# **HIVE CASE STUDY** Click stream data analysis Submitted by: Dyutimaya Das Steps:

# ☆ Starting an EMR Cluster

- Select region as N. Virginia(us-east-1c).
- Create a key pair
- Go to AWS account and Create an EMR cluster with 1 master node and 1 core node having m4. large instance type.
  - ★ Select the same key pair as created before.
- Enable SSH by editing an inbound rule and add SSH as port 22 to the rule.
- Open the putty terminal add the IP address of hostname and Put the .PPK file in the Auth section

```
Using username "hadoop".
🚜 Authenticating with public key "demo keypair"
Last login: Sun Apr 25 07:08:20 2021
                 Amazon Linux 2 AMI
https://aws.amazon.com/amazon-linux-2/
2 package(s) needed for security, out of 27 available
Run "sudo yum update" to apply all updates.
EEEEEEEEEEEEEEEEEE MMMMMMMM
                                 M::::::: M R::::::::::::R
EE:::::EEEEEEEEE:::E M:::::::M
                               M:::::::M R:::::RRRRRR:::::R
            EEEEE M:::::::M
                              E::::E
 E::::E
                 M::::::M:::M
                             M:::M:::::M R:::R
                                                   R::::R
 E:::::EEEEEEEEE M:::::M M:::M M::::M R:::RRRRRR:::::R
 E::::EEEEEEEEE
                 M:::::M
                         M:::::M
                                  M:::::M
                                          R:::RRRRRR::::R
 E::::E
                 M:::::M
                          M:::M
                                  M:::::M R:::R
 E::::E EEEEE M:::::M
                           MMM
                                  M:::::M R:::R
                                                   R::::R
EE:::::EEEEEEEE::::E M:::::M
                                          R:::R
                                                   R::::R
M:::::M RR::::R
                                                   R::::R
EEEEEEEEEEEEEEEEEE MMMMMMM
                                  MMMMMMM RRRRRRR
                                                   RRRRRR
```

# Hadoop Commands Using Putty Terminal

We have clickstream data from 2 months in form of 2 .csv files. Thus, we will create a directory in HDFS first.

Step 1: Create a directory in HDFS in to which we can copy the dataset which is present in the S3 bucket. By using the command

hadoop fs -mkdir /user/hive/casestudy

```
[hadoop@ip-172-31-74-162 ~]$ hadoop fs -mkdir /user/hive/casestudy
```

Step 2: Copy the data present in S3 bucket to the HDFS by using this command

hadoop distcp 's3://e-commerce-events-ml/\* /user/hive/casestudy/

Step 3: Check whether the data set has been moved to HDFS or not.

```
[hadoop@ip-172-31-74-162 ~]$ hadoop fs -ls /user/hive/casestudy/
Found 2 items
-rw-r--r- 1 hadoop hdfsadmingroup 545839412 2021-04-25 07:13 /user/hive/casestudy/2019-Nov.csv
-rw-r--r- 1 hadoop hdfsadmingroup 482542278 2021-04-25 07:13 /user/hive/casestudy/2019-Oct.csv
```

# 

Step 1: Launch the Hive CLI using the command

Hive

```
[hadoop@ip-172-31-65-148 ~]$ hive

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

Step 2: Create a data base for storing our tables metadata by using this command

create database if not exists sales;

```
hive> create database if not exists sales;
OK
Time taken: 0.595 seconds
```

Step 3: Check the created database by using the command

Show databases;

```
hive> show databases;
OK
default
sales
Time taken: 0.229 seconds, Fetched: 2 row(s)
hive>
```

Salles database has been created perfectly.

Type *use sales;* to start working on this database.

```
hive> use sales;
OK
Time taken: 0.037 seconds
```

#### step 4: Create a table to load all the data

- Create an Internal table here. We don't need to run any other applications on the top of this table so its better to create an internal table rather than an external table.
- Here we used OpenCSVSerde library to read .csv file
- We don't want to contain first line of csv file into the table because it contains the header file so we skip the line count as 1.

```
CREATE TABLE IF NOT EXISTS sales_info (event_time TIMESTAMP, event_type STRING, product_id STRING, category_id STRING, category_code STRING, brand STRING, price

FLOAT, user_id BIGINT, user_session STRIN) ROW FORMAT SERDE

'org.apache.hadoop.hive.serde2.OpenCSVSerde' STORED AS TEXTFILE

LOCATION
```

'/user/hive/casestudy/' TBLPROPERTIES ("skip.header.line.count"="1");

hive> create table if not exists sales\_info (event\_time timestamp,event\_type string,product\_id string,category\_id string,category\_code string,brand string,price float,user\_id bigint ,user\_s ession string) row format SERDE 'org.apache.hadoop.hive.serde2.OpenCSVSerde' stored as textfile location '/user/hive/casestudy/' tblproperties ("skip.header.line.count"="1");

OK
Time taken: 0.089 seconds

Step 5: Check the table creation by using command

show tables;

```
hive> show tables;
OK
sales_info
Time taken: 0.044 seconds, Fetched: 1 row(s)
```

The table sales\_info is created.

Step 6: Check data in table. select

\* from sales info limit 3;

Took 2.246 s to fetch first 3 records from the non-optimized table sales\_info.

Step 7: Now we shall create our Optimized table.

- For creating Dynamic partitioned table, we need to first set the command as Set hive.exec.dynamic.partition=true; Set hive.exec.dynamic.partition.mode=nonstrict;
- We will be using even\_type as our field for partitioning. It contains 4 labels and most of our queries will be around purchases and 'purchase' is an event type. So, it's a better choice to use partitioning on event\_type.
- We will be using bucketing and clustering simultaneously on price and will divide it into 50 buckets.
- This table is an internal table as all the data of previous table need to insert on the same table.

CREATE TABLE IF NOT EXISTS opt\_sales\_info (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 50 BUCKETS ROW FORMAT SERDE 'org.apache.hadoop.hive.serde2.OpenCSVSerde' STORED AS TEXTFILE LOCATION '/user/hive/casestudy/" TBLPROPERTIES ("skip.header.line.count"="1");

```
nive> set hive.exec.dynamic.partition=true;
nive> set hive.exec.dynamic.partition.mode=nonstrict;
nive> create table if not exists opt_sales_info (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 50 buckets row format SERDE 'org.apache.hadoop.hive.serde2.OpenCSVSerde' stored as textfile location '/user/hive/casestudy/' t
plproperties ("skip.header.line.count"="1");
```

Step 8: Now check the table creation by command

Show tables:

```
hive> show tables;
OK
tab_name
opt_sales_info
sales_info
Time taken: 0.038 seconds, Fetched: 2 row(s)
hive>
```

Both the tables are created base table (sales-info) and optimized table (opt\_sales\_info)

Step 9: Insert data into the optimize table by using the command.

```
INSERT INTO table opt_sales_info
PARTITION (event_type)
SELECT event time,
product_id, category_id,
category_code, brand,
price, user_id,
user_session,
event_type FROM
sales info;
```

Step 10: Exit from the hive terminal and check the buckets and partitions in hdfs. hadoop

fs -ls '/user/hive/casestudy/'

As we can see 4 partitions have been dynamically created on the field event\_type in the form of directories with name format 'event\_type=value'

Step 11: Check buckets of any one partition: hadoop fs -ls '/user/hive/casestudy/event\_type=cart'

```
hadoop@ip-172-31-65-148 ~]$ hadoop fs -ls /user/hive/casestudy/event_type=cart
                                                      2054764 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000000 3705883 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000001
rwxr-xr-x
rwxr-xr-x
                   hadoop hdfsadmingroup
                                                    12579390 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000002
                                                     3617299 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000003 4426146 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000004
                   hadoop hdfsadmingroup
                                                      7573931 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000005
7350044 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000006
rwxr-xr-x
                  hadoop hdfsadmingroup
                                                      3028836 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000007
                1 hadoop hdfsadmingroup
rwxr-xr-x
                                                     7028319 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000008 6282173 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000009
rwxr-xr-x
                  hadoop hdfsadmingroup
                1 hadoop hdfsadmingroup
rwxr-xr-x
                                                     4240684 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000010
rwxr-xr-x
                1 hadoop hdfsadmingroup
                                                     4822418 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000011
rwxr-xr-x
                                                      3002275 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000012
rwxr-xr-x
                1 hadoop hdfsadmingroup
                                                     2750488 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000013
rwxr-xr-x
                1 hadoop hdfsadmingroup
                                                    2730488 2021-03-02 09:07 /user/hive/casestudy/event_type=cart/000014
11330763 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000015
3202119 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000016
5746591 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000017
rwxr-xr-x
                1 hadoop hdfsadmingroup
rwxr-xr-x
rwxr-xr-x
                  hadoop hdfsadmingroup
rwxr-xr-x
                1 hadoop hdfsadmingroup
                                                     2559685 2021-05-02 09:07 /user/hive/casestudy/event type=cart/000018
rwxr-xr-x
                1 hadoop hdfsadmingroup
                                                    10987168 2021-05-02 09:07 /user/hive/casestudy/event type=cart/000019
rwxr-xr-x
                                                     1096/168 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000012

9592355 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000023

3241246 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000022

5565655 2021-05-02 09:08 /user/hive/casestudy/event_type=cart/000023
                1 hadoop hdfsadmingroup
rwxr-xr-x
                  hadoop hdfsadmingroup
rwxr-xr-x
                   hadoop hdfsadmingroup
rwxr-xr-x
                   hadoop hdfsadmingroup
                                                      9369608 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000024
                   hadoop hdfsadmingroup
rwxr-xr-x
                                                      3486334 2021-05-02 09:07 /user/hive/casestudy/event type=cart/000025
                   hadoop hdfsadmingroup
                                                      8442021 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000026
3774273 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000027
                   hadoop hdfsadmingroup
                   hadoop hdfsadmingroup
                                                    10254871 2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000028
                   hadoop hdfsadmingroup
                                                                2021-05-02 09:07 /user/hive/casestudy/event_type=cart/000029
```

50 buckets have been created on the 'price' field for each partition.

## Step 12: Check data in table:

SELECT \* FROM opt info sales LIMIT 3;

```
hive> select * from opt_sales_info limit 3 ;

OK

2019-10-14 14:29:52 UTC 5622687 1487580007281722301 severina 1.81 549308589 47f4f19c-7fed-4a07-9b1c-1b49ad78ce13 cart
2019-10-13 16:28:31 UTC 5881733 1842735760499802745 kinetics 8.57 494832412 1f14b28e-18ab-4417-afc0-d241ea059329 cart
2019-10-12 16:46:23 UTC 5668346 1487580007701152718 blixz 1.81 70197834 06ea5f35-6576-4a79-9bee-e15d352f44f3 cart
Time taken: 0.248 seconds, Fetched: 3 row(s)
hive>
```

Took 0.248 s to fetch first 3 records from the non-optimized table sales\_info whereas it took 2.246 s to fetch first 3 record from non-optimised table.

# ☆ Running Hive queries

Query 1: Find the total revenue generated due to purchase made in October.

We need to compare the performance of the non-optimized table and the optimized table.

Query on non-optimized table:

```
Select sum(price) as oct_revenue from sales_info where month(event time)=10 and event type='purchase'; Screenshot:
```

#### Result:

The total revenue generated due to purchases made in October is: 1211538.4299.

Here, the Map Reduce job has to run through the entire table to filter out purchase records from October and hence, the time taken is on the higher side i.e. 125.985 seconds

#### Query on optimized table:

Select sum(price) as oct\_revenue from opt\_sales\_info where month (event\_time) =10 and event\_type='purchase';

# Screenshot:

```
hive> select sum(price) as oct_revenue from opt_sales_info where month(event_time)=10 and event_type='purchase';
Query ID = hadoop_20210425080207_72c331cb-ae4e-45cc-97b8-5167146e7482
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application_1619331103777_0010)

VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

Map 1 ...... container SUCCEEDED 6 6 6 0 0 0 0 0
Reducer 2 .... container SUCCEEDED 1 1 0 0 0 0
VERTICES: 02/02 [===========>>] 100% ELAPSED TIME: 23.57 s

OK
oct_revenue
1211320.6099998376
Time taken: 24.789 seconds, Fetched: 1 row(s)
```

#### Result:

Here we can see that the bucketing on 'price' field and the partitioning on the 'event\_type' field helped speed up the querying by almost 5 folds. (24s vs 126s)

NOTE: As the performance on the optimized table (opt\_sales-info) is much better, we will be running rest of our queries on this table itself.

# Query 2: Write a query to yield the total sum of purchases per month in a single output.

# Query:

select month(event\_time), sum(price) as total\_revenue
from opt\_sales\_info where event\_type='purchase'
group by month(event\_time);

## Screenshot:

#### Results:

Here we can see that the total sum of purchases for October is 1211320.60 and November is 1530861.029.

We have used GROUP BY here to get the sums of both months in a single output. It can also be seen that the partitioning and bucketing have again helped to keep the query execution time just under 25s.

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

### Query:

With revenue table as (

```
Select sum (case when month(event_time) = 10

then price
else 0 end) as oct_total_sale,

sum (case when month(event_time) = 11

then price
else 0 end) as nov_total_sale

from opt_sales_info where event_type =
'purchase')

select nov_total_sale - oct_total_sale as change_in_revenue from revenue_table;
```

# Screenshot:

```
hive> with revenue table as (
    > select sum( case when month(event time) = 10
                        then price
                      else 0 end) as oct total sale,
            sum ( case when month (event time) = 11
                           then price
                       else 0 end) as nov total sale
   > from opt_sales_info where event_type='purchase')
> select nov_total_sale-oct_total_sale as change_in_revenue from revenue_table;
Query ID = hadoop 20210425162208 bf946f64-b4c4-48c2-b8fa-e8fd2ba5578e
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 1619365146236 0004)
        VERTICES
                      MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
Map 1 ..... container SUCCEEDED
Reducer 2 ..... container
                              SUCCEEDED
VERTICES: 02/02 [===============>>] 100% ELAPSED TIME: 25.06 s
change in revenue
319540.4200000339
Time taken: 33.432 seconds, Fetched: 1 row(s)
```

# Results:

In this query, firstly we have used a common table expression which contains 2 conditional sum variables, one is sum of the prices only when month is October and the other is sum of the prices only when its November. This is implemented using CASE. Once we have these 2 values, we are then going to query this CTE and fetch the difference in the 2 calculated sums. Prices are summed only when 'event\_type' is a purchase.

It can be seen that purchases in Nov have increased by an amount of 319540.42 over October.

Query 4: Find the distinct categories of products. Categories with null category can be ignored.

### Query by using distinct function:

```
Select distinct category_code as category_code from opt_sales_info where category_code != ";
```

#### Screenshot:

```
nive> select distinct category_code as category_code from opt_sales_info where category_code != '';
Query ID = hadoop 20210425082127 6153293d-9f48-4a08-ad8d-b3c81f2d246f
Fotal 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 1619331103777 0012)
                                  STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
       VERTICES
Map 1 ..... container SUCCEEDED
Reducer 2 ..... container SUCCEEDED
ERTICES: 02/02 [=====
                                 ======>>] 100% ELAPSED TIME: 58.03 s
category code
accessories.bag
accessories.cosmetic bag
apparel.glove
appliances.environment.air conditioner
appliances.environment.vacuum
appliances.personal.hair cutter
urniture.bathroom.bath
urniture.living room.cabinet
furniture.living room.chair
sport.diving
stationery.cartrige
Fime taken: 66.902 seconds, Fetched: 11 row(s)
```

#### Result:

From the query, it can be seen that we used DISTINCT function to fetch distinct categories of the products. It took 66.9 s.

Here we got 11 distinct product categories as result.

# Query 5: Find the total number of products available under each category.

### Query:

```
select category_code, count(product_id) as product_count
from opt_sales_info where category_code != "
group by category_code;
```

#### Screenshot:

```
nive> select category_code, count(product id)as product count from op
Query ID = hadoop_20210425082706_1b424f94-6393-4196-98c5-0b1a2a81e388
                                                               _count from opt_sales_info where category_code != "" group by category_code;
Total jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application_1619331103777_0012)
        VERTICES
                        MODE
                                      STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
Map 1 ..... container Reducer 2 ..... container
ategory_code product_count
accessories.bag 11680
appliances.environment.air_conditioner 332
appliances.environment.vacuum 59759
appliances.personal.hair cutter 1642
furniture.bathroom.bath 9857
furniture.living_room.chair
sport.diving \overline{2}
 tationery.cartrige
 ime taken: 57.308 seconds, Fetched: 11 row(s)
```

#### Result:

Here in this query, we used the COUNT function for counting the product\_ids under each category\_code here.

# Query 6: Which brand had the maximum sales in October and November combined.

#### Query:

```
select brand, sum (price) as total_sales
from opt_sales_info
where event_type= 'purchase' and brand != "
group by brand order by total_sales desc
limit 1;
```

## **Screenshot:**

#### Result:

Runail has the most sales in both months combined. Sales value is 148258.94

# Query 7: Which brands increased their sales from October to November

# Query:

# Screenshot:

```
hive> with brand_sales_info 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
                           else 0 end) as nov_sales
   > from opt_sales_info where event_type='purchase' group by brand)
   > select brand from brand_sales_info where nov_sales > oct_sales;
Query ID = hadoop_20210425163\overline{143}_9d\overline{81}ca3b-32d5-478\overline{c}-9d26-9d638\overline{02}12965
Launching Job 1 out of 1
Tez session was closed. Reopening...
Session re-established.
Status: Running (Executing on YARN cluster with App id application_1619365146236_0005)
       VERTICES MODE
                           STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
/ERTICES: 02/02 [===========>>] 100% ELAPSED TIME: 26.80 s
brand
art-visage
artex
barbie
batiste
peautix
```

```
igrobeauty
ingarden
jessnail
kapous
kerasys
kims
kinetics
kocostar
koelcia
koelf
kosmekka
lador
ladykin
latinoil
levrana
marutaka-foot
matreshka
metzger
neoleor
oniq
polarus
profepil
rasyan
refectocil
roubloff
s.care
severina
shary
skinity
solomeya
staleks
swarovski
tertio
veraclara
vilenta
yu-r
Time taken: 34.711 seconds, Fetched: 161 row(s)
```

# Result:

In this query, firstly we are using a common table expression which contains 2 conditional sum variables, one is the sum of prices only when month is October and the other sum of prices only when its November. This is implemented using CASE. Once we have these 2 values FOR EACH BRAND, we are then going to query this cpmmon table expression and fetch those brands whose NOV\_SALES value is greater than their OCT\_SALES value. Prices are summed only when 'event\_type' is a purchase.

Here, we can see that total 161 brands increased their revenue from Oct to Nov

Query 8: 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.

# Query:

```
select user_id, sum(price) as total_spend
from opt_sales-info where event_type =
'purchase' group by user_id order by
total_spend desc
limit 10;
```

# Screenshot:

```
hive> select user_id, sum(price)as total_spend from opt_sales info where event_type='purchase' group by user_id order by total_spend desc limit 10;

Query ID = hadoop_20210425083938_fladea3b=e6cf-4959-a564-86427c28a31f
Total_jobs = 1
Launching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application_1619331103777_0012)

VERTICES MODE STATUS TOTAL COMPLETED RUNNING FENDING FAILED KILLED

Map 1 ...... container SUCCEEDED 6 6 6 0 0 0 0
Reducer 2 .... container SUCCEEDED 2 2 0 0 0 0 0
Reducer 3 .... container SUCCEEDED 1 1 0 0 0 0 0

VERTICES: 03/03 [========>>>] 100% ELAPSED TIME: 23.62 s

DK

user_id total_spend
S57790271 2715.86999999997
150318419 1645.9700000000003
562167663 1352.849999999999
531900524 1329.45
557850743 1295.479999999998
532130011 1185.389999999999
561592095 1109.7
5131950134 1097.59
566576008 1056.36
521347209 1040.9099999999999
Time taken: 24.267 seconds, Petched: 10 row(s)
```

# Result:

The Top 10 users are listed as above.

A simple order user by descending, Sum(price) and then doing LIMIT 10 did the job.

# **廿 Clean-up: Last Steps:**

Drop the tables:

```
hive> drop table sales_info;
OK
Time taken: 0.266 seconds
hive> drop table opt_sales_info;
OK
Time taken: 0.17 seconds
hive> show tables;
OK
Time taken: 0.054 seconds
hive>
```

# No more tables left.

Drop the database:

```
hive> drop database if exists sales;
OK
Time taken: 0.063 seconds
hive> show databases;
OK
default
Time taken: 0.03 seconds, Fetched: 1 row(s)
hive>
```

Our database is dropped now.

Terminate the database:

Go to the Cluster's page -> Click on Terminate -> Disable the Termination protection -> Terminate the cluster

