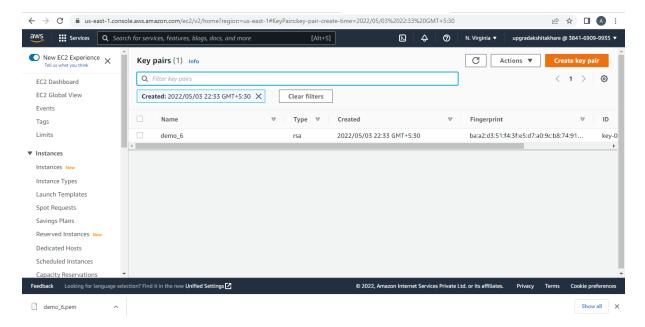
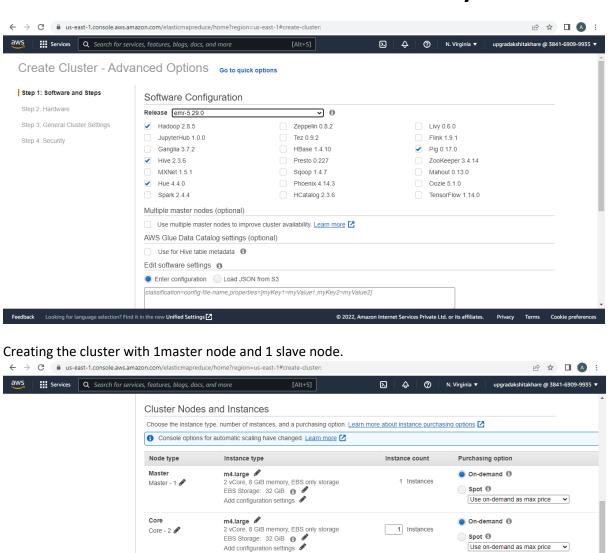
Problem Statement

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. Needless to say, the role of big data analysts is among the most sought-after job profiles of this decade. Therefore, 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 e-commerce company.



Creation of key pair

Creating the EMR version 5.29.0 as suggested.



m5.xlarge

✓ 4 vCore, 16 GiB memory, EBS only storage

EBS Storage: 64 GiB ① Add configuration settings

Task - 3 🎤

Feedback Looking for language selection? Find it in the new Unified Settings ☑

+ Add task instance group

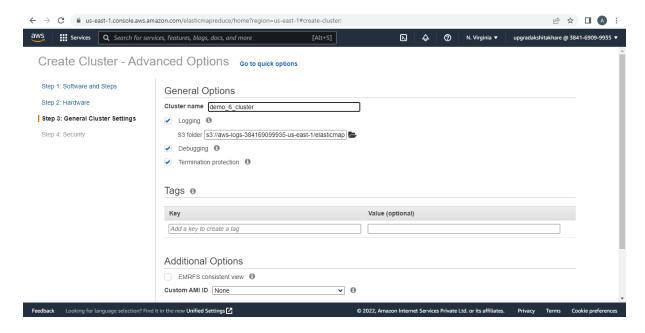
On-demand 🛈

Use on-demand as max price 🔻

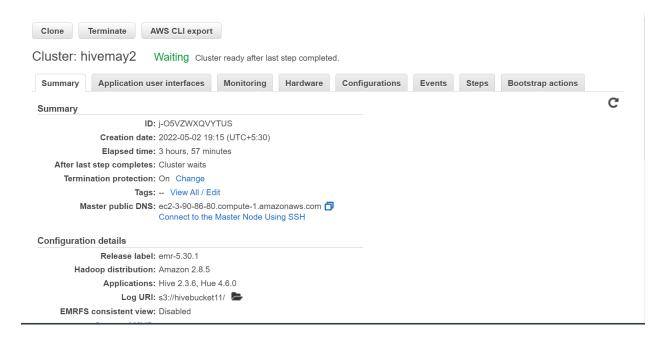
Spot 1

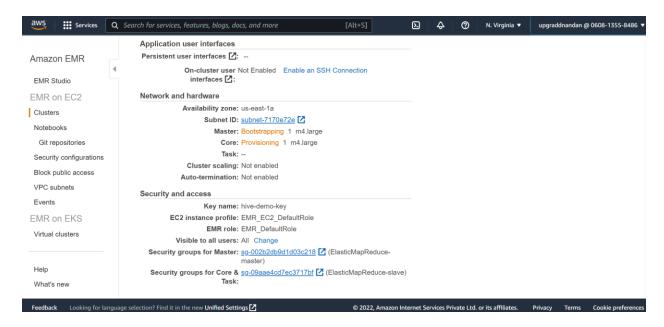
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0 Instances



Loading the 2019-Oct.csv and 2019-Nov.csv to s3 hivebucket 11





We loaded the files in S3 and then loaded the hadoop EMR Cluster

```
Using username "hadoop".
  Authenticating with public key "imported-openssh-key"
                  Amazon Linux 2 AMI
https://aws.amazon.com/amazon-linux-2/
                                  EEEEEEEEEEEEEEEEEEE MMMMMMMM
EE:::::EEEEEEEEE:::E M:::::::M
                                 M:::::::M R:::::RRRRRR:::::R
 E::::E
           EEEEE M:::::::M
                                R::::R
                  M::::::M:::M
                              M:::M:::::M
                                           R:::R
 E::::EEEEEEEEE
                  M:::::M M:::M M::::M
                                           R:::RRRRRR::::R
 E:::::E
                  M:::::M M::::M:::M M:::::M
                                           R:::::::::::::::::::RR
 E::::EEEEEEEEE
                                           R:::RRRRRR::::R
                  M:::::M
                          M:::::M
                                   M:::::M
                                            R:::R
 E::::E
                  M:::::M
                           M:::M
                                   M:::::M
                                                     R::::R
            EEEEE M:::::M
                            MMM
                                   M:::::M
                                            R:::R
                                                     R::::R
EE:::::EEEEEEEE::::E M:::::M
                                   M:::::M
                                   M:::::M RR::::R
EEEEEEEEEEEEEEEEE MMMMMMM
                                   MMMMMM RRRRRR
                                                     RRRRRR
[hadoop@ip-172-31-44-20 ~]$ pwd
/home/hadoop
[hadoop@ip-172-31-44-20 ~]$ aws s3 cp s3://hivebucket11/2019-Oct.csv .
download: s3://hivebucket11/2019-Oct.csv to ./2019-Oct.csv
[hadoop@ip-172-31-44-20 \sim]$ ls
2019-Oct.csv
[hadoop@ip-172-31-44-20 \sim]$ aws s3 cp s3://hivebucket11/2019-Nov.csv.
download: s3://hivebucket11/2019-Nov.csv to ./2019-Nov.csv
[hadoop@ip-172-31-44-20 ~]$ ls
2019-Nov.csv 2019-Oct.csv
```

Creating the database ,table and loading the Oct and Nov data to the same table. Altering the table with time stamp format.

```
Logging initialized using configuration in file:/etc/hive/conf.dist/hive-log4j2.properties Async: false hive> create database if not exists edata;
Time taken: 0.895 seconds
hive> use edata;
Time taken: 0.12 seconds
hive> CREATE EXTERNAL TABLE IF NOT EXISTS ecom events (
     > 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
     > )
> COMMENT 'ecom_events Table'
> ROW FORMAT DELIMITED
> FIELDS TERMINATED BY ','
> LOCATION '/tmp/ecomdata'; ;
hive> show tables;
Time taken: 0.137 seconds, Fetched: 1 row(s)
hive> Load data local inpath '/home/hadoop/2019-Oct.csv' into table ecom_events;
Loading data to table edata.ecom_events
Time taken: 9.987 seconds
hive> ALTER TABLE ecom_events SET TBLPROPERTIES ("skip.header.line.count"="1");
Time taken: 0.14 seconds
     > ALTER TABLE ecom events SET TBLPROPERTIES ("timestamp.formats"="yyyy-MM-d
Time taken: 0.114 seconds
         set hive.cli.print.header=true;
```

Describing the table:

```
hive> describe ecom events;
OK
                       timestamp
event_time
event_type
                       string
product_id
                       string
category_id
                       string
category_code
                       string
brand
                       string
                       float
price
                       bigint
user_id
user session
                        string
Time taken: 0.103 seconds, Fetched: 9 row(s)
hive> describe ecom_events;
```

Checking for the loaded table by using cmd select * from ecom events limit 10;

```
hive> use edata;
OK
Time taken: 0.118 seconds
hive> show tables;
OK
ecom_events
Time taken: 0.343 seconds, Fetched: 1 row(s)
hive> load data local inpath'/home/hadoop/2019-Oct.csv' into table ecom_events;
Loading data to table edata.ecom_events
OK
Time taken: 9.365 seconds
hive> select * from ecom_events limit 10;
OK
NULL event_type product_id category_id category_id runail 2.62 463240011 26dd6e6e-4dac-4778-8d2c-92e149dab885
NULL cart 5773353 1487580005134238553 runail 2.62 463240011 26dd6e6e-4dac-4778-8d2c-92e149dab885
NULL cart 573359 1487580005134238553 runail 2.62 463240011 26dd6e6e-4dac-4778-8d2c-92e149dab885
NULL cart 5881589 2151191071051219817 lovely 13.48 429681830 49e8d43-adf3-428b-a2c3-fe8bc6a307c9
NULL cart 5881589 148758001534238553 runail 2.62 463240011 26dd6e6e-4dac-4778-8d2c-92e149dab885
NULL cart 5885269 1487580005134238553 runail 2.62 463240011 26dd6e6e-4dac-4778-8d2c-92e149dab885
NULL cart 5885269 1487580005134238553 runail 2.62 430174032 73deale7-664e-4364-8b30-ad329d5af04f
NULL cart 5885269 1487580008246412266 kapous 4.75 377667011 81326ac6-daa4-4f0a-b488-f0956a78733
NULL cart 569899 1487580008445982239 0.56 467916806 2f5b5546-b8cb-9ee7-7ecd-64276f8ef486
NULL cart 569899 1487580006317032337 1.27 38598599 d30965e8-1101-44ab-b45d-cclbb9fae694
Time taken: 2.362 seconds, Fetched: 10 row(s)
```

Hive Assignment Questions:

Running the queries before partition and screen shots for same:

QUERY1: Find the total revenue generated due to purchases made in October.

select sum(price) as oct_revenue from ecom_events where event_type='purchase' and month(event time)=10;

```
hive> select sum(price) as oct_revenue from ecom_events where event_type='purchase' and month(event_time)=10;
Query ID = hadoop_20220503163341_flc35306-5239-4204-97de-a656ea6dfc50
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_1651587703400_0004)

VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

Map 1 ...... container SUCCEEDED 1 1 0 0 0 0 0
Reducer 2 ..... container SUCCEEDED 1 1 0 0 0 0 0
VERTICES: 02/02 [==========>>] 100% ELAPSED TIME: 13.72 s

OK
oct_revenue
1211538.4295325726
Time taken: 27.1 seconds, Fetched: 1 row(s)
hive>
```

QUERY2: Write a query to yield the total sum of purchases per month in a single output.

select month(event_time), sum(price) as purchase_sum from ecom_events where event_type='purchase' group by month(event_time);

QUERY3: Write a query to find the change in revenue generated due to purchases from October to November.

```
WITH month_revenue AS

(SELECT

SUM(case when MONTH(event_time) = '10' then price else 0 end) AS Oct_Revenue,

SUM(case when MONTH(event_time) = '11' then price else 0 end) AS Nov_Revenue

FROM ecom_events

WHERE event_type= 'purchase'

AND MONTH(event_time) in ('10', '11')

)

SELECT (Oct_Revenue - Nov_Revenue) FROM month_revenue;
```

```
hive> WITH month revenue AS
              SUM(case when MONTH(event_time) = '10' then price else 0 end) AS Oct_Revenue,
               SUM(case when MONTH(event time) = '11' then price else 0 end) AS Nov Revenue
         FROM ecom_events
          WHERE event_type= 'purchase'
            AND MONTH (event_time) in ('10', '11')
         SELECT (Oct_Revenue - Nov_Revenue) FROM month_revenue ;
Query ID = hadoop_20220503174218 afe197b0-3f5e-44d6-951f-74080def4716
Launching Job 1 out of 1
Session re-established.
Status: Running (Executing on YARN cluster with App id application_1651587703400_0007)
        VERTICES
                     MODE
                                 STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
Map 1 ..... container SUCCEEDED Reducer 2 .... container SUCCEEDED
 -319478.469592195
Time taken: 46.791 seconds, Fetched: 1 row(s)
```

QUERY4: Find distinct categories of products. Categories with null category code can be ignored.

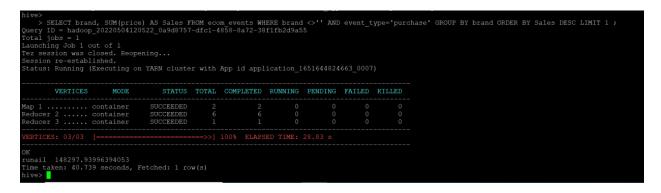
SELECT DISTINCT split(category_code, '\\.')[0] AS category FROM ecom_events WHERE split(category_code, '\\.')[0]<>";

QUERY5: Find the total number of products available under each category.

SELECT split(category_code,'\\.')[0] AS category, COUNT(product_id) AS prd FROM ecom_events GROUP BY split(category_code,'\\.')[0] ORDER BY prd DESC;

QUERY6: Which brand had the maximum sales in October and November combined?

SELECT brand, SUM(price) AS Sales FROM ecom_events WHERE brand <>" AND event_type='purchase' GROUP BY brand ORDER BY Sales DESC LIMIT 1;



QUERY7: Which brands increased their sales from October to November?

WITH monthly_diff AS (SELECT brand, SUM(CASE WHEN date_format(event_time,'MM')=10 THEN price ELSE 0 END) AS October, SUM(CASE WHEN date_format(event_time,'MM')=11 THEN price ELSE 0 END) AS November FROM ecom_events WHERE event_type='purchase' GROUP BY brand) SELECT brand, October, November, (November-October) as Sales_diff FROM monthly_diff WHERE (November-October) >0 ORDER BY Sales_diff;

```
931.0899903774261
         902.3799939155579
                                                            28.709996461868286
                                  90.30999946594238
                                                            29.659997940063477
beautyblender 78.73999977111816
                                        109.41000175476074
                                                                  30.670001983642578
vilenta 197.59999787807465
                                 231.20999908447266
                                                            33.61000120639801
mavala 409.0400023460388
likato 296.0599980354309
                                  446.32000255584717
                                  340.9699954986572
                                                            44.90999746322632
ladykin 125.64999961853027
                                  170.56999969482422
                                                            44.920000076293945
foamie 35.03999996185303
elskin 251.0900001525879
                                                            45.44999980926514
                                  307.6499996781349
                                                            56.55999952554703
               155.33000373840332
balbcare
                                          212.3800015449524
                                                                    57.04999780654907
koelcia 55.5 112.75 57.25
profhenna 679.2300038337708
kares 0.0 59.45000076293945
                                           736.8500001430511
                                                                     57.619996309280396
                                      109.33000040054321
61.28999876976013
                                                                     60.11000061035156
        0.0 61.28999876976013 61.289998769
288.01999855041504 351.2099983692169
                                                            63.18999981880188
laboratorium 246.49999952316284 312.5199975967407
                                                                    66.01999807357788
cutrin 299.3700017929077 367.6199998855591 68.24999809265137
egomania
                 77.46999835968018 146.04000091552734 68.57000255584717
konad 739.8300001621246
                                                            70.83999770879745
nirvel 163.04000329971313
                                  234.33000826835632
                                                            71.29000496864319
koelf 422.7300081253052
plazan 101.37000036239624
                                                            84.55999374389648
                                 194.010000705719
177.5100040435791
                                                            92.64000034332275
        83.95000076293945
                                                            93.56000328063965
kerasys 430.9100044965744
                                  525.2000050544739
                                                            94.29000055789948
enjoy 41.34999966621399
                                  136.57000184059143
                                                            95.22000217437744
                               106384 2803.7799961566925
152.60999727249146 98.
243.3599967956543 98.
                                                            925 96.70999884605408
98.26999711990356
depilflax
             2707.0699973106384
        54.34000015258789
                                                            98.27999782562256
                                                            101.76999497413635
batiste 772.400013923645
                                  874.1700088977814
        645.5800037384033
osmo
        819.1300112009048
                                  945.5100176334381
dizao
                                                            126.38000643253326
igrobeauty 513.6600003838539
                                       645.0699995160103 131.40999913215637
                                                            132.00000047683716
nefertiti 233.51999759674072 366.64000034332275 133.12000274658203 elizavecca 70.52999973297119 204.29999923706055 133.76999950408936
miskin 158.04000186920166 293.0700011253357 latinoil 249.5199966430664 384 589000
                                                           135.02999925613403
            249.5199966430664 384.5899987220764 135.07000207901
farmona 1692.46000289917 1843.4299907684326 150.9699878692627
cristalinas 427.63000297546387 584.950008392334 157.32000541687012
      358.93999576568604 538.6099972724915
                                                           179.67000150680542
               0.0 182.66999757289886 182.66999757289886
matreshka
                                           502.3399975299835
                 318.69999980926514
freshbubble
                                                                    183.63999772071838
                                                            193.4699993133545
        66.79000186920166
```

```
metzger 5373.4499744176865
                               6457.159960865974
                                                       1083.709986448288
                               2775.510024756193
de.lux 1659.7000161707401
                                                       1115.810008585453
swarovski
               1887.9299856424332
                                       3043.1599831581116
                                                              1155.2299975156784
                                       1782.8599914312363
beauty-free
               554.1699986457825
                                                              1228.6899927854538
                               2009.6300013065338
                                                     1300.9699981212616
        705.5200037956238
                               2015.1000146865845
                                                       1309.5800108909607
              4775.8799668848515 6120.479953020811
                                                            1344.5999861359596
severina
irisk 45591.96021157503
                               46946.04018642008
                                                      1354.0799748450518
       8425.409879207611
                               9841.649902820587
levrana 2243.5599967837334
                               3664.0999879837036
                                                      1420.5399911999702
roubloff
               3491.3600150346756
                                       4913.770027637482
                                                              1422.410012602806
                                                      1444.8799936771393
       4457.259982824326
                               5902.139976501465
        3341.199989080429
                               4839.720018148422
                                                      1498.5200290679932
shik
domix
       10472.05003106594
                               12009.170008182526
                                                      1537.1199771165848
                                                       1596.6100018024445
artex
        2730.6399517059326
                               4327.249953508377
beautix 10493.949965000153
                               12222.95004272461
                                                       1729.0000777244568
        3904.940046072006
                               5642.01002573967
                                                       1737.0699796676636
milv
masura
       31266.079910814762
                               33058.469878435135
                                                       1953.0500071048737
       11927.159952402115
                               14093.079938054085
                                                       2165.91998565197
concept 11032.14000660181
                               13380.400002479553
                                                       2348.2599958777428
       21756.749947547913
                               24142.66994935274
                                                       2385.9200018048286
kaypro 881.3400187492371
                               3268.700007915497
                                                       2387.3599891662598
benovy 409.619996547699
                               3259.969982147217
                                                       2850.349985599518
italwax 21940.239994883537
                                                       2859.130049407482
                               24799.37004429102
     8756.910053431988
                               11707.88005465269
                                                      2950.970001220703
yoko
               9390.690077126026
                                      12352.910059452057
                                                              2962.2199823260307
haruyama
                                      10273.099990844727
               7280.749939441681
marathon
lovely 8704.380010932684
                               11939.059989094734
                                                      3234.6799781620502
bpw.style
               11572.1500659585 14837.440190911293 3265.290124952793
                                                     3355.8799889683723
staleks 8519.730030417442
               3421.7800273299217
freedecor
                                    7671.800070524216
                                                             4250.020043194294
                               76758.65991047397
                                                      5219.379857007414
polarus 6013.720007181168
                                                       5358.210015535355
cosmoprofi 8322.80991601944
               26287.840348243713
                                       33345.23023867607
                                                              7057.389890432358
iessnail
                               38671.27037525177
                                                      9474.640277385712
                                                             10404.820005103946
ingarden
                               16394.239884018898
lianail 5892.839952707291
                                                       10501.399931311607
       35302.029363155365
                               51039.74947929382
                                                       15737.720116138458
grattol 35445.53947067261
                               71472.70888674259
                                                       36027.169416069984
        474679.05964545906
                               619509.2397020273
                                                       144830.18005656824
Time taken: 31.606 seconds, Fetched: 161 row(s)
hive>
```

QUERY8: 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.

SELECT user_id, SUM(price) AS expense FROM ecom_events WHERE event_type='purchase' GROUP BY user_id ORDER BY expense DESC LIMIT 10;

Performance analysis running query before partition and after partition and screen shots for same:

```
hive> select * from clickStream2019 limit 5;

OK

OK

2019-11-01 00:00:02 UTC view 5802432 1487580009286598681 0
.32 562076640 09fard6c-6c99-46h1-834f-33527fddc241

2019-11-01 00:00:00 UTC cart 54397 1487580006317032337 2
.38 553329724 2067216c-31b5-455d-alco-af0575a34ffb

2019-11-01 00:00:10 UTC view 583716c 1733999064100130704 pnb 2
2.22 556138645 57ed222e-a54a-4907-9944-5a875c2d7f4f

2019-11-01 00:00:11 UTC cart 5876812 1487580010100285687 jesnail
3.16 54506666 186c1951-0802-4354-63056410547

2019-11-01 00:00:24 UTC remove from_cart 5826182 1487580007483048900 3
.33 553329724 2067216c-3155-4555-4510c-af0575a34ffb

Time taken: 2.292 seconds, Fetched: 5 row(s)
```

Select * on non-partitioned table took 2.292 sec

Select * on partitioned table took 0.229 sec

QUERY2: Write a query to yield the total sum of purchases per month in a single output.

select month(event_time), sum(price) as purchase_sum from ecom_events where event_type='purchase' group by month(event_time);

2nd query on non-partitioned table took 74.247 sec

```
hive> select month, sum(price) as sum purchase from part_cs_2019 where event_type='purchase' group by month;

Query ID = hadcop_20220504115946_8241a2a6-bc66-41f1-a57f-de5aab88e4c3
Total jobs = 1

Launching Job 1 out of 1

Status: Running (Executing on YARN cluster with App id application_1651662329711_0003)

VERTICES MODE STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED

Map 1 ....... container SUCCEEDED 7 7 0 0 0 0

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

VERTICES: 02/02 [=======>>] 100% ELAPSED TIME: 29.45 s

OK

10 1211538.430
11 1531016.900
Time taken: 30.092 seconds, Fetched: 2 row(s)
```

2nd query on partitioned table took 30.092 sec

Partitioning on the basis of month:

Partitioning segregates all the entries for the various columns of the dataset on the basis of the parameter chosen for partitioning and stores the data in their respective partitions. Hence, While we write the query to fetch the values from the table, only the required partitions of the table are queried. Thus, it reduces the time taken by the query to yield the result. In dynamic partitioning, the values of partitioned columns exist within the table. So, it is not required to pass the values of partitioned columns manually. It works well for columns having low cardinality.

There are two modes of dynamic partitioning:

Strict: This needs at least one column to be static while loading the data.

Non-strict: This allows us to have dynamic values of all the partition columns.

In the next snapshot, we have created a dynamic partitioned table named "part_cs_2019" and how data has been loaded into it using the "insert" clause:

By default, dynamic partitioning is not allowed in Hive. So it has to be enabled .

```
Aveo create table if not exists part_C2_2019 (event_time timestamp, event_type string, product_id string, category_sid string, category_code string, brind string, price decimal(10,3) user_id bigint, user_session string) partitioned by (month int) row format delimited fields terminated by "\" lines terminated by "\" stored as textfile;
or a staken: 0.141 seconds hives and tables;
or clickstream2019
part_c3_2019
Time taken: 0.068 seconds, Fetched: 2 row(s)
Time taken: 0.068 seconds, Fetched: 2 row(s)
Time taken: 10.068 secon
```

After partitioning the time taken for the execution of query is less.

Accessing the content of the table is much faster after partition and buckecting.

Query 4) after partitioning

Query 6) after partitioning

Bucketing:

Bucketing provides flexibility to further segregate the data into more manageable sections called buckets or clusters. **CLUSTERED BY** clause is used to divide the table into buckets. It works well for the columns having high cardinality.

```
As the property of the propert
```

Running query 6 after bucketing