 92

```
[186]: query2 = """
SELECT artists, AVG(popularity) AS avg_popularity
FROM dsoaws_parquet.amazon_reviews_parquet
GROUP BY artists
```

```
HAVING AVG(popularity) = 92
"""

df2 = wr.athena.read_sql_query(query2, database="dsoaws_parquet")
print("Artists with an average popularity of 92:")
print(df2)
```

1.10 List the Top 10 genres with the highest average energy

```
[187]: query3 = """
    SELECT track_genre, AVG(energy) AS avg_energy
    FROM dsoaws_parquet.amazon_reviews_parquet
    GROUP BY track_genre
    ORDER BY avg_energy DESC
    LIMIT 10
    """

    df3 = wr.athena.read_sql_query(query3, database="dsoaws_parquet")
    print("Top 10 genres with highest average energy:")
    print(df3)
```

```
Top 10 genres with highest average energy:
```

```
track_genre avg_energy
    death-metal
0
                  0.931470
1
      grindcore 0.924201
2
      metalcore 0.914485
          happy 0.910971
3
      hardstyle 0.901246
4
  drum-and-bass 0.876635
5
6
    black-metal 0.874897
7
    heavy-metal 0.874003
8
                  0.871237
          party
9
                  0.868677
         j-idol
```

1.11 How many tracks is Bad Bunny on?

```
[194]: query4 = """
    SELECT COUNT(*) AS track_count
    FROM dsoaws_parquet.amazon_reviews_parquet
    WHERE artists LIKE '%Bad Bunny%'
    """
    df4 = wr.athena.read_sql_query(query4, database="dsoaws_parquet")
    print("Number of tracks featuring Bad Bunny:")
    print(df4)
```

```
Number of tracks featuring Bad Bunny:
track_count
0 416
```

1.12 Show the top 10 genres in terms of popularity, sorted by their most popular track

Top 10 genres by most popular track:

```
track_genre max_popularity
0
                            100
           pop
                            100
1
        dance
2
      hip-hop
                             99
3
                             98
       reggae
4
    reggaeton
                             98
5
           edm
                             98
6
        latin
                             98
7
       latino
                             98
8
        piano
                             96
9
         rock
                             96
```

[]: