HIVE CASE STUDY

SUBMITTED BY:

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PROBLEM STATEMENT:

On the clickstream cosmetic data, we are required to gain insights and analyse our customer browsing patter which will help to increase profit of the company. This can be done through analysing customer browsing pattern on which products, purchases and views when they are browsing the tracking the clicks and pattern.

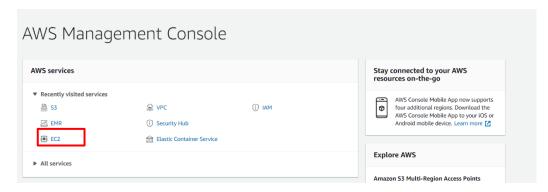
IMPLEMENTATION PHASE:

The implementation phase can be divided into the following parts:

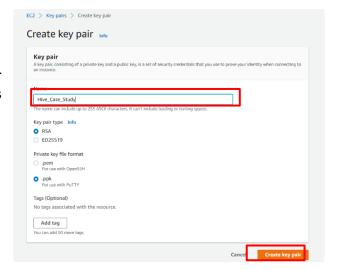
- Copying the data set into the HDFS:
- Launch an EMR cluster that utilizes the Hive services, and
- Move the data from the S3 bucket into the HDFS
- Creating the database and launching Hive queries on your EMR cluster:
- Create the structure of your database,
- Use optimized techniques to run your queries as efficiently as possible
- Show the improvement of the performance after using optimization on any single query.
- Run Hive queries to answer the questions given below.
- Cleaning up
- Drop your database, and
- Terminate your cluster

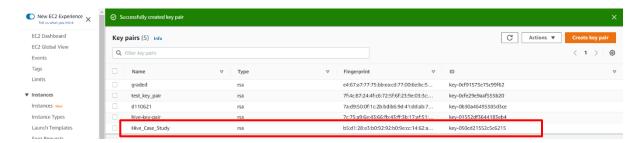
STEPS: CREATING KEY-PAIR

On the AWS Services, click on EC2 then 'Create key pair'.



Fill the Name section, as in our case, our key pair name is 'Hive_Case_Study'. It is directly created as .ppk file. Next click on 'Create key pair'.

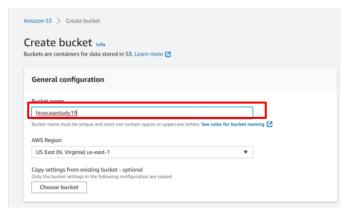




Our key pair 'Hive Case Study' has been created and downloaded.

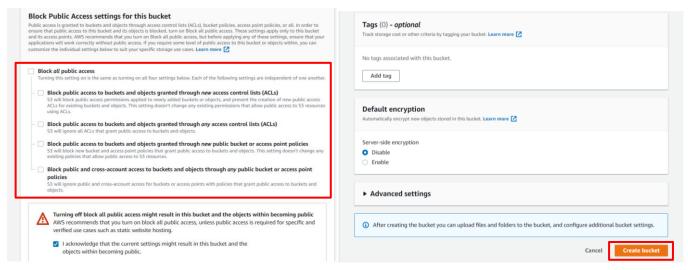
STEPS: CREATING BUCKET FOR THE CASE STUDY

We need to store our files in S3, thus the first step to serve the purpose is to 'Create Bucket'. On 'Amazon S3' go to 'Create bucket'.

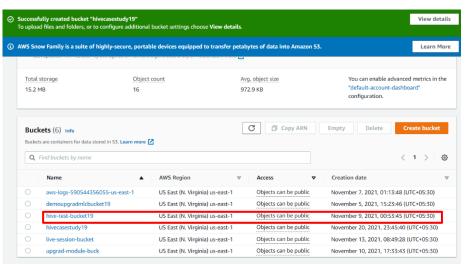


Field our Bucket name, do note that the bucket name should be unique.

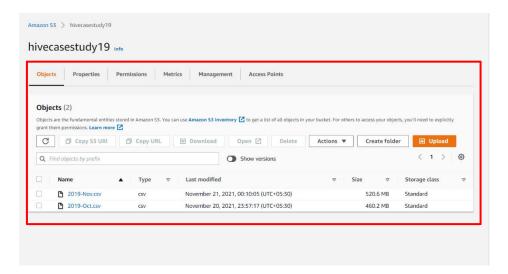
When we are creating our Bucket, we need to enable our Public Access settings for the Bucket. Untick the following option > Create Bucket.



As we can see, our bucket 'hivecasestudy19' has successfully created. We should now be ready to upload our 2019-Oct.csv and 2019-Nov.csv files into the Bucket.

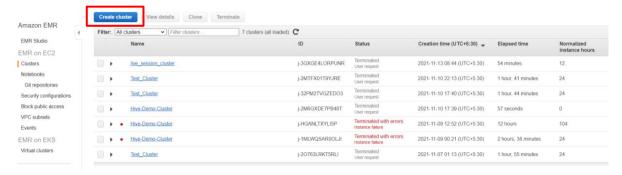


In our Bucket, we uploaded both of our files successfully. We will proceed now with our cluster creation.



STEPS: EMR (ELASTIC MAP REDUCE) - CREATING CLUSTER

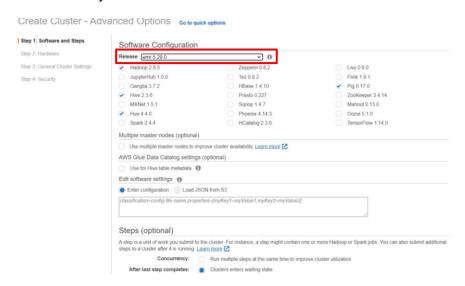
When we type EMR in the AWS Services, the below screen will pop up. As you can see, we are trying to create a cluster for this purpose of case study. Hence, below are the next steps.



Click on the 'Create Cluster' then 'Go to advanced options'.

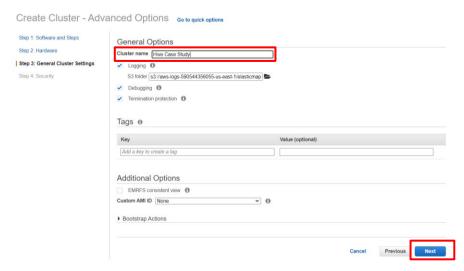


On the Advance Options, 'Step 1: Software and Steps' we will choose the *EMR RELEASED 5.29.0* for this case study. And we clicked 'Next'.

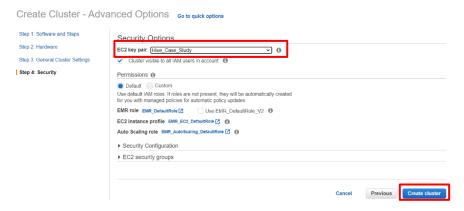


Note that for the purpose of case study, we have selected EMR Released 5.29.0 as some queries might run longer in the latest released.

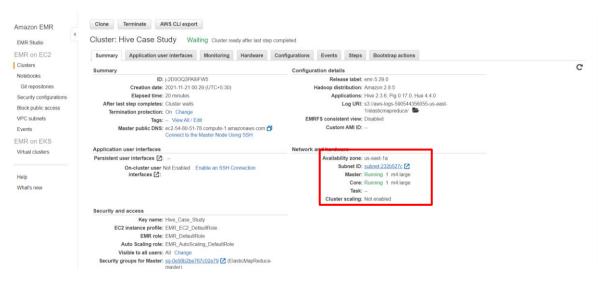
Our cluster name in the case study is 'Hive Case Study'. And then, we clicked 'Next'



In the Security Options, we will be using our EC2 Key-Pair 'Hive_Case_Study' which had been downloaded earlier, and we clicked on 'Create Cluster'.



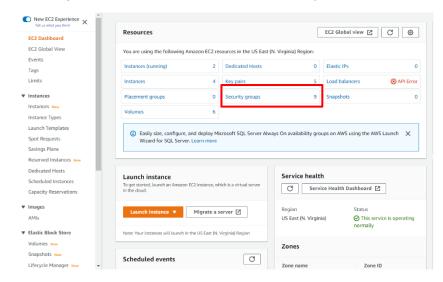
As you can see, our cluster is ready. For this case study, we will be using **2** node cluster which consist of **1** Master Node (m4.large) and **1** Core Node (m4.large). And both of these nodes are ready and running.



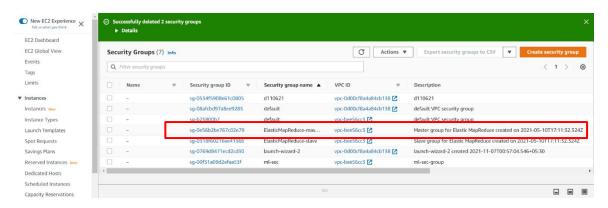
STEPS: SECURITY GROUPS

Once the cluster has successfully created, we have to edit the 'inbound rule'.

We will go to the EC2 and click on the 'Security Groups'.



Click on the Master Group.

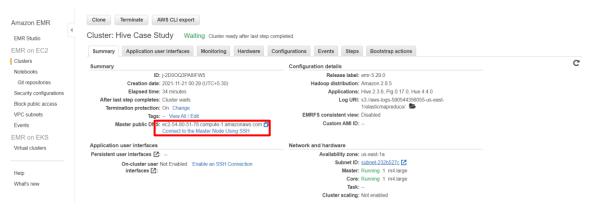


Save the 'SSH' inbound rules to 'Anywhere'



STEPS: CONNECTING THE MASTER NODE WITH PUTTY

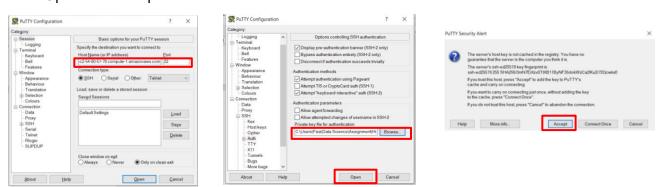
Go to our ready cluster and click on the 'Connect to the Master Node Using SSH', another screen will pop up:



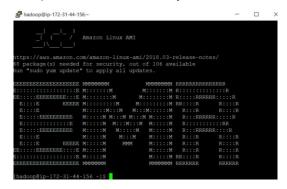
Important notes to be remembered here is our 'Host Name field' as well as our PPK file name saved earlier. Copied the 'Host Name field'.



Launch 'Putty', paste the 'Host Name' address which had been copied earlier in the 'Host Name' field. On the left side, click on 'SSH' > 'Auth' then browse the PPK file, in our case 'Hive_Case_Study.ppk', click 'Open' the 'Accept'.



EMR (Elastic Map Reduce) CLI has successfully launched.



STEPS: QUERYING IN HADOOP AND HIVE

Check the services running in Hadoop cluster

Command: sudo initctl list

```
[hadoop@ip-172-31-44-156 ~]$ sudo initctl list
rc stop/waiting
tty (/dev/tty3) start/running, process 4946
tty (/dev/tty2) start/running, process 4944
tty (/dev/tty1) start/running, process 4941
tty (/dev/tty6) start/running, process 4957
tty (/dev/tty5) start/running, process 4957
tty (/dev/tty4) start/running, process 4952
tty (/dev/tty4) start/running, process 4948
update-motd stop/waiting
hive-server2 start/running, process 14480
hadoop-mapreduce-historyserver start/running, process 12867
hadoop-yarn-timelineserver start/running, process 12154
plymouth-shutdown stop/waiting
whisper-server stop/waiting
control-alt-delete stop/waiting
hive-hcatalog-server start/running, process 15245
```

Load the data sets into HDFS from S3

Verifying the inbuilt file directories in HDFS.
 Command: *hadoop fs -ls* /

```
[hadoop@ip-172-31-44-156 ~]$ hadoop fs -ls /
Found 4 items
drwxr-xr-x - hdfs hadoop 0 2021-11-20 19:05 /apps
drwxrwxrwt - hdfs hadoop 0 2021-11-20 19:07 /tmp
drwxr-xr-x - hdfs hadoop 0 2021-11-20 19:05 /user
drwxr-xr-x - hdfs hadoop 0 2021-11-20 19:05 /var
```

Creating a directory for our case study
 Create directory command: hadoop fs -mkdir /user/hive/hivecasestudy

Verifying the directory command: hadoop fs -ls /user/hive/

```
[hadoop@ip-172-31-44-156 ~]$ hadoop fs -mkdir /casestudy /user/hive/hivecasestud

y
[hadoop@ip-172-31-44-156 ~]$ hadoop fs -ls /user/hive/
Found 2 items
drwxr-xr-x - hadoop hadoop 0 2021-11-20 19:45 /user/hive/hivecasestud

y
drwxrwxrwt - hdfs hadoop 0 2021-11-20 19:05 /user/hive/warehouse
[hadoop@ip-172-31-44-156 ~]$
```

From the above screenshot, we can see that the new directory is successfully created.

- Load the data from S3 bucket to the HDFS
 - A. For our clickstream data, Month of Oct-2019. Command:

hadoop distcp s3://hivecasestudy19/2019-Oct.csv User/hive/hivecasestudy/2019-Oct.csv

```
drwxrwxrwt - hdfs hadoop 0 2021-11-20 19:05 /user/hive/warehouse [hadoop@ip-172-31-44-156 ~]$ hadoop distcp s3://hivecasestudy19/2019-Oct.csv /user/hive/hivecasestudy/2019-Oct.csv 21/11/20 19:50:11 INFO tools.DistCp: Input Options: DistCpOptions{atomicCommit=false, syncFolder=false, delete Missing=false, ignoreFailures=false, overwrite=false, skipCRC=false, blocking=true, numListstatusThreads=0, ma xMaps=20, mapBandwidth=100, sslConfigurationFile='null', copyStrategy='uniformsize', preserveStatus=[], preser veRawXattrs=false, atomicWorkPath=null, logPath=null, sourceFileListing=null, sourcePaths=[s3://hivecasestudy19/2019-Oct.csv], targetPath=/user/hive/hivecasestudy/2019-Oct.csv, targetPathExists=false, filtersFile='null'} 21/11/20 19:50:11 INFO client.RMProxy: Connecting to ResourceManager at ip-172-31-44-156.ec2.internal/172.31.4
```

B. Clickstream data, Month of Nov-2019.

Command:

hadoop distcp s3://hivecasestudy19/2019-Nov.csv /user/hive/hivecasestudy/2019-Nov.csv

[hadoop@ip-172-31-44-156 ~]\$ hadoop distcp s3://hivecasestudy19/2019-Nov.csv /user/hive/hivecasestudy/2019-Nov.csv 21/11/20 19:53:39 INFO tools.DistCp: Input Options: DistCpOptions{atomicCommit=false, syncFolder=false, deleteMissing=false, ignoreFailures=false, overwrite=false, skipCRC=false, blocking=true, numListstatusThreads=0, maxMaps=20, mapBandwidth=100, sslConfigurationFile='null', copyStrategy='uniformsize', preserveStatus=[], preserveRawXattrs=false, atomicWorkPath=null, logPath=null, sourceFileListing=null, sourcePaths=[s3://hivecasestudy19/2019-Nov.csv], targetPath=/user/hive/hivecasestudy/2019-Nov.csv, targetPathExists=false, filtersFile='null'} 21/11/20 19:53:40 INFO client.RMProxy: Connecting to ResourceManager at ip-172-31-44-156.ec2.internal/172.31.44.156:8032

Verifying the loaded data in HDFS. Command: hadoop fs -ls /user/hive/hivecasestudy

From the above screenshots, we can see that both of our files has been successfully loaded.

View the loaded data

Command: hadoop fs -cat /user/hive/hivecasestudy/2019-Oct.csv |head

Command: hadoop fs -cat /user/hive/hivecasestudy/2019-Nov.csv |head

```
Tat: Unable to write to output stream.

[hadoop@ip-172-31-44-156 ~]$ hadoop fs -cat /user/hive/hivecasestudy/2019-Nov.csv | head

event time, event_type, product_id, category_id, category_code, brand, price, user_id, user_session

2019-11-01 00:00:02 UTC, view, 5802432, 1487580009286598681,,,0.32,562076640,09fafd6c-6c99-46b1-834f-33527f4de241

2019-11-01 00:00:09 UTC, cart, 5844397,1487580006317032337,,2.38,553329724,2067216c-31b5-455d-a1cc-af0575a34ffb

2019-11-01 00:00:10 UTC, view, 5837166,1783999064103190764,,pnb,22.22,556138645,57ed222e-a54a-4907-9944-5a875c2d7f4f

2019-11-01 00:00:11 UTC, cart,5876812,1487580010100293687,jessnail,3.16,564506666,186c1951-8052-4b37-adce-dd9644b1d5f7

2019-11-01 00:00:24 UTC, remove_from_cart,5826182,1487580007483048900,,,3.33,553329724,2067216c-31b5-455d-a1cc-af0575a34ff

2019-11-01 00:00:24 UTC, remove_from_cart,5826182,1487580007483048900,,,3.33,553329724,2067216c-31b5-455d-a1cc-af0575a34ff

2019-11-01 00:00:25 UTC, view,5856189,1487580009026551821,,runail,15.71,562076640,09fafd6c-6c99-46b1-834f-33527f4de241

2019-11-01 00:00:32 UTC, view,5837835,1933472286753424063,,,3.49,514649199,432a4e95-375c-4b40-bd36-0fc039e77580

2019-11-01 00:00:34 UTC,remove_from_cart,5870838,1487580007675986893,,milv,0.79,429913900,2f0bff3c-252f-4fe6-afcd-5d8a6a9

2039-
204: Unable to write to output stream.

[hadoop@ip-172-31-44-156 ~]$
```

Create database

Launch Hive: Hadoop > Hive.

```
cat: Unable to write to output stream.
[hadoop@ip-172-31-44-156 ~]$ hive

Logging initialized using configuration in file:/etc/hive/conf.dist/hive-log4j2.properties Async: false
hive>
```

Creating a database

Command: Create database if not exists ecom;

2. Use database created

Command: use ecom;

3. Verifying the database created

Command: show databases;

```
Logging initialized using configuration in file:/etc/hive/conf.dist/hive-log4j2.properties Async: false hive> create database if not exists ecom;
OK
Time taken: 0.944 seconds hive> use ecom;
OK
Time taken: 0.052 seconds hive> show databases;
OK
default ecom
Time taken: 0.173 seconds, Fetched: 2 row(s) hive>
```

Create a table

1. Create a table using CSVSerde

Command:

CREATE EXTERNAL TABLE IF NOT EXISTS retail (event_time timestamp, event_type string, product_id string, category_id string, category_code string, brand string, price float,user_id bigint, user_session string) ROW FORMAT SERDE 'org.apache.hadoop.hive.serde2.OpenCSVSerde' STORED AS TEXTFILE LOCATION '/user/hive/hivecasestudy' tblproperties("skip.header.line .count"="1");

```
hive> CREATE EXTERNAL TABLE IF NOT EXISTS retail (event_time timestamp, event_type string, product_id string, category id string, category_code string, brand string, price float, user_id bigint, user_session string) ROW FORMAT SERDE 'org.apach e.hadoop.hive.serde2.OpenCSVSerde' STORED AS TEXTFILE LOCATION '/user/hive/hivecasestudy' tblproperties("skip.header.line .count"="1");
OK
Time taken: 0.371 seconds
```

2. Set the display for the header column Command: set hive.cli.print.header = true;

```
Time taken: 0.371 seconds
hive> set hive.cli.print.header = true;
hive>
```

3. Verifying the table creation by checking the top 5 rows in the table. Command: **select** * **from retail limit 5**;

STEPS: APPLYING OPTIMIZATION TECHNIQUES (PARTITIONING & BUCKETING)

Enable Dynamic Partitioning

Command:

set hive.exec.dynamic.partition.mode = nonstrict; set hive.exec.dynamic.partition = true;

Enable Bucketing

Command: **set hive.enforce.bucketing = true**;

```
hive> set hive.exec.dynamic.partition.mode = nonstrict;
hive> set hive.exec.dynamic.partition = true;
hive> set hive.enforce.bucketing = true;
hive>
```

Create an optimized table

Command:

CREATE TABLE IF NOT EXISTS dynpart_bucket_retail(event_time timestamp, product_id string, category_id string, category_code string, brand string, price float, user_id bigint, user_session string) PARTITIONED BY (event_type string) CLUSTERED BY (price) INTO 10 BUCKETS ROW FORMAT SERDE 'org.apache.hadoop.hive.serde2.OpenCSVSerde' STORED AS TEXTFILE LOCATION '/user/hive/hivecasestudy' tblproperties('skip.header.line.count' = '1');

```
hive> set hive.enforce.bucketing = true;
hive> CREATE TABLE IF NOT EXISTS dynpart_bucket_retail(event_time timestamp, product_id string, category_id string, category_code
string, brand string, price float, user_id bigint, user_session string) PARTITIONED BY (event_type string) CLUSTERED BY (price) IN
TO 10 BUCKETS ROW FORMAT SERDE 'org.apache.hadoop.hive.serde2.OpenCSVSerde' STORED AS TEXTFILE LOCATION '/user/hive/hivecasestudy'
tblproperties('skip.header.line.count' = '1');
OK
Time taken: 0.071 seconds
```

For this table optimization based on the partitioning and bucketing, we have decided to create a partition on 'event_type) into 10 buckets and clustered by 'price'.

Verifying the table Command: show tables;

```
hive> show tables;

OK

tab_name
dynpart_bucket_retail
retail
Time taken: 0.043 seconds, Fetched: 2 row(s)
```

As you can see, the table with partitions and bucketing named 'dynpart_bucket_retail' has been successfully created.

Insert the data into the optimized table Command:

INSERT INTO TABLE dynpart_bucket_retail PARTITION (event_type) SELECT event_time,product_id, category_id, category_code, brand, price, user_id, user_session, event_type FROM retail;

 Verifying the table created in Hadoop Command: hadoop fs -ls /user/hive/hivecasestudy

As you can see the partitioned files has been created and partitioned by 'event_type' which consists of cart, purchase, remove_from_cart and view. Let's explore further on the partitioning.

Command: hadoop fs -ls /user/hive/hivecasestudy/event_type=purchase

```
hadoop@ip-172-31-44-156 ~]$ hadoop fs -ls /user/hive/hivecasestudy/event_type=purchase
ound 10 items
                  hadoop hadoop
                                         6241877 2021-11-20 20:18 /user/hive/hivecasestudy/event_type=purchase/000000_0
                                         7235640 2021-11-20 20:19 /user/hive/hivecasestudy/event_type=purchase/000001_0
                                        7231471 2021-11-20 20:19 /user/hive/hivecasestudy/event_type=purchase/000002_0
                                        7526313 2021-11-20 20:18 /user/hive/hivecasestudy/event_type=purchase/000003_0 7227979 2021-11-20 20:18 /user/hive/hivecasestudy/event_type=purchase/000004_0 7310389 2021-11-20 20:18 /user/hive/hivecasestudy/event_type=purchase/000005_0
rwxr-xr-x
                1 hadoop hadoop
rwxr-xr-x
                1 hadoop hadoop
rwxr-xr-x
                                        8915123 2021-11-20 20:19 /user/hive/hivecasestudy/event_type=purchase/000006_0
                 hadoop hadoop
                                        5366094 2021-11-20 20:19 /user/hive/hivecasestudy/event_type=purchase/000007_0 6469070 2021-11-20 20:18 /user/hive/hivecasestudy/event_type=purchase/000008_0
rwxr-xr-x
rwxr-xr-x
               1 hadoop hadoop
                 hadoop hadoop
                                        8004214 2021-11-20 20:18 /user/hive/hivecasestudy/event_type=purchase/000009_0
rwxr-xr-x
```

For the 'event_type=purchase', there are exactly total of 10 buckets has been created.

Verifying the performance for the tables before and after optimization.

A. Retail table without optimization

Command: select * from retail limit 5;

```
ve> select * from retail limit 5;
                                 5802432 1487580009286598681
                                                                                                               09fafd6c-6c99-46b1-834f-3352
019-11-01 00:00:09 UTC cart
                                 5844397 1487580006317032337
                                                                                                               2067216c-31b5-455d-a1cc-af0
019-11-01 00:00:10 UTC view
c2d7f4f
                                 5837166 1783999064103190764
                                                                                                               57ed222e-a54a-4907-9944-5a8
                                                                                     22.22
019-11-01 00:00:11 UTC cart
                                 5876812 1487580010100293687
                                                                                                                        186c1951-8052-4b37-
lce-dd9644b1d5f7
019-11-01 00:00:24 UTC remove from cart
-455d-a1cc-af0575a34ffb
Pime taken: 2.355 seconds, Fetched: 5 row(s)
```

B. Dynpart_bucket_retail table with optimization

Command: select * from dynpart bucket retail limit 5;

Insights:

1. Before optimization for the 'retail table' it tooks 2.355 seconds to retrieve the data from the query, however with the dynpart_bucket_table (after optimization), it tooks only 0.204 seconds to read the data. Difference of 2.151 seconds, it's too early to make any assumptions, lets look into further queries.

QUESTIONS

Question 1: Find the total revenue generated due to purchases made in October.

Without Optimization (Retail Table)

Command:

select sum(price) as october_revenue from retail where month(event_time) = '10' AND event_type = 'purchase';

With Optimization (Dynpart_bucket_retail Table)

Command:

select sum(price) as october_revenue from dynpart_bucket_retail where month(event_time) = '10' AND event_type = 'purchase';

- 1. In this query, we are required to find the total revenue generated on October 2019 based on the Purchases.
- 2. The time taken to read the data for the 'retail table (without optimization)' is 135.866 seconds, while for the 'dynpart_bucket_retail table (with optimization)' is 24.25 seconds, which is 5x lesser than the retail table.
- 3. To answer the above questions, there are total of 1211538.43 +/- revenue generate on October on Purchases.
- 4. There are huge differences in the time taken to retrieve the data from both tables, in this query, the optimization table performed faster than 'retail table' which is without optimization.

Question 2: Write a guery to yield the total sum of purchases per month in a single output.

Without Optimization (Retail Table)

Command:

select month(event_time) as month, count(event_type) as total_purchases from retail where event_type
= 'purchase' group by month(event_time);

With Optimization (Dynpart_bucket_retail Table)

Command:

select month(event_time) as month, count(event_type) as total_purchases from dynpart_bucket_retail where event_type = 'purchase' group by month(event_time);

```
hive> select month(event_time) as month, count(event_type) as total_purchases from dynpart_bucket_retail where event_type = 'purchase' group by month (event_time);

Query ID = hadoop_20211120204702_b836adb6-2635-4453-b78a-053b18fe0890
Total jobs = 1

Launching Job 1 out of 1

Status: Running (Executing on YARN cluster with App id application_1637435175999_0007)

VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

Map 1 ....... container SUCCEEDED 3 3 0 0 0 0

Reducer 2 ..... container SUCCEEDED 1 1 0 0 0 0

VERTICES: 02/02 [------->] 100% ELAPSED TIME: 23.28 s

OK

10 245619
11 322412
Time taken: 24.022 seconds, Fetched: 2 row(s)
```

- 1. In this next question, we are required to find the total sum for the purchases for the month of October and November.
- 2. The Optimized table again prove the faster query with only 24.02 seconds for the time taken and the ratil table took 102.85 seconds to retrieve the data.
- 3. For the month of October, the total purchases is 245624 and 322417 for the month of November. There are a significant improvement on the purchase value which is around 30%.

Question 3: Write a query to find the change in revenue generated due to purchases from October to November.

Without Optimization (Retail Table)

Command:

select (sum(case when month(event_time)=11 then price else 0 end) - sum(case when month(event_time)=10 then price else 0 end)) as change_in_rev from retail where event_type = 'purchase' and month(event_time) in ('10','11');

With Optimization (Dynpart_bucket_retail Table)

Command:

select (sum(case when month(event_time)=11 then price else 0 end) - sum(case when month(event_time)=10 then price else 0 end)) as change_in_rev from dynpart_bucket_retail where event_type = 'purchase' and month(event_time) in ('10','11');

```
hive> select (sum(case when month(event_time)=11 then price else 0 end) - sum(case when month(event_time)=10 then price else 0 end)) as change_in_r
ev from dynpart bucket retail where event_type = 'purchase' and month(event_time) in ('10','11');
Query ID = hadoop_20211120205605_3d20ec40-d72c-436c-9fd2-de05c52d2018
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application_1637435175999_0008)

VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

Map 1 ....... container SUCCEEDED 3 3 0 0 0 0
Reducer 2 ..... container SUCCEEDED 1 1 0 0 0 0
VERTICES: 02/02 [===========>>] 100% ELAPSED TIME: 25.44 s

OK
319437.7899997565
Time taken: 26.109 seconds, Fetched: 1 row(s)
```

- 1. In this next question, we will find out on the differences in the revenue generated based on purchases for the month of October and November.
- 2. The difference of revenue generated between the month of October and November 2019 are 319478.47 +/-.
- 3. It took 112.009 second for the Retail table to read the data from the above query, however only 26.109 seconds time taken from the optimization table. Again, the Optimized table performed faster in this analysis. Hence, for the next remaining questions, we will be using the optimized tables to run the queries.

Question 4: Find distinct categories of products. Categories with null category code can be ignored.

With Optimization (Dynpart_bucket_retail Table)

Command:

select distinct split(category_code,'\\.')[0] as distinct_category from dynpart_bucket_retail where category_code != ";

```
hive> select distinct split(category_code,'\\.')[0] as distinct category from dynpart_bucket_retail where category_code != '';
Query ID = hadoop_20211120205910_a54Ic0d4-a105-4d6f-ba75-a28001eae34c
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application_1637435175999_0008)

VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

Map 1 ...... container SUCCEEDED 6 6 6 0 0 0 0
Reducer 2 ..... container SUCCEEDED 5 5 0 0 0 0
VERTICES: 02/02 [===========>>] 100% ELAPSED TIME: 64.51 s

OK
furniture
appliances
accessories
apparel
sport
stationery
Time taken: 65.263 seconds, Fetched: 6 row(s)
```

- 1. The time take to retrieve the data to execute the query is 65.263 seconds.
- 2. This question required us to find out the categories present in the data, and there are total of 6 categories available, which are:
 - Furniture
 - Appliances
 - Accessories
 - Sport
 - Apparel
 - Stationery

Question 5: Find the total number of products available under each category

With Optimization (Dynpart_bucket_retail Table)

Command:

select split(category_code,'\\')[0] as category, count(product_id) as num_of_products from dynpart_bucket_retail where category_code != "group by split(category_code,'\\')[0] order by num_of_products desc;

Insights:

- 1. Based from the categories listed from the previous question, we are required to find the total number of products in each of the 6 categories.
- 2. To execute this query, it took 65.36 seconds.
- 3. And the output are as follows:

No.	Category	Number of Products
1.	Appliances	61736
2.	Stationery	26722
3.	Furniture	23604
4.	Apparel	18232
5.	Accessories	12928
6.	Sports	2

Appliances lead the total number of products with 61736, followed closely by Stationery and Furniture. Surprisingly Sports only consists of 2 products.

Question 6: Which brand had the maximum sales in October and November combined?

With Optimization (Dynpart_bucket_retail Table)

Command:

WITH total_sales_summary AS(

select brand, round((sum(case when month(event_time)=10 then price else 0 end) + sum(case when month(event_time)=11 then price else 0 end)),2) as total_sales from dynpart_bucket_retail where event_type = 'purchase' and month(event_time) in ('10','11') and brand != '' group by brand)

select brand, total_sales from total_sales_summary order by total_sales desc limit 1;

- 1. In this question, we are required to find the name of the brand which had the total number of sales for BOTH month of October and November (combine).
- 2. The time taken to retrieve the data from the above query is only 35.324 seconds.
- 3. It turns out that brand Runail possessed the highest sales for the month of October AND November.

Question 7: Which brands increased their sales from October to November?

With Optimization (Dynpart_bucket_retail Table)

Command:

WITH brand_sales_summary AS(

select brand, round(sum(case when month(event_time)=10 then price else 0 end),2) as sales_october,round(sum(case when month(event_time)=11 then price else 0 end),2) as sales_november from dynpart_bucket_retail where event_type = 'purchase' and month(event_time) in ('10','11') and brand != '' group by brand)

select brand, sales_october, sales_november, round((sales_november - sales_october),2)as sales_differences from brand_sales_summary where sales_november - sales_october > 0 order by sales_differences desc;

```
tatukan 39.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.00 12.0.0
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182.67
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584.95
150.97
384.59
135.03
204.3
366.64
132.0
645.07
126.38
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Farmona 1692.46
Latinoil
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1843.43
249.52
293.07
70.53
233.52
230.38
513.66
945.51
762.31
874.17
243.36
152.61
2707.07
136.57
525.2
      farmona 1692.46
latinoil
miskin 158.04
elizavecca
mefertiti
finish 98.38
igrobeauty
dizao 819.13
osmo 645.58
carmex 145.08
eos 54.34
depilflax
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101.77
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98.27

2803.78 96.71

95.22

94.29

93.56

92.64

84.56

71.29

70.84

146.04 68.57

68.25

312.52 66.02

63.19

61.29

109.33 60.11

59.45

736.85 57.62

57.25

212.38 57.05

56.56

45.45

44.91

27.28
   carmex 145.08
54.34
depilflax
enjoy 41.35
kerasys 430.91
aura 83.95
plazan 101.37
nirvel 163.04
konad 739.83
agomania
rutrin 299.37
laboratorium
imm 288.02
dewal 0.0
tarutaka-foot
kares 0.0
                                                                                                                                                                                                                 177.51
194.01
507.29
234.33
810.67
77.47
367.62
246.5
351.21
      inm 288.02 351.21
dewal 0.0 61.29
marutaka-foot 49.22
kares 0.0 59.45
profhenna 679.23
koelcia 55.5 112.75
balbcare 155.33
elskin 251.09 307.65
foamie 35.04 80.49
ladykin 125.65 170.57
likato 296.06 340.97
mavala 409.04 446.32
kinetics 6334.25 6945.26 611.01 browmenna 14331.37 14916.73 airmails 5118.9 5691.52 572.62 uskusi 5142.27 5690.31 548.04 coifin 903.0 1428.49 525.49 11moni 1308.9 1796.6 487.7 matrix 3243.25 3726.74 483.49 gehwol 1089.07 157.68 468.61 greymy 29.21 489.49 460.28 bloaqua 942.89 1398.12 455.23 farmavita 837.37 1291.97 454.6 sophin 1067.86 1515.52 447.66 yu-r 271.41 673.71 402.3 kiss 421.55 817.33 395.78 naomi 0.0 389.0 389.0 lador 0.0 389.0 lador 
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78.74

90.31

931.09

444.81

93.36

63.4

24.26

401.22

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163.37

50.37
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iore 60.65
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rly 902.38
stelare
rofepil
lixz 38.95
inacil 0.0
rodefroy
rlysolid
eraclara
uno 0.0
ramill 63.01
reaclemoon
upertan
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471.802
24.45
24.25.12
91.59
71.21
21.08
18.48
181.49
66.51
12.39
329.17
10.14
10.03
9.64
8.39
4.57
4.3
   reaclemoon
upertan
arbie 0.0
eoproce
asyan 18.8
1y 17.14
ertio 236.16
aguar 1102.11
oleo 204.2
eoleor 43.41
ooyou 5.71
oodyton 1376.34
ekinity 8.88
elloganic
grace 100.92
rosima 20.23
vale 2.54
rime taken: 36.6
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316.84
28.94
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245.8
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212.53
51.7
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1380.64
12.44
0.0
102.61
20.93
3.1
                                                                                                                                                                                                                                                                                                                              3.56
3.1 3.1
1.69
0.7
0.56
nds, Fetched:
```

Insights:

- 1. To answer this question, we need to find the name of the brand which has increased their sales from the month of October and November.
- 2. It took only 36.658 seconds to retrieve the data from the Optimized table.
- 3. From the output, we can see that there are the total of 160 rows, which means, there are the total of 160 brands which has increased their sales in the subsequent month.
- 4. Grattol if the leading brand with the total increment of 2x which is 36027.17 increment in both of the month, followed closely by Uno brand with 15737.32 and Lianail with 10501.40. The least brand which has the lowest differences in Sales for the month of October and November are Ovale with 0.56, Cosima 0.70 and Grace with 1.69.

Question 8: Your company wants to reward the top 10 users of its websites with a golden customer plan. Write a query to generate a list of top 10 users who spend the most.

With Optimization (Dynpart_bucket_retail Table)

Command:

select user_id, round(sum(price),2) as amount_spends from dynpart_bucket_retail where event_type = 'purchase' group by user_id order by amount_spends desc limit 10;

Insights:

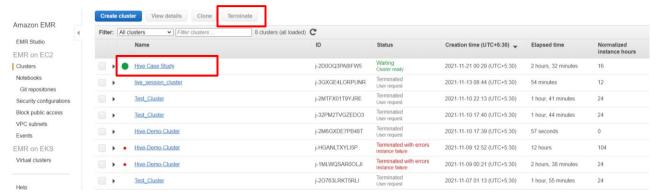
- 1. We need to find out the Top 10 users who spend the most on the website, which will won the Golden Customer Plan from our company.
- 2. The time taken for this query to be completed is 34.113 seconds.
- 3. Based from the output, we can see that the Top user spend 2715.87, followed by 1645.97 and 1352.85.

We have completed all of the questions, which needs to be run on the system, and we can conclude that Partitioning and Bucketing helps in increasing the performance by providing faster analysis on the query despite the huge amount of data provided. And it is efficient and convenient, as well as easy to code despite the volume of data loaded.

STEPS: TERMINATING EMR CLUSTER

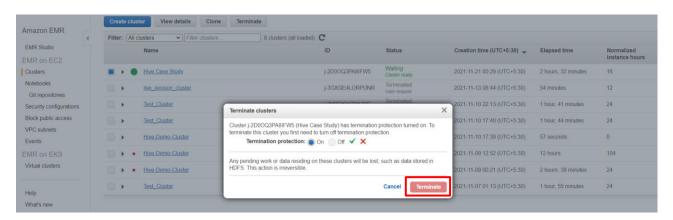
It is very important to terminate the cluster, because of the charges that will be charged into our account. Thus, we followed the below procedure to complete the above process:

On AWS Services, type EMR and go back to the main page of our Cluster.



Tick on our cluster, in this case 'Hive_Case_Study' cluster > click on Terminate.

A screen will pop up, we have unticked the 'On' and switch to 'Off' to allow the termination procedure to be completed, then click on 'Terminate'.



Our cluster and both of our Master and Core node has been terminated and with that we have completed ur hive Analysis on ClickStream dataset successfully.

