Hive Case Study

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Problem Statement:

With online sales gaining popularity, tech companies are exploring ways to improve their sales by analysing customer behaviour and gaining insights about product trends. Furthermore, the websites make it easier for customers to find the products they require without much scavenging. As part of this assignment, we will be challenging you, as a big data analyst, to extract data and gather insights from a real-life data set of an ecommerce company.

This is done by tracking their clicks on the website and searching for patterns within them. The clickstream data contains all the logs as to how the customer navigated through the website. It also contains other details such as time spent on every page, etc. From this, tech companies make use of data ingesting frameworks such as Apache Kafka or AWS Kinesis in order to store it in frameworks such as Hadoop. From there, machine learning engineers or business analysts use this data to derive valuable insights.

Data for this case study was available in:

https://e-commerce-events-ml.s3.amazonaws.com/2019-Oct.csv https://e-commerce-events-ml.s3.amazonaws.com/2019-Nov.csv

Objective:

To extract data and gather insights from an e-commerce company dataset, using AWS EMR and S3, Hadoop and Hive systems.

Steps Involved:

- 1. Launching an EMR Cluster
- 2. Loading the data into S3 and ingesting the data to HDFS

- 3. Launching and creating Hive Schema
- **4.** Optimization of tables
- **5.** Querying Assignment Questions
- **6.** Improvement of performance after using optimization on tables
- 7. Cleaning-up

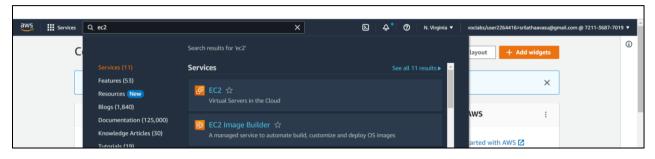
1. Launching an EMR Cluster:

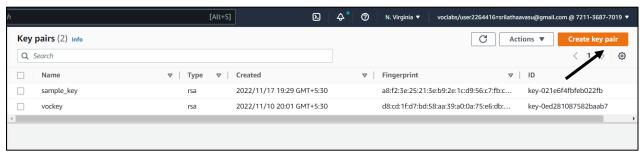
The following steps are followed while launching an EMR cluster,

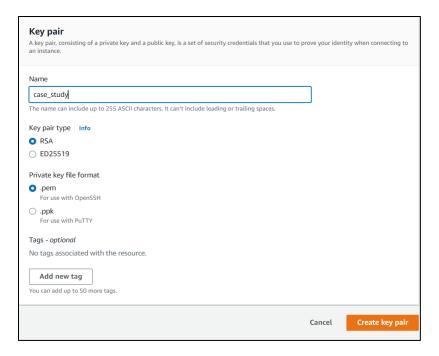
- i. Key-pair creation
- ii. Configuring & Launching an EMR cluster
- iii. Connecting to master node using SSH
- iv. Launching of Hive shell

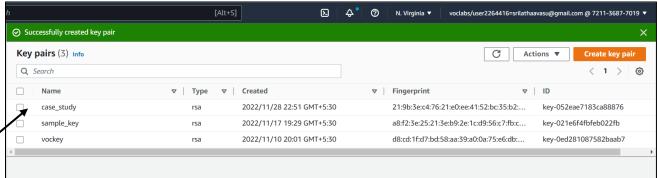
i. Key-pair creation:

We need to create an EC2 key-pair file and download it as .pem file.





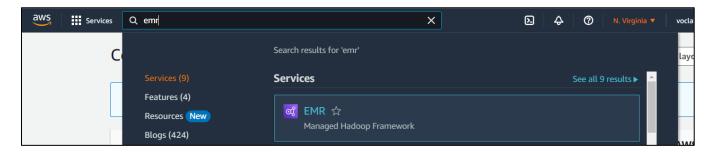


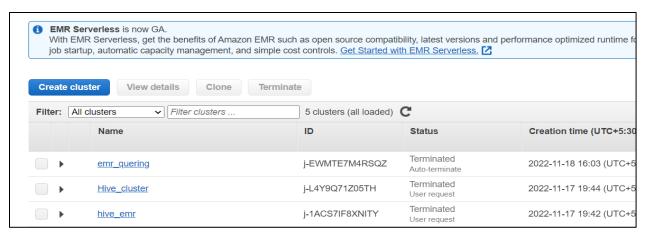


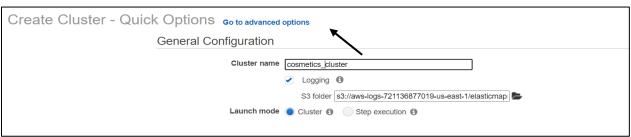
We will use this private key to securely connect to the EMR cluster through SSH.

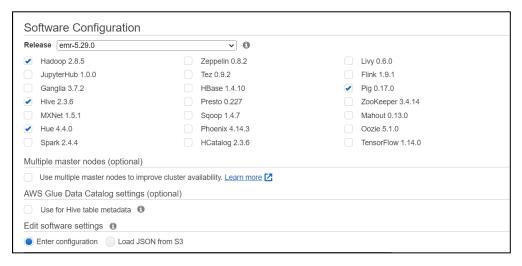
ii. Configuring & Launching an EMR cluster:

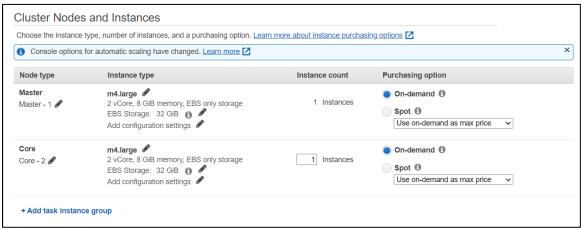
Amazon EMR (previously called Amazon Elastic MapReduce) is a managed cluster platform that simplifies running big data frameworks, such as Hadoop and Apache Spark on AWS to process and analyze vast amounts of data.



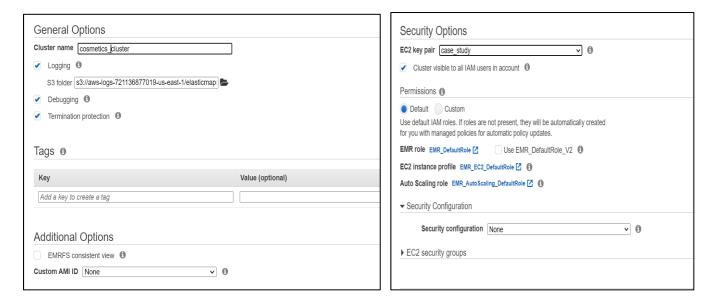




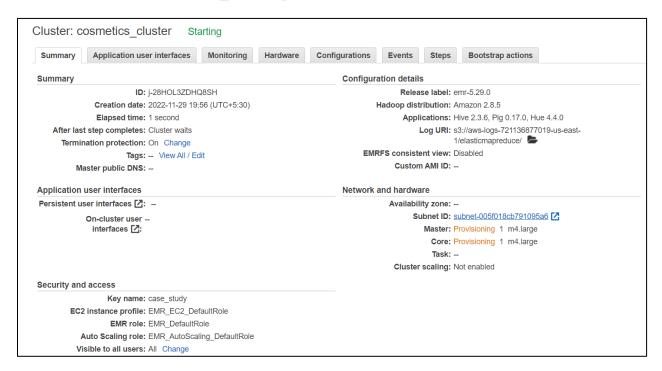




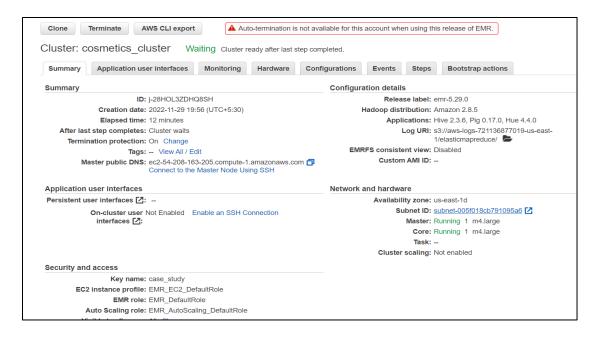
We have chosen emr-5.29.0 release for this case study.



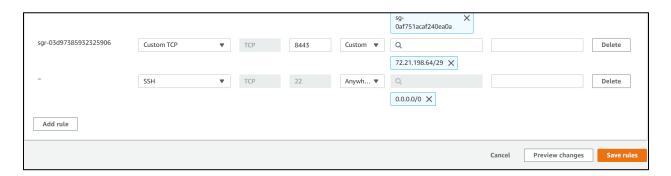
We have selected the "case_study" private key which we have created earlier.



Our cluster Hive Assignment is created and launched successfully and is now in "Waiting" state.

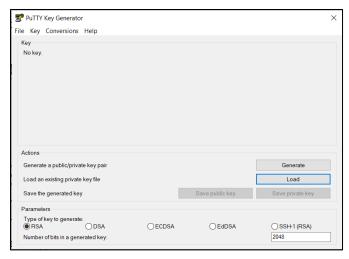


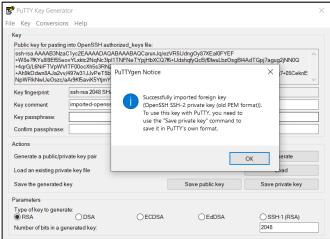
We need to ensure that the port is open to establish a connection. Under the cluster information page, click on the security groups of the master node. Here, we add an additional inbound rule as below.

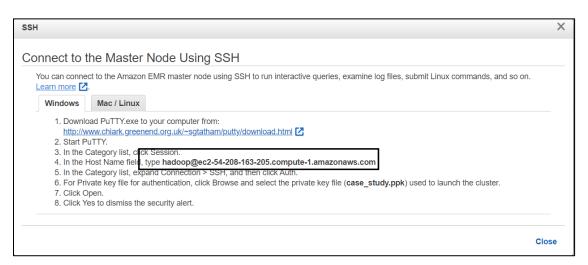


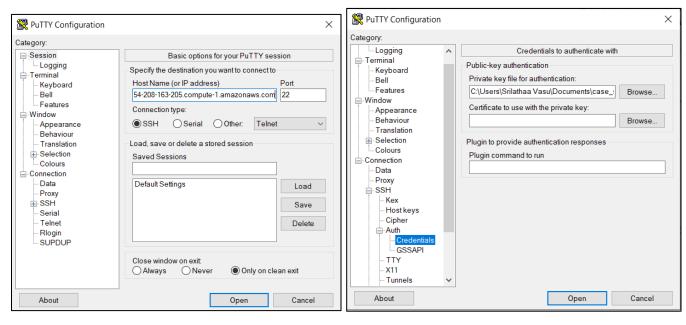
iii. Connecting to master node using SSH:

For connecting the master node using SSH, we need to use two programs namely, PuTTY and PuTTYgen. We first need to open PuTTYgen, then load the private key(case_study) and save it as .ppk file at a desired location. Next, we need to open PuTTY to connect to master node. We need to copy Master DNS address from the EMR cluster summary page.





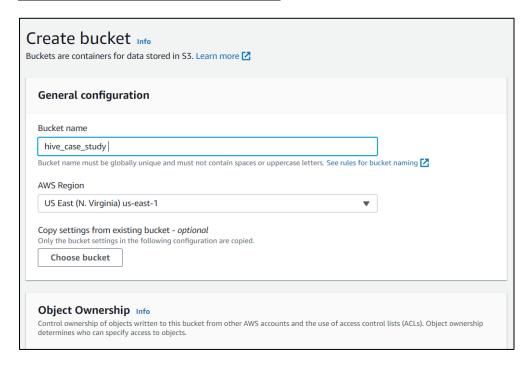


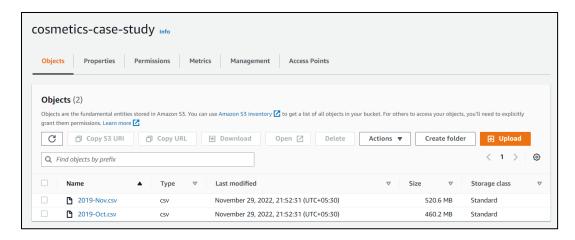


iv. Launching of Hive Shell:

2. Loading the data into S3 and ingesting the data to HDFS:

Uploading databases to S3 buckets:





Creating a temporary directory in HDFS:

Checking the directories already present in HDFS.

Command: hadoop fs -ls /

Output:

Creating a temporary directory.

Command: hadoop fs -mkdir /user/hivecasestudy hadoop fs -ls /user/

```
[hadoop@ip-172-31-87-236 ~]$ hadoop fs -mkdir /user/hivecasestudy
[hadoop@ip-172-31-87-236 ~]$ hadoop fs -ls /user/
Found 7 items
drwxrwxrwx - hadoop hadoop
                                    0 2022-11-29 14:33 /user/hadoop
drwxr-xr-x - mapred mapred
                                    0 2022-11-29 14:33 /user/history
                                    0 2022-11-29 14:33 /user/hive
drwxrwxrwx - hdfs
                    hadoop
                                    0 2022-11-29 16:50 /user/hivecasestudy
drwxr-xr-x - hadoop hadoop
drwxrwxrwx - hue
                     hue
                                    0 2022-11-29 14:33 /user/hue
                                    0 2022-11-29 14:33 /user/oozie
drwxrwxrwx
            - oozie oozie
                                    0 2022-11-29 14:33 /user/root
drwxrwxrwx
            - root
                     hadoop
[hadoop@ip-172-31-87-236 ~]$
```

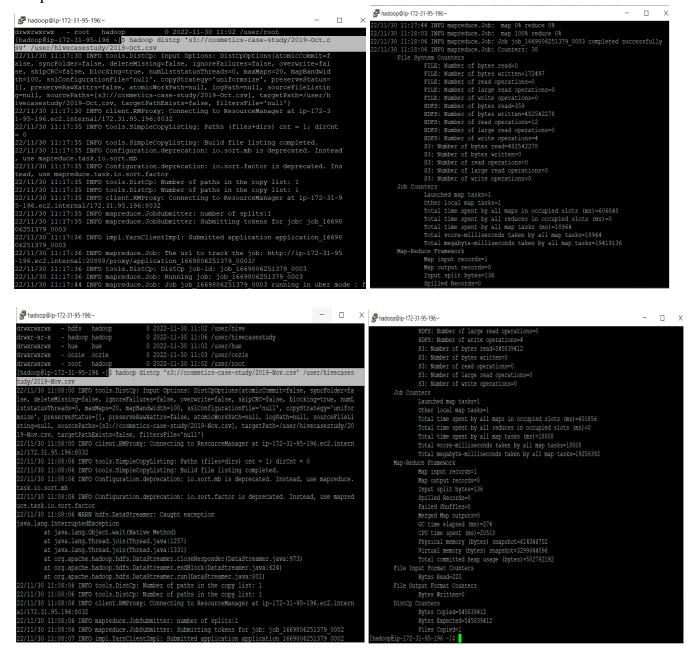
Loading the data to HDFS:

Since the size of the data is large we'll load the data into HDFS from S3 an into the local storage.

Command:

hadoop distcp 's3://cosmetics-case-study/2019-Oct.csv' /user/hivecasestudy/2019-Oct.csv hadoop distcp 's3://cosmetics-case-study/2019-Nov.csv' /user/hivecasestudy/2019-Nov.csv

Output:



Checking the loaded files:

Command: *hadoop fs -ls /user/hivecasestudy/*

Output:

We can see that the datasets have been loaded successfully.

3. Launching and creating Hive Schema:

Launching Hive:

Command: hive

Output:

```
[hadoop@ip-172-31-95-196 ~]$ hive

Logging initialized using configuration in file:/etc/hive/conf.dist/hive-log4j2.

properties Async: false

hive>
```

Creating a new database for the case study:

Command: *create database if not exists hive_casestudy; use hive_casestudy;*

Output:

```
hive> create database if not exists hive_casestudy;
OK
Time taken: 1.123 seconds
hive> use hive_casestudy;
OK
```

Create an external table to load the data:

Command: CREATE EXTERNAL TABLE IF NOT EXISTS sales_details(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 '/hivecasestudy' 
>TBLPROPERTIES("skip.header.line.count"="1");
```

```
hive> CREATE EXTERNAL TABLE IF NOT EXISTS sales_details(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/hivecasestudy/'

> TBLPROPERTIES ("skip.header.line.count"="1");
OK
Time taken: 0.542 seconds
```

Checking the table 'sales details':

Command: desc sales details;

Output:

```
hive> desc sales details;
OK
event time
                                                 from deserializer
event type
                        string
                                                 from deserializer
product id
                                                 from deserializer
category_id
                        string
                                                 from deserializer
category_code
                        string
                                                 from deserializer
brand
                                                 from deserializer
price
                                                 from deserializer
                                                 from deserializer
user_session
                        string
                                                 from deserializer
Fime taken: 0.18 seconds, Fetched: 9 row(s)
```

Loading the data into the 'sales_details' table and checking:

Command: load data inpath '/user/hivecasestudy/2019-Oct.csv' into table sales_details; load data inpath '/user/hivecasestudy/2019-Nov.csv' into table sales_details;

Output:

```
hive> load data inpath '/user/hivecasestudy/2019-Oct.csv' into table sales_details;
Loading data to table default.sales_details
OK
Time taken: 1.511 seconds
hive> load data inpath '/user/hivecasestudy/2019-Nov.csv' into table sales_details;
Loading data to table default.sales_details
OK
Time taken: 0.842 seconds
```

Command: select * from sales_details > limit 5:

```
* from sales details
sales details.event time
                                      sales_details.event_type
                                                                                        sales_details.product
sales_details.category_id sales_details.category_code sales_details.b
cand sales_details.price sales_details.user_id sales_details.user_session
2019-11-01 00:00:02 UTC view 5802432 1487580009286598681 0.32 5
52076640 09fafd6c-6c99-46b1-834f-33527f4de241
62076640
2019-11-01 00:00:09 UTC cart 5844397 1487580006317032337
53329724 2067216c-31b5-455d-a1cc-af0575a34ffb
                                                                                                               22.22
                   57ed222e-a54a-4907-9944-5a875c2d7f4f
56138645
                                                                                                    jessnail
                                 186c1951-8052-4b37-adce-dd9644b1d5f7
019-11-01 00:00:24 UTC remove from cart
                                                            5826182 1487580007483048900
                                2067216c-31b5-455d-a1cc-af0575a34ffb
      taken: 0.632 seconds, Fetched: 5 row(s)
```

Hive Commands for easy view:

```
Commands: set hive.cli.print.header=True;

set hive.resultset.use.unique.column.names=false;

set hive.exec.dynamic.partition=true;

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

set hive.enforce.bucketing=true;
```

Output:

```
hive> set hive.cli.print.header=True;
hive> set hive.resultset.use.unique.column.names=false;
hive> set hive.exec.dynamic.partition=true;
hive> set hive.exec.dynamic.partition.mode=nonstrict;
hive> set hive.enforce.bucketing=true;
```

Explanation:

- set hive.cli.print.header=True; To display the column header in output
- set hive.resultset.use.unique.column.names=false; To remove table name before column name separated by '.'
- set hive.exec.dynamic.partition=true;
- set hive.exec.dynamic.partition.mode=nonstrict; To enable Dynamic partitioning
- set hive.enforce.bucketing=true; To enable bucketing

4. Optimization of Tables – Partitioning and Bucketing:

Partitioning and Bucketing are query optimisation techniques. Bucketing is a method of segregating the tables or partitions in hive into parts based on the values of a column which in turn reduces the query execution time.

Attribute 'event_type' has unique set of categorical values and it will give us a definite number of partitions (4). So, partitioning will be performed on the attribute 'event_type'. This will also reduce the load to the name node.

Bucketing will be performed on the attribute 'price'. Depending on the data size and the size of HDFS block, the number of buckets will be created. We have calculated to create a minimum of eight buckets.

Command: create external table cosmetics_par(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 6 buckets

> row format serde 'org.apache.hadoop.hive.serde2.OpenCSVSerde'

> stored as textfile location '/user/hivecasestudy/'

> tblproperties ("skip.header.line.count"="1");

Output:

```
hive> create external table cosmetics_par (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 (user_id) into 6 buckets

> row format serde 'org.apache.hadoop.hive.serde2.OpenCSVSerde'

> stored as textfile location '/user/hivecasestudy/'

> tblproperties ("skip.header.line.count"="1");

OK

Time taken: 0.307 seconds
```

Checking the new table cos_opt:

Command: desc cosmetics_par;

```
hive> desc cosmetics par;
OK
event time
                        string
                                                 from deserializer
product id
                        string
                                                 from deserializer
                                                 from deserializer
category id
                        string
category code
                                                 from deserializer
                        string
                                                 from deserializer
brand
                        string
price
                        string
                                                 from deserializer
user id
                                                 from deserializer
                        string
user session
                                                 from deserializer
                        string
event type
                        string
# Partition Information
# col name
                        data type
                                                 comment
                        string
event type
Time taken: 0.467 seconds, Fetched: 14 row(s)
```

Loading data into partitioned and bucketed table 'cosmetics par':

Command: insert into table cosmetics_par partition(event_type)

> select event_time, product_id, category_id, category_code, brand, price,
user_id, user_session, event_type

> from sales_details;

```
Reducer 2: 4(+3)/9
Map 1: 4/4
Map 1: 4/4
              Reducer 2: 4(+3)/9
Map 1: 4/4 Reducer 2: 4(+3)/9
             Reducer 2: 4(+3)/9
Map 1: 4/4
Map 1: 4/4
               Reducer 2: 5(+3)/9
               Reducer 2: 5(+3)/9
Map 1: 4/4
              Reducer 2: 5(+3)/9
Map 1: 4/4
Map 1: 4/4
              Reducer 2: 5(+3)/9
Map 1: 4/4
              Reducer 2: 5(+3)/9
Map 1: 4/4
Map 1: 4/4
              Reducer 2: 5(+3)/9
Map 1: 4/4
               Reducer 2: 5(+3)/9
               Reducer 2: 5(+3)/9
Map 1: 4/4
              Reducer 2: 6(+2)/9
Map 1: 4/4
Map 1: 4/4
              Reducer 2: 6(+3)/9
Map 1: 4/4
              Reducer 2: 6(+3)/9
Map 1: 4/4
             Reducer 2: 6(+3)/9
             Reducer 2: 6(+3)/9
Reducer 2: 7(+2)/9
Reducer 2: 7(+2)/9
Map 1: 4/4
Map 1: 4/4
Map 1: 4/4
             Reducer 2: 7(+2)/9
Map 1: 4/4
             Reducer 2: 7(+2)/9
Map 1: 4/4
Map 1: 4/4
             Reducer 2: 8(+1)/9
Map 1: 4/4 Reducer 2: 8(+1)/9
Map 1: 4/4 Reducer 2: 9/9
Map 1: 4/4
Loading data to table hive casestudy.cosmetics par partition (event type=null)
        Time taken to load dynamic partitions: 0.383 seconds
        Time taken for adding to write entity: 0.006 seconds
```

Verifying partitioning in Hive:

Command: *show partitions cosmetics_par*;

Output:

```
hive> show partitions cosmetics_par;

OK

event_type=cart

event_type=purchase

event_type=remove_from_cart

event_type=view

Time taken: 0.144 seconds, Fetched: 4 row(s)
```

Verifying partitioning and bucketing in Hadoop:

Command: *hadoop fs -ls /user/hivecasestudy/*

```
[hadoop@ip-172-31-83-157 ~]$ hadoop fs -ls /user/hivecasestudy/
Found 4 items
                                    0 2022-12-01 12:51 /user/hivecasestudy/event typ
drwxr-xr-x - hadoop hadoop
e=cart
            - hadoop hadoop
                                    0 2022-12-01 12:52 /user/hivecasestudy/event typ
drwxr-xr-x
e=purchase
                                    0 2022-12-01 12:52 /user/hivecasestudy/event typ

    hadoop hadoop

drwxr-xr-x
e=remove from cart
            - hadoop hadoop
                                    0 2022-12-01 12:52 /user/hivecasestudy/event typ
drwxr-xr-x
e=view
```

We can observe that four partitions have been created.

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

Output:

From the output, we can observe that 6 buckets have been created with index starting from 0-5.

Command: select * from cosmetics_par limit 5;

```
hive> select * from cosmetics par
   > limit 5;
2019-10-06 20:46:50 UTC 5853630 1487580006317032337
                                                                               557505
                                                                       11.10
     dfee5efd-eb4a-4d7c-8dc9-0d976ff1ead8 cart
2019-10-08 09:48:05 UTC 5885214 2069804417665728971
                                                               cosmoprofi
                                                                               5.56 5
              a7a775fa-c671-43ae-928d-667eab5e6e48
                                                       cart
2019-10-01 07:00:05 UTC 5847446 1487580009286598681
                                                                               555511
       d29c7b8f-041d-4cbb-acf3-67859a4cc5db
2019-10-08 09:48:03 UTC 5885022 2069804417665728971
                                                               cosmoprofi
                                                                               5.56 5
               f0cda670-7691-41d0-b7d8-c445c0e34926
2019-10-06 20:46:49 UTC 5853630 1487580006317032337
                                                                       11.10
                                                                               557505
       dfee5efd-eb4a-4d7c-8dc9-0d976ff1ead8
Time taken: 3.334 seconds, Fetched: 5 row(s)
```

Now, we have created two tables namely, sales_details and cosmetics_par. We will be using 'cosmetics_par' table for analysing the data which is a partitioned and bucketed table.

```
hive> show tables;
OK
cosmetics_par
sales_details
Time taken: 0.059 seconds, Fetched: 2 row(s)
```

5. Querying – Assignment Questions:

i. Find the total revenue generated due to purchases made in October.

```
Command: SELECT ROUND(SUM(price),2) AS total_sales

> FROM cosmetics_par

> WHERE event_type='purchase' AND

> month(event_time)=10;
```

Output

• The total revenue generated in the month of October is Rs.1211534.31

ii. Write a query to yield the total sum of purchases per month in a single output.

```
Command: SELECT Month(event_time) as Month, ROUND(SUM(price),2) as total_purchases

>FROM cosmetics_par

>WHERE event_type='purchase'

>GROUP BY month(event_time);
```

Output:

```
hive> SELECT month(event time) as month, ROUND(SUM(price),2) AS total purchases
    > FROM cosmetics par
    > WHERE event type='purchase'
    > GROUP BY month (event time);
Query ID = hadoop 20221201171315 af38aabd-820a-4376-90fb-eb057953f474
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application 1669909606668 0010)
         VERTICES
                          MODE
                                         STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

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

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

0K
month total purchases
         1211534.31
         1531007.31
Time taken: 26.668 seconds, Fetched: 2 row(s)
```

- In the month of October, the total purchases is Rs.1211534.31
- In the month of November, the total purchases is Rs.1531007.31

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

```
Command: SELECT ROUND((Rev_Nov - Rev_Oct),2) AS revenue_change
>FROM (
>SELECT SUM (CASE WHEN month(event_time)=10 THEN price ELSE 0
END) AS Rev_Oct,
>SUM (CASE WHEN month(event_time)=11 THEN price ELSE 0
```

```
END) AS Rev_Nov,

>FROM cosmetics_par,

>WHERE event_type='purchase' AND

>month(event_time) in (10,11)

> )s;
```

```
hive> SELECT ROUND((Rev Nov - Rev Oct),2) AS revenue change
   > FROM (
   > SELECT SUM(CASE when month(event time)=10 then price else 0 END) AS Rev Oct,
   > SUM(CASE when month(event time)=11 then price else 0 END) AS Rev Nov
   > FROM cosmetics par
   > WHERE event type='purchase' AND
   > month(event time) in (10,11)
Query ID = hadoop 20221201170815 2b6b05f1-dc12-4bb6-9e93-e1b3824a1316
Total jobs = 1
Launching Job 1 out of 1
Tez session was closed. Reopening...
Session re-established.
Status: Running (Executing on YARN cluster with App id application 1669909606668 0010)
                      MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
        VERTICES
Map 1 ..... container SUCCEEDED
Reducer 2 ..... container SUCCEEDED
OK
revenue change
319473.0
lime taken: 37.808 seconds, Fetched: 1 row(s)
```

• From the output, we can infer that the difference in revenue is Rs. 3,19,473.00

iv. Find distinct categories of products. Categories with null category code can be ignored.

Command: SELECT DISTINCT SPLIT (category_code, '\\.')[0] AS Distinct_category_list >FROM cosmetics_par;

```
hive> SELECT DISTINCT SPLIT(category code, '\\.')[0] AS Distinct category_list
   > FROM cosmetics par;
Query ID = hadoop 20221201171658 87fa2462-8b39-4361-ac9f-1fdf5851b5f7
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application 1669909606668 0010)
       VERTICES
                    MODE
                               STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
Map 1 ..... container SUCCEEDED
                                          6 6
Reducer 2 ..... container SUCCEEDED
VERTICES: 02/02 [====
                                 ======>>] 100% ELAPSED TIME: 74.57 s
distinct category list
furniture
appliances
accessories
apparel
sport
stationery
Time taken: 75.222 seconds, Fetched: 7 row(s)
```

- We can observe that there are 6 distinct categories namely, furniture, appliances, accessories, apparel, sport and stationery.
- v. Find the total number of products available under each category.

```
Command: SELECT SPLIT (category_code, '\\.')[0] AS category, COUNT(product_id) AS total_prod

>FROM cosmetics_par

>GROUP BY SPLIT(category_code, '\\.')[0]

>ORDER BY total_prod;
```

```
hive> SELECT SPLIT(category code,'\\.')[0] AS category, COUNT(product id) AS total prod
    > FROM cosmetics par
     > GROUP BY SPLIT(category code, '\\.')[0]
     > ORDER BY total prod;
Query ID = hadoop 20221201172308 d41edd06-4e6b-4b40-9b0c-5252c4285efd
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application 1669909606668 0010)
          VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

        Map 1 ......
        container
        SUCCEEDED
        6
        6
        0
        0
        0

        Reducer 2 .....
        container
        SUCCEEDED
        5
        5
        0
        0
        0

        Reducer 3 .....
        container
        SUCCEEDED
        1
        1
        0
        0
        0

category total_prod
sport 2
accessories 12929
apparel 18232
furniture 23604
stationery 26722
appliances 61736
          8594871
Time taken: 75.845 seconds, Fetched: 7 row(s)
```

The total number of products under each category are as follows:

• Sport : 2

• Accessories: 12929

• Apparel: 18232

• Furniture : 23604

• Stationery : 26722

• Appliances: 61736

vi. Which brand had the maximum sales in October and November combined?

```
Command: SELECT brand, ROUND(SUM(price),2) AS sales_total

>FROM cosmetics_par

>WHERE event_type= 'purchase'

>GROUP BY brand

>ORDER BY sales_total DESC

>LIMIT 5;
```

```
hive> SELECT brand,ROUND(SUM(price),2) as sales total
    > FROM cosmetics par
    > WHERE event_type='purchase'
    > GROUP BY brand
    > ORDER BY sales total DESC
    > LIMIT 5;
Query ID = hadoop 20221201173150 1986cdf5-d678-408e-b7b6-66e6c8ca9de5
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application 1669909606668 0010)
        VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
Map 1 ...... container SUCCEEDED 3 3 0

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

Reducer 3 container SUCCEEDED 1 1 0
Reducer 2 ..... container SUCCEEDED 1
Reducer 3 ..... container SUCCEEDED 1
OK
brand sales total
        1094179.77
runail 148297.94
grattol 106918.25
irisk 92538.0
        86341.78
Fime taken: 24.547 seconds, Fetched: 5 row(s)
```

• From the output, we can observe that 'Runail' is a popular brand with high sales in both months. The total sales is Rs.148297.94

vii. Which brands increased their sales from October to November?

```
Command: WITH sales_total AS (

>SELECT brand,

>SUM (CASE WHEN month(event_time)=10 THEN price ELSE 0 END) AS oct_sales,

>SUM (CASE WHEN month(event_time)=11 THEN price ELSE 0 END) AS nov_sales

>FROM cosmetics_par

>WHERE event_type = 'purchase' AND month(event_time) in (10,11)

>GROUP BY brand )

>SELECT brand, ROUND(oct_sales,2), ROUND(nov_sales,2), ROUND((nov_sales-oct_sales),2) AS increased_sales

>FROM sales_total

>WHERE (nov_sales-oct_sales)>0

>ORDER BY increased_sales DESC;
```

```
hive> WITH sales total AS (
    > SELECT brand,
    > SUM (CASE WHEN month (event time) = 10 THEN price ELSE 0 END) AS oct sales,
    > SUM (CASE WHEN month (event time) = 11 THEN price ELSE 0 END) AS nov sales
    > FROM cosmetics par
    > WHERE event type='purchase' AND
    > month(event time) in (10,11)
    > GROUP BY brand
    > SELECT brand, ROUND(oct sales,2), ROUND(nov sales,2), ROUND((nov sales-oct sales),2) AS increased
sales
    > FROM sales total
    > WHERE (nov sales-oct sales)>0
   > ORDER BY increased sales DESC;
Query ID = hadoop 20221201175202 0fb2b8e6-a47c-4960-8afa-0183562c6b88
Total jobs = 1
Launching Job 1 out of 1
Tez session was closed. Reopening...
Session re-established.
Status: Running (Executing on YARN cluster with App id application 1669909606668 0011)
         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

        Reducer 3
        Container
        SUCCEEDED
        1
        1
        0
        0
        0
        0

brand c1 c2 increased_sales

474677.31 619502.46 144825.15

grattol 35445.54 71472.71 36027.17

uno 35302.03 51039.75 15737.72
lianail 5892.84 16394.24 10501.4
ingarden 23161.39 33566.21 10404.82
strong 29196.63 38671.27 9474.64
jessnail 26287.84 33345.23 7057.39 cosmoprofi 8322.81 14536.99 6214.18
polarus 6013.72 11371.93 5358.21
runail 71539.28 76758.66 5219.38
freedecor 3421.78 7671.8 4250.02
staleks 8519.73 11875.61 3355.88
bpw.style 11572.15 14836.36 3264.21
```

```
staleks 8519.73 11875.61
                                3355.88
bpw.style
                11572.15
                                14836.36
                                                3264.21
lovely 8704.38 11939.06
                                3234.68
                7280.75 10273.1 2992.35
marathon
                9390.69 12352.91
                                        2962.22
haruyama
yoko 8756.91 11707.88
                            2950.97
italwax 21940.24
                        24799.37
                                        2859.13
benovy 409.62 3259.97 2850.35
kaypro 881.34 3268.7 2387.36
estel 21756.75
                        24142.67
                                        2385.92
concept 11032.14
                        13380.4 2348.26
kapous 11927.16
                        14093.08
                                        2165.92
f.o.x
        6624.23 8577.28 1953.05
masura 31263.71
                        33056.74
                                        1793.03
milv
        3904.94 5642.01 1737.07
beautix 10493.95
                        12222.95
                                        1729.0
       2730.64 4327.25 1596.61
artex
                      12009.17
domix 10472.05
                                        1537.12
shik
        3341.2 4839.72 1498.52
smart 4457.26 5902.14 1444.88
roubloff
              3491.36 4913.77 1422.41
levrana 2243.56 3664.1 1420.54
oniq 8425.41 9841.65 1416.24
      45591.96
irisk
                        46946.04
                                        1354.08
               4775.88 6120.48 1344.6
severina
joico 705.52 2015.1 1309.58 zeitun 708.66 2009.63 1300.97
             554.17 1782.86 1228.69
beauty-free
                1887.93 3043.16 1155.23
swarovski
de.lux 1659.7 2775.51 1115.81
metzger 5373.45 6457.16 1083.71
markell 1768.75 2834.43 1065.68
sanoto 157.14 1209.68 1052.54
                4369.74 5327.68 957.94
nagaraku
ecolab 262.85 1214.3 951.45
art-visage 2092.71 2997.8 905.09
levissime
missha 1293.83 2150.28 856.45
               1899.7 2685.8 786.1
solomeya
rosi 3077.04 3841.56 764.52
refectocil
               2716.18 3475.58 759.4
kaaral 4412.43 5086.07 673.64
                1181.44 1813.37 631.93
kosmekka
                6334.25 6945.26 611.01
kinetics
browxenna
                14331.37
                                14916.73
                                                585.36
airnails
               5118.9 5691.52 572.62
```

```
1181.44 1813.37 631.93
kosmekka
kinetics
               6334.25 6945.26 611.01
               14331.37
                             14916.73
                                              585.36
browxenna
airnails
               5118.9 5691.52 572.62
uskusi 5142.27 5690.31 548.04
coifin 903.0 1428.49 525.49
s.care 412.68 913.07 500.39
limoni 1308.9 1796.6 487.7
matrix 3243.25 3726.74 483.49
gehwol 1089.07 1557.68 468.61
greymy 29.21 489.49 460.28
bioagua 942.89 1398.12 455.23
               837.37 1291.97 454.6
farmavita
sophin 1067.86 1515.52 447.66
yu-r
      271.41 673.71 402.3
       421.55 817.33 395.78
            389.0 389.0
naomi 0.0
lador 2083.61 2471.53 387.92
ellips 245.85 606.04 360.19
       3318.96 3657.43 338.47
lowence 242.84 567.75 324.91
nitrile 847.28 1162.68 315.4
shary 871.96 1176.49 304.53
kims 330.04 632.04 302.0
happyfons 801.92 1091.59 289.67
happyfons
               310.85 594.93 284.08
insight 1443.7 1721.96 278.26
candy 534.96 799.38 264.42
bluesky 10307.24
                       10565.53
                                      258.29
             511.51 768.35 256.84
beauugreen
protokeratin
               201.25 456.79 255.54
trind 298.07 542.96 244.89
entity 479.71 719.26 239.55
               651.94 890.45 238.51
skinlite
provoc 827.99 1063.82 235.83
fedua 52.38 263.81 211.43
               41.16 241.95 200.79
ecocraft
keen
       236.35 435.62 199.27
mane
       66.79
               260.26 193.47
               318.7 502.34 183.64
freshbubble
matreshka
               0.0
                      182.67 182.67
       358.94 538.61 179.67
cristalinas
               427.63 584.95 157.32
farmona 1692.46 1843.43 150.97
```

```
cristalinas
               427.63 584.95 157.32
farmona 1692.46 1843.43 150.97
               249.52 384.59 135.07
latinoil
miskin 158.04
              293.07 135.03
elizavecca
               70.53
                       204.3 133.77
nefertiti
               233.52 366.64 133.12
finish 98.38
               230.38 132.0
               513.66 645.07 131.41
igrobeauty
dizao
       819.13 945.51 126.38
       645.58 762.31 116.73
osmo
batiste 772.4
               874.17 101.77
carmex 145.08 243.36 98.28
       54.34
               152.61 98.27
depilflax
               2707.07 2803.78 96.71
               136.57 95.22
enjoy
       41.35
kerasys 430.91 525.2
                      94.29
aura
       83.95
               177.51 93.56
plazan 101.37 194.01 92.64
       422.73 507.29 84.56
koelf
nirvel 163.04
              234.33 71.29
konad
       739.83 810.67 70.84
egomania
               77.47
                      146.04 68.57
cutrin 299.37 367.62 68.25
laboratorium
               246.5
                      312.52 66.02
       288.02 351.21 63.19
inm
dewal
       0.0
               61.29
                       61.29
marutaka-foot
               49.22
                      109.33 60.11
kares
       0.0
               59.45
                       59.45
profhenna
               679.23 736.85 57.62
koelcia 55.5
               112.75 57.25
balbcare
               155.33 212.38 57.05
elskin 251.09
              307.65 56.56
foamie 35.04
               80.49
                      45.45
ladykin 125.65
              170.57 44.92
               340.97 44.91
likato 296.06
mavala 409.04 446.32
                      37.28
vilenta 197.6
               231.21 33.61
beautyblender
               78.74
                      109.41 30.67
biore
               90.31
                       29.66
       60.65
       902.38 931.09 28.71
orly
```

```
ladykin 125.65 170.57
                        44.92
likato 296.06
                340.97
                        44.91
mavala 409.04
                446.32
                        37.28
vilenta 197.6
                231.21
                        33.61
beautyblender
                78.74
                        109.41
                                 30.67
biore
        60.65
                90.31
                        29.66
orly
        902.38
                931.09
                        28.71
estelare
                444.81
                        471.87
                                 27.06
profepil
                93.36
                        118.02
                                 24.66
blixz
        38.95
                63.4
                        24.45
binacil 0.0
                24.26
                        24.26
godefroy
                401.22
                        425.12
                                 23.9
glysolid
                69.73
                        91.59
                                 21.86
                50.11
                        71.21
                                 21.1
veraclara
juno
        0.0
                21.08
                        21.08
                81.49
kamill 63.01
                        18.48
                163.37
                        181.49
treaclemoon
                                 18.12
                50.37
                        66.51
                                 16.14
supertan
                12.39
                        12.39
barbie 0.0
                316.84
                        329.17
deoproce
                                 12.33
rasyan 18.8
                28.94
                        10.14
fly
        17.14
                27.17
                        10.03
        236.16
                245.8
                        9.64
tertio
jaguar 1102.11 1110.65 8.54
        204.2
                212.53 8.33
soleo
neoleor 43.41
                51.7
                        8.29
        5.71
                10.28
                        4.57
moyou
bodyton 1376.34 1380.64 4.3
skinity 8.88
                12.44
                        3.56
helloganic
                0.0
                        3.1
                                 3.1
        100.92
                        1.69
                102.61
grace
        20.23
                20.93
cosima
                        0.7
        2.54
                3.1
                        0.56
ovale
Time taken: 45.29 seconds, Fetched: 161 row(s)
```

- From the output, we can observe that 161 brands increased their sales from the month of October to November.
- 'Grattol' brand has the highest per month increase of Rs. 36027.17
- 'Ovale' brand has the lowest per month increase of Rs. 0.56
- viii. Your company wants to reward the top 10 users of its website with a Golden Customer plan. Write a query to generate a list of top 10 users who spend the most.

```
Command: SELECT user_id, ROUND(SUM(price),2) AS spent_money
>FROM cosmetics_par
>WHERE event_type= 'purchase'
>GROUP BY user_id
>ORDER BY spent_money DESC
>LIMIT 10;
```

```
hive> SELECT user id, ROUND(SUM(price),2) AS spent money
     > FROM cosmetics par
     > WHERE event type='purchase'
     > GROUP BY user id
     > ORDER BY spent money DESC
     > LIMIT 10;
Query ID = hadoop 20221201173522 9d4770b6-1149-458a-a367-16c45b53c2f3
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application 1669909606668 0010)
        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

      Reducer 3 ..... container
      SUCCEEDED
      1
      1
      0
      0
      0
      0

 VERTICES: 03/03 [=============>>] 100% ELAPSED TIME: 27.07 s
user id spent money
557790271 2715.87
150318419 1645.97
562167663 1352.85
531900924 1329.45
557850743 1295.48
522130011
                   1185.39
561592095
                   1109.7
431950134
                   1097.59
                     1056.36
566576008
521347209
                     1040.91
Time taken: 27.757 seconds, Fetched: 10 row(s)
```

The top 10 users of the website are as follows:

557790271: 2715.87
150318419: 1645.97
562167663: 1352.85
531900924: 1329:45

• 557850743 : 1295.48

• 522130011:1185.39

• 561592095 : 1109.7

• 431950134 : 1097.59

• 566576008 : 1056.36

• 521347209 : 1040.91

6. Improvement of performance after using optimization on tables

The execution time of query are reduced by using Hive optimization techniques. We had created an optimized table "cosmetics_par" through partitioning and bucketing in the start of the case study.

We will now observe the importance of optimization techniques, by querying two questions using the table "sales_details" and "cosmetics_par". We have selected Question 6 & Question 8 for the demonstration.

Q1. Write a query to yield the total sum of purchases per month in a single output.

Command: SELECT Month(event_time) as Month, ROUND(SUM(price),2) as total_purchases

- >FROM sales_details
- >WHERE event_type='purchase'
- >GROUP BY month(event_time);

Command: SELECT Month(event_time) as Month, ROUND(SUM(price),2) as total_purchases

- >FROM cosmetics_par
- >WHERE event type='purchase'
- >GROUP BY month(event_time); (Optimised querying)

```
hive> SELECT month(event time) as month, ROUND(SUM(price),2) AS total purchases
    > FROM cosmetics par
    > WHERE event type='purchase'
    > GROUP BY month (event time);
Query ID = hadoop 20221201171315 af38aabd-820a-4376-90fb-eb057953f474
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application 1669909606668 0010)
         VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

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

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

 ERTICES: 02/02 [======
month total purchases
        1211534.31
11
         1531007.31
Time taken: 26.668 seconds, Fetched: 2 row(s)
```

- We can observe that the results are similar for both queries but the query execution time differs.
- We see the query execution time for non optimized query is 60.192 seconds. We observed that, the query execution time for optimized query is 26.668 seconds. This is a huge difference, as the data size increase the increase in time will be substantially higher.

Q2. Your company wants to reward the top 10 users of its website with a Golden Customer plan. Write a query to generate a list of top 10 users who spend the most.

```
Command: SELECT user_id, ROUND(SUM(price),2) AS spent_money
>FROM sales_details
>WHERE event_type= 'purchase'
>GROUP BY user_id
>ORDER BY spent_money DESC
>LIMIT 10;
```

```
hive> SELECT user_id, ROUND(SUM(price),2) AS spent_money
   > FROM sales details
   > WHERE event_type='purchase'
   > GROUP BY user id
   > ORDER BY spent money DESC
Query ID = hadoop 20221201175852 2da3a69c-bd1a-4902-a346-4d77986518dc
Total jobs = 1
Launching Job 1 out of 1
Tez session was closed. Reopening...
Session re-established.
Status: Running (Executing on YARN cluster with App id application 1669909606668 0012)
                     MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
        VERTICES
Map 1 ..... container SUCCEEDED
Reducer 2 .... container SUCCEEDED
Reducer 3 .... container SUCCEEDED
                                         =====>>] 100% ELAPSED TIME: 62.68 s
user_id spent_money
557790271
                2715.87
150318419
562167663
                1352.85
531900924
557850743
                1295.48
561592095
431950134
                1097.59
566576008
521347209
                1040.91
 ime taken: 71.187 seconds, Fetched: 10 row(s)
```

```
Command: SELECT user_id, ROUND(SUM(price),2) AS spent_money
>FROM cosmetics_par
>WHERE event_type= 'purchase'
>GROUP BY user_id
>ORDER BY spent_money DESC
>LIMIT 10; (Optimized query)
```

```
hive> SELECT user id, ROUND(SUM(price),2) AS spent money
    > FROM cosmetics par
    > WHERE event type='purchase'
    > GROUP BY user id
    > ORDER BY spent money DESC
    > LIMIT 10;
Query ID = hadoop 20221201173522 9d4770b6-1149-458a-a367-16c45b53c2f3
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application 1669909606668 0010)
         VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

        Map 1
        Container
        SUCCEEDED
        3
        3
        0
        0
        0

        Reducer 2
        Container
        SUCCEEDED
        1
        1
        0
        0
        0

        Reducer 3
        Container
        SUCCEEDED
        1
        1
        0
        0
        0

 user id spent money
557790271 2715.87
150318419
                 1645.97
562167663
                 1352.85
                 1329.45
531900924
557850743
                 1295.48
522130011
                 1185.39
561592095
                  1109.7
431950134
                  1097.59
566576008
                  1056.36
521347209
                  1040.91
Fime taken: 27.757 seconds, Fetched: 10 row(s)
```

- We can observe that the results are similar for both queries but the query execution time differs.
- We see the query execution time for non optimized query is 71.187 seconds. We observed that, the query execution time for optimized query is 27.757 seconds.

This is a huge difference, as the data size increase the increase in time will be substantially higher.

 Thus, partitioning and bucketing plays a crucial role to reduce the query execution time.

7. Cleaning-up:

Once we are done with the analysis, we can drop the databases and quit hive and then terminate the EMR Cluster.

```
hive> show databases;

OK

default

hive_casestudy

Time taken: 0.03 seconds, Fetched: 2 row(s)

hive> drop database hive_casestudy CASCADE;

OK

Time taken: 0.356 seconds

hive> show databases;

OK

default

Time taken: 0.017 seconds, Fetched: 1 row(s)
```

