

Amazon Web Services Athena Workshop

Lab3.Exploring Data Lake with Athena (Advanced)

August 2020

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Introduction

This additional lab on Amazon Athena covers best practices on partitioning, columnar formats, and compression that can improve query performance and can get significant cost savings.

Prerequisites:

Each student should receive individual AWS Access URL/credentials.

Getting Started

In this lab, you will complete the following tasks: You would use Athena Console to perform all these tasks.

- 1. Logon to console and view public dataset in S3:
- 2. Create Athena Tables
- 3. Verify Tables in Glue Catalog
- 4. Query Data with Amazon Athena
- 5. Create View
- 6. Run CTAS to create new partitioned table
- 7. Run CTAS to explore Bucketing
- 8. <u>Use Athena Workgroups</u>

Logon to console and view public dataset in S3:

In this lab, we will be using data from - <u>Amazon Customer Reviews</u>. This dataset provides both TSV (tab separated values) and Parquet versions of over 130 million customer reviews since 1995.

1. To view source dataset in S₃, access below URL https://console.aws.amazon.com/s₃/home?region=us-east-1&bucket=amazon-reviews-pds

tsv folder has multiple files compressed using gzip. Also notice that file size varies from 12 MB to 2.6 GB.

Parquet folder has sub-folders on product category and going down one level, you would notice that files are compressed using snappy. File size is more uniform.

| Format | Compression | Total Size |
|---------------------------|-------------|------------|
| Tsv, row storage | gzip | 32.2 GB |
| Parquet, columnar storage | snappy | 47.4 GB |

Gzip and Snappy are non-splittable file formats with gzip having higher compression ratio than snappy. But Snappy is significantly faster in compression and decompression speeds.

Compressing your data can speed up your queries significantly, as long as the files are either of an optimal size (see the next section), or the files are splittable. The smaller data sizes reduce network traffic from Amazon S₃ to Athena.

Splittable files allow the execution engine in Athena to split the reading of a file by multiple readers to increase parallelism. If you have a single unsplittable file, then only a single reader can read the file while all other readers sit idle. Not all compression algorithms are splittable. The following table lists common compression formats and their attributes.

| Algorithm | Splittable? | | Compress + Decompress speed |
|-------------------|-------------|-----------|-----------------------------------|
| Gzip (DEFLATE) | No | High | Medium |
| bzip2 | Yes | Very high | Slow |
| LZO | No | Low | Fast |
| Snappy | No | Low | Very fast |

Queries run more efficiently when reading data can be parallelized and when blocks of data can be read sequentially. Ensuring that your file formats are splittable helps with parallelism regardless of how large your files may be.

However, if your files are too small (generally less than 128 MB), the execution engine might be spending additional time with the overhead of opening Amazon S₃ files, listing directories, getting object metadata, setting up data transfer, reading file headers, reading compression dictionaries, and so on. On the other hand, if your file is not splittable and the files are too large, the query processing waits until a single reader has completed reading the entire file. That can reduce parallelism.

Create Athena Tables:

When you create a database and table in Athena, you describe the schema and location of the data, making the data in the table ready for querying.

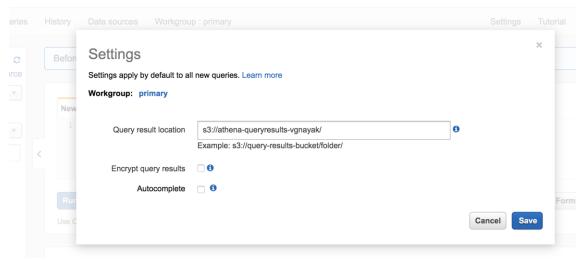
In this exercise, we will create table using Athena Query editor and then explore an alternate option of automatically creating tables using Glue Crawler.

Once the table is created, it is made available in the centralized Glue Catalog. Glue catalog is shared by services like Athena, Redshift Spectrum, EMR, Glue ETL and Hive compatible stores.

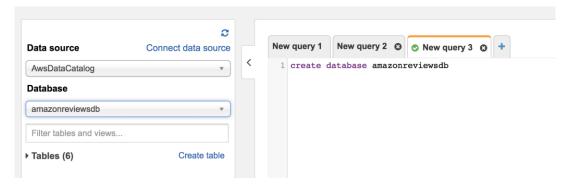
Create Athena Tables Using Query Editor:

1. Before starting Athena, create s3 bucket to store Athena results. Name : athena-queryresults-<yourname>

- 2. In AWS services, search & select **Athena**.
- 3. When you get to query editor, go to settings and update Query result location with bucket name created in Step 1.



- 4. In query editor, enter create database amazonreviewsdb
- 5. When you run above command, it should create new database.
- 6. Select your database *amazonreviewsdb* from Database drop down on the left



When you run a CREATE TABLE query in Athena, you register your table with the AWS Glue Data Catalog. Let's start creating tables.

7. We will create our first table by running following statement that point to TSV files. In create statement LOCATION clause, you'll notice that we are using TSV files.

```
CREATE EXTERNAL TABLE amazon_reviews_tsv (
marketplace string,
customer_id string,
review_id string,
product_id string,
product_parent string,
product_title string,
```

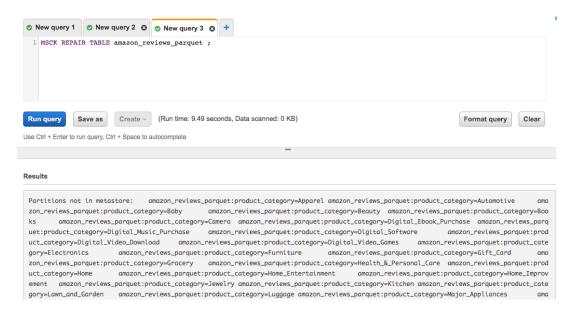
```
product_category string,
 star_rating int,
 helpful_votes int,
 total_votes int,
 vine string,
 verified_purchase string,
 review headline string,
 review_body string,
 review date date,
 year int)
ROW FORMAT DELIMITED
 FIELDS TERMINATED BY '\t'
 ESCAPED BY '\\'
 LINES TERMINATED BY '\n'
LOCATION
 's3://amazon-reviews-pds/tsv/'
TBLPROPERTIES ("skip.header.line.count"="1");
```

8. Create the second table in the database, this time pointing to parquet files

```
CREATE EXTERNAL TABLE `amazon_reviews_parquet`(
 `marketplace` string,
 `customer_id` string,
 `review_id` string,
 `product_id` string,
 `product_parent` string,
 `product_title` string,
 `star_rating` int,
 `helpful_votes` int,
 `total votes` int,
 `vine` string,
 `verified_purchase` string,
 `review_headline` string,
 `review_body` string,
 `review_date` bigint,
 `year` int)
PARTITIONED BY ( `product_category` string )
STORED AS PARQUET
LOCATION 's3://amazon-reviews-pds/parquet';
```

9. In the previous step, although query is successful you could notice a message stating that partitions need to be loaded to this table. To load the partitions, run below statement

```
MSCK REPAIR TABLE amazon_reviews_parquet;
```

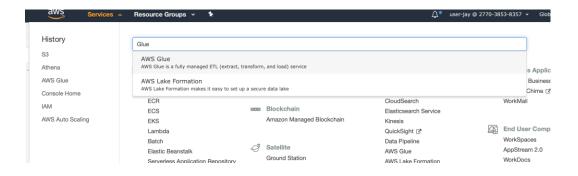


Athena leverages Hive for partitioning data. You can partition your data by any key. Also, if you query a partitioned table and specify the partition in the WHERE clause, Athena scans the data only from that partition thus leading to less cost per query.

Create Athena Tables Using Glue Crawler:

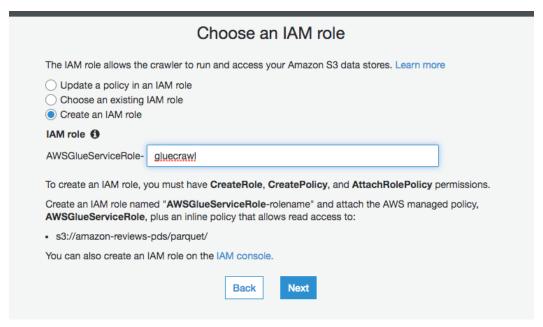
Now, let's leverage crawler in AWS Glue to retrieve schema information automatically. The crawlers go through your data, and inspect portions of it to determine the schema. In addition, the crawler can detect and register partitions.

1. In the AWS services console, search for Glue and select AWS Glue

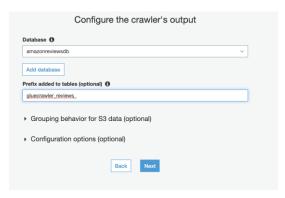


- 2. In the left pane, click "crawlers" and then select "Add Crawler".
- 3. Enter the crawler name "amazon_reviews_parquet_gluecrawler" and select Next.
- 4. Select "Data Stores" and select Next.
- 5. Enter the S₃ location as "s3://amazon-reviews-pds/parquet/" and select Next

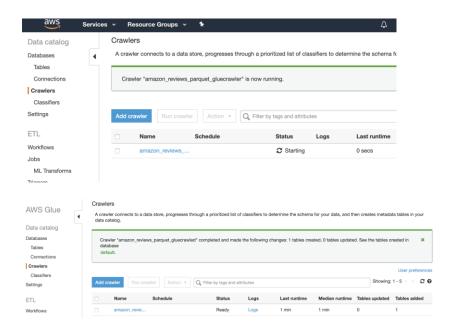
- 6. Select Next on "Add Another data store" with default option selected as No
- 7. Now create a new IAM role for the crawler and hit Next



- 8. Choose a schedule for crawler. For this exercise just select "Run on Demand" and hit next.
- 9. Select database "amazonreviewsdb", enter table name prefix "gluecrawler_reviews_". Click finish to Add the crawler.

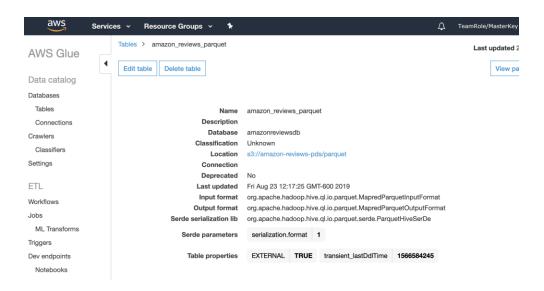


10. Find the new crawler listed with status "Ready". Select the checkbox next to the crawler and select "RunCrawler". The crawler would show initial "starting" status and it would turn to "stopping" and then turn to "Ready". This should create table in catalog. Approximate run time ~ 3 mins.

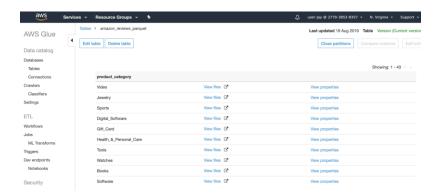


Verify Tables in Glue Catalog

- 1. If you are in AWS Glue, then navigate to tables and you should see table created by crawler with prefix gluecrawler_reviews_
- 2. Also, you should see tables created in earlier step from Athena query editor
- 3. Select table "amazon_reviews_parquet" created in first step and view the table definitions. Notice the location where the table data is referred.



At the right top, click on "view partitions", to view the various partition values available for product_category.



Query Data with Amazon Athena

Let's start running queries

- 1. In Athena Query Editor, select the database "amazonreviewsdb"
- 2. Now let's run a simple guery on tsv table to preview data

```
SELECT * FROM "amazon_reviews_tsv"
WHERE marketplace = 'US'
limit 10;
```

Run time: 2.66 seconds, Data scanned: 236.87 KB

 Now let's try a query with aggregations on tsv and parquet tables. This is common pattern in analytical/reporting applications.
 First, let's run this on tsv to capture run time and data scanned

```
SELECT product_id, product_title, count(*) as num_reviews, avg(star_rating) as avg_stars
FROM amazon_reviews_tsv where product_category='Toys'
GROUP BY 1, 2
ORDER BY 3 DESC
limit 100;
```

(Run time: 1 minute 26 seconds, Data scanned: 32.23 GB)

4. Run the same query now against Parquet partitioned table

```
SELECT product_id, product_title, count(*) as num_reviews, avg(star_rating) as avg_stars
FROM amazon_reviews_parquet where product_category='Toys'
GROUP BY 1, 2
ORDER BY 3 DESC
limit 100;
```

(Run time: 4.53 seconds, Data scanned: 215.04 MB)

Significant improvement in performance and data scanned can be attributed to parquet columnar storage format and effective partitions on predicates. When Athena executes a query on a partitioned table, it first checks to see if any partitioned columns were used in the WHERE clause of the query. If partitioned columns were used, Athena requests the AWS Glue Data Catalog to return the partition specification matching the specified partition columns. The partition specification includes the LOCATION property that tells Athena which Amazon S₃ prefix to use when reading data. In this case, *only* data stored in this prefix is scanned.

Create view

- A view in Amazon Athena is a logical, not a physical table. The query that defines a view runs each time the view is referenced in a query. Create views when you want to hide underlying complexity and minimize maintenance problems if underlying table/column names change.
- Open a new Athena Query editor tab and run the below query to retrieve TopReviewedStarRated products

```
Create view TopReviewedStarRatedProducts_v as
SELECT product_category, product_id, product_title, count(*) TotalReviews FROM
amazon_reviews_parquet
WHERE star_rating=5
group by product_category, product_id, product_title
order by TotalReviews desc
```

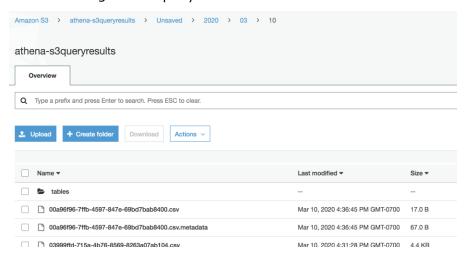
3. You can now run query on this view to extract anytime the Top N reviewed products among those high rated over the years.

```
select * from topreviewedstarratedproducts_v limit 10
Run time: 6.06 seconds, Data scanned: 6.39 GB
```

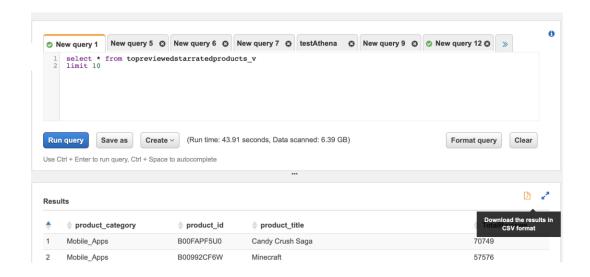
Query Results and History

- 1. For every query run, Athena automatically saves query results in S₃ at query result location. Files are saved to the query result location in Amazon S₃ based on the name of the query, the query ID, and the date that the query ran. Files for each query are named using the QueryID, which is a unique identifier that Athena assigns to each query when it runs.
- 2. To identify QueryID Choose history from Query Editor and then click on state "Failed" or "Succeeded"

3. Navigate to S₃ location *athena-queryresults-*<*yourname*> and identify the query results saved in S₃ for last query execution.



4. You can also optionally download the results from Query editor via the download link



Run CTAS to create new partitioned parquet table

In earlier step, we started by accessing parquet formatted dataset directly in Athena. But there will be instances where files are staged in CSV/TSV/TXT and we need to run a step of converting them to parquet/orc format.

Objective of this section is to show you an approach to accomplish this conversion.

We would use CTAS (Create Table As Select) command to create a new table that first queries TSV data and writes the results in Parquet format in S₃. Also partitions the dataset by "marketplace" and "year"

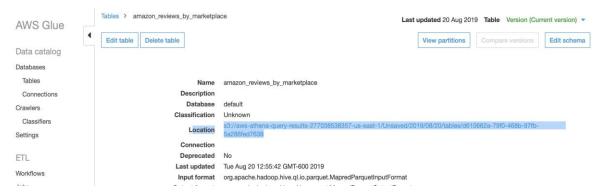
1. Run below query in Athena. This query takes ~5 mins to complete.

```
CREATE TABLE amazon_reviews_by_marketplace
WITH (
format='PARQUET',
partitioned_by = ARRAY['marketplace', 'year']
) AS
SELECT customer_id, review_id, product_id, product_parent, product_title,
product_category, star_rating, helpful_votes, total_votes, verified_purchase,
review_headline, review_body, review_date,
marketplace,
year(review_date) as year
FROM amazon_reviews_tsv
WHERE "$path" LIKE '%tsv.gz'
```

<u>Note:</u> CTAS command writes results to S₃ under S₃://athena-queryresults-<yourname>/location. You can specify your own S₃ location by specifying external location = 's3://...', in the CTAS command.

Partitioning strategy depends on the query access patterns. In this case, we want to analyze the data by marketplace and year. The following queries will show how proper partitioning can help reduce data scan and improve performance for this access pattern.

- 2. Once table is created, go to glue catalog and view the newly created table. Select the new table "amazon_reviews_by_marketplace".
- 3. Find the S₃ location where the table has created new partitioned data.



4. Click the S₃ location to view the new partitioned structure. The first level partition displays "marketplace" and selecting one of the folders display its inner partition by "year" and the contents within.

5. Now go back to Athena Query Editor and try the following queries over the newly created table.

```
SELECT product_id, COUNT(*) FROM amazon_reviews_by_marketplace
WHERE marketplace='US' AND year = 2013
GROUP BY 1 ORDER BY 2 DESC LIMIT 10
```

[parquet + partitioned by marketplace] - > (Run time: 5.27 seconds, Data scanned: 145.66 MB)

Vs

```
SELECT product_id, COUNT(*) FROM amazon_reviews_parquet
WHERE marketplace='US' AND year = 2013
GROUP BY 1 ORDER BY 2 DESC LIMIT 10
```

[parquet] - > (Run time: 5.88 seconds, Data scanned: 882 MB). Though performance is comparable, data scan is high

```
vs
SELECT product_id, COUNT(*) FROM amazon_reviews_tsv
WHERE marketplace='US' AND extract(year from review_date) = 2013
GROUP BY 1 ORDER BY 2 DESC LIMIT 10
```

[tsv] -> (Run time: 1 minute 43 seconds, Data scanned: 32.23 GB). Performance poor and data scan is higher.

In above examples, we saw partitioning saves you query costs and query time. But how would you optimize if you have high cardinality fields? (large number of distinct values). For such need, we can explore bucketing. With this data can be split for storage into many buckets that will have roughly the same amount of data. Please note that Athena leverages Hive bucketing, which is different from S₃ buckets.

Run CTAS to explore bucketing

 Let's assume a case where we have need to query by Review_id more often to pull its specific information. Since Review_id has high cardinality, bucketing is better strategy to split the data. But before optimizing, let's run the below query on original TSV data to fetch product ID by review Id.

```
SELECT "$path", product_id FROM amazon_reviews_tsv
WHERE review_id='RWKE7RNMWTQYT'
```

Query Takes Run time: 1 minute 30 seconds, Data scanned: 32.21 GB

Note: "spath" returns the path in S₃ where the data for the review_id is located.

2. Now run the same query on amazon_reviews_by_marketplace query which is already partitioned by marketplace and year and is parquet.

```
SELECT "$path", product_id FROM amazon_reviews_by_marketplace
WHERE review_id='RWKE7RNMWTQYT'
```

This query takes Run time: 5.51 seconds, Data scanned: 1.46 GB

parquet being columnar format is able to retrieve results much faster.

3. Now let's use CTAS to create bucketed table on "Review_ID" as our query is just by review_id. We can't use normal partition as it would result in creating too many partitions and fail on max limit.

```
CREATE TABLE amazon_reviews_unsorted
WITH (
format='Parquet',
bucketed_by = ARRAY['review_id'],
bucket_count = 30
) AS
SELECT review_id, product_id, customer_id, product_parent, star_rating, helpful_votes,
total_votes, verified_purchase, marketplace, product_category, review_date
FROM amazon_reviews_by_marketplace
```

4. Now try the guery on the bucket partitioned table and check it run time stats.

```
SELECT "$path", product_id FROM amazon_reviews_unsorted WHERE review_id='RWKE7RNMWTQYT'
```

Run time: 3.43 seconds, Data scanned: 51.92 MB

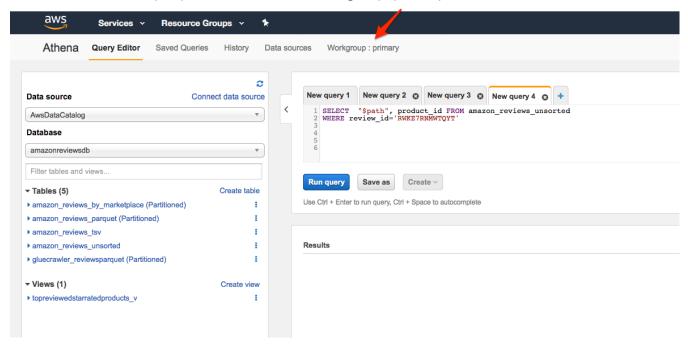
As we have bucketed data on reviewId, Athena only queries the specific buckets by first checking the parguet meta data thereby reducing time and scan size.

Athena Workgroups

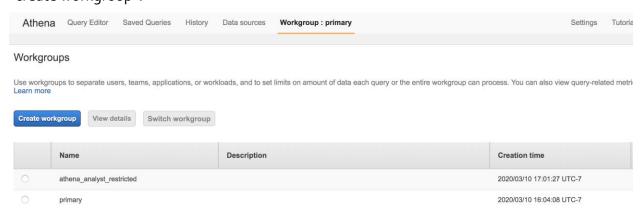
Use workgroups to separate users, teams, applications, or workloads, to set limits on amount of data each query or the entire workgroup can process, and to track costs. Because workgroups act as resources, you can use resource-level identity-based policies to control access to a specific workgroup

Creating workgroup and setting controls

- 1. In this lab, we will create a new workgroup and define data usage controls.
- 2. Go to Athena query editor and click on workgroup: primary

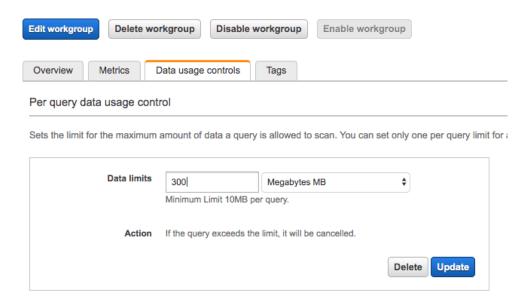


Select "create workgroup" and provide workgroup name - "athena_analyst_restricted".
 Add Query result location. Other options can be left default. Go ahead and select "create workgroup".



4. Once group is created. Select the group and view details. Select tab "data usage controls". Enter 300 in Data limits for Megabytes MB and update.

Workgroup: athena_analyst_restricted



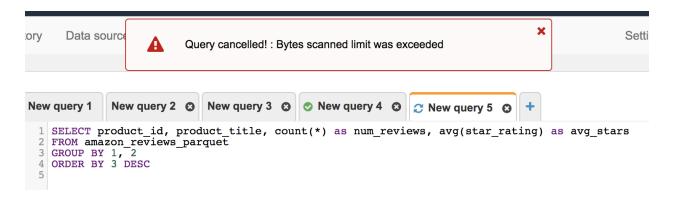
- 5. Go back to query editor and select workgroup and switch to this new group. You will notice, it doesn't carry over the query history as workgroups separate out query execution and query history.
- 6. Execute following query. Query should run successfully and scan 215 MB. Notice we have filter on partition column and this is parquet formatted table

SELECT product_id, product_title, count(*) as num_reviews, avg(star_rating) as avg_stars FROM amazon_reviews_parquet where product_category='Toys' GROUP BY 1, 2 ORDER BY 3 DESC

(Run time: 8.05 seconds, Data scanned: 215.04 MB)

Execute following query without filter on partition column. You should get error message "Bytes Scanned limit was exceeded"

SELECT product_id, product_title, count(*) as num_reviews, avg(star_rating) as avg_stars FROM amazon_reviews_parquet GROUP BY 1, 2 ORDER BY 3 DESC



Above steps shows control limits on per-query execution within a workgroup. You can also define workgroup-wide data usage control limit.