

Scripts Execution

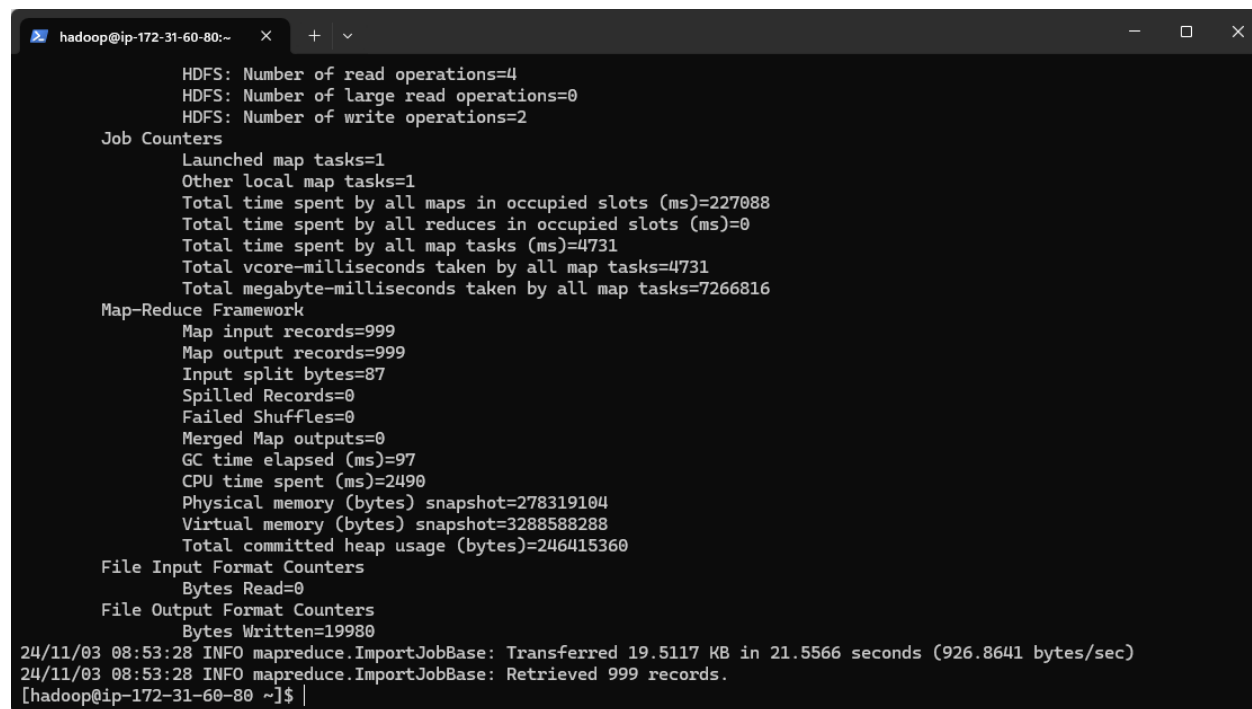
Explanation of the solution to the batch layer problem

<Properly explain the step by step process followed for the completion of tasks till **task 4**. The steps should be properly documented here.>

- Sqoop command to import tables from RDS to HDFS:

Table 1: Command for member_score table:

```
sqoop import \
--connect jdbc:mysql://upgradawsrds1.cyaie1c9bmnf.us-east-
1.rds.amazonaws.com/cred_financials_data \
--table member_score \
--username upgraduser --password upgraduser \
--target-dir /user/root/cap_project/member_score \
-m 1
```

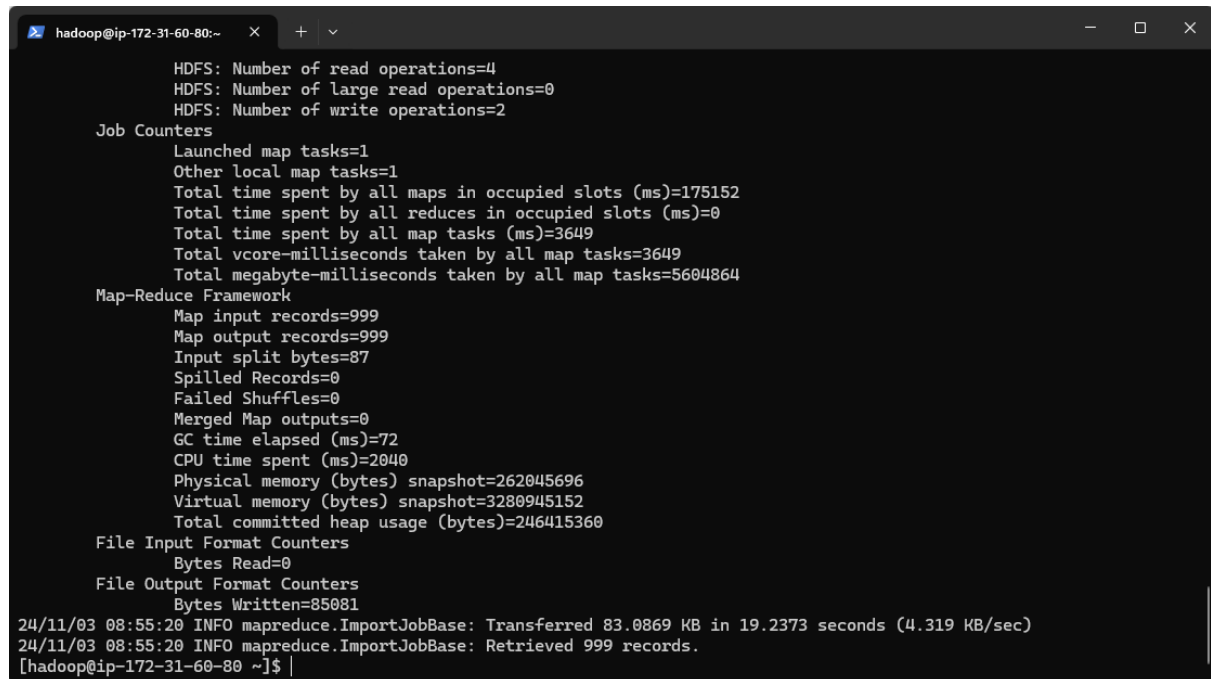


```
hadoop@ip-172-31-60-80:~$ sqoop import \
--connect jdbc:mysql://upgradawsrds1.cyaie1c9bmnf.us-east-
1.rds.amazonaws.com/cred_financials_data \
--table member_score \
--username upgraduser --password upgraduser \
--target-dir /user/root/cap_project/member_score \
-m 1

HDFS: Number of read operations=4
HDFS: Number of large read operations=0
HDFS: Number of write operations=2
Job Counters
  Launched map tasks=1
  Other local map tasks=1
  Total time spent by all maps in occupied slots (ms)=227088
  Total time spent by all reduces in occupied slots (ms)=0
  Total time spent by all map tasks (ms)=4731
  Total vcore-milliseconds taken by all map tasks=4731
  Total megabyte-milliseconds taken by all map tasks=7266816
Map-Reduce Framework
  Map input records=999
  Map output records=999
  Input split bytes=87
  Spilled Records=0
  Failed Shuffles=0
  Merged Map outputs=0
  GC time elapsed (ms)=97
  CPU time spent (ms)=2490
  Physical memory (bytes) snapshot=278319104
  Virtual memory (bytes) snapshot=3288588288
  Total committed heap usage (bytes)=246415360
File Input Format Counters
  Bytes Read=0
File Output Format Counters
  Bytes Written=19980
24/11/03 08:53:28 INFO mapreduce.ImportJobBase: Transferred 19.5117 KB in 21.5566 seconds (926.8641 bytes/sec)
24/11/03 08:53:28 INFO mapreduce.ImportJobBase: Retrieved 999 records.
[hadoop@ip-172-31-60-80 ~]$
```

Table 2: Command for card_member table:

```
sqoop import \
--connect jdbc:mysql://upgradawsrds1.cyaieic9bmnf.us-east-
1.rds.amazonaws.com/cred_financials_data \
--table card_member \
--username upgraduser --password upgraduser \
--target-dir /user/root/cap_project/card_member \
-m 1
```



```
hadoop@ip-172-31-60-80:~$
HDFS: Number of read operations=4
HDFS: Number of large read operations=0
HDFS: Number of write operations=2
Job Counters
  Launched map tasks=1
  Other local map tasks=1
  Total time spent by all maps in occupied slots (ms)=175152
  Total time spent by all reduces in occupied slots (ms)=0
  Total time spent by all map tasks (ms)=3649
  Total vcore-milliseconds taken by all map tasks=3649
  Total megabyte-milliseconds taken by all map tasks=5604864
Map-Reduce Framework
  Map input records=999
  Map output records=999
  Input split bytes=87
  Spilled Records=0
  Failed Shuffles=0
  Merged Map outputs=0
  GC time elapsed (ms)=72
  CPU time spent (ms)=2040
  Physical memory (bytes) snapshot=262045696
  Virtual memory (bytes) snapshot=3280945152
  Total committed heap usage (bytes)=246415360
File Input Format Counters
  Bytes Read=0
File Output Format Counters
  Bytes Written=85081
24/11/03 08:55:20 INFO mapreduce.ImportJobBase: Transferred 83.0869 KB in 19.2373 seconds (4.319 KB/sec)
24/11/03 08:55:20 INFO mapreduce.ImportJobBase: Retrieved 999 records.
[hadoop@ip-172-31-60-80 ~]$
```

- Verify the data inside the imported directories:

- `hadoop fs -cat /user/root/cap_project/member_score/part-m-00000`
- `hadoop fs -cat /user/root/cap_project/card_member/part-m-00000`

```
973974062665904,565
974087224071871,682
975111321017949,306
976362362966586,210
976740397894593,303
978465390240911,270
978786807247400,652
979218131207765,552
979274357975688,353
981612675682175,265
982171321477033,683
983248478716258,473
983390343419665,183
985095728641698,325
985619195348046,310
987837606639725,623
988661034778977,496
988666952112281,531
989174793157941,533
989548647998824,506
990571384923260,469
990740662283435,202
991784667284668,621
991958482478439,215
992104321998571,658
992552823055811,694
994983851226493,687
996411635289270,335
997128952368160,683
[hadoop@ip-172-31-60-80 ~]$
```

```
6562510549485881,659982919406634,2015-07-04 06:39:15.0,11/16,United States,Mount Vernon
6566718331696237,335526117403543,2013-09-19 10:34:17.0,08/14,United States,Sand Springs
6568438963325843,183038102441540,2017-07-06 05:25:55.0,07/17,United States,North Las Vegas
6570061543816002,722794052951029,2017-01-10 03:01:36.0,02/17,United States,Sedalia
6573615549671948,985619195348046,2015-09-29 08:21:37.0,12/16,United States,Belleville
6574255180086418,891702243060747,2012-07-19 07:47:35.0,07/15,United States,Selma
6574656060675651,555127039477442,2010-06-02 00:11:01.0,11/10,United States,Coral Terrace
6577397607792660,479331531880555,2013-06-13 00:02:59.0,02/16,United States,Norco
6578328793401039,149907932138330,2016-03-21 02:51:11.0,11/16,United States,Wasco
6580194731591489,218337397923739,2018-01-09 00:32:14.0,01/18,United States,Oxford
6582135370631324,389050970153068,2016-07-17 10:26:54.0,08/16,United States,Roy
6582950745622331,185289866247329,2012-04-17 03:37:04.0,03/13,United States,Hutchinson
6583283909793527,398607904778893,2015-08-27 03:19:32.0,01/17,United States,Galesburg
6585229076432173,887283455220105,2015-03-06 00:53:06.0,05/16,United States,Security-Widefield
6587114664896609,623414156260264,2015-10-05 06:45:35.0,03/17,United States,Hobart
6588637504808417,678537053778506,2012-05-30 04:37:45.0,04/14,United States,Concord
6589539191695876,832393339334706,2014-01-26 11:59:35.0,09/15,United States,Roseville
6589855658684769,193386908982058,2010-02-07 09:52:17.0,11/15,United States,Westfield
6589894320960430,276119478113309,2011-08-08 02:25:29.0,11/15,United States,Blaine
6590907016354002,849289511810887,2016-10-28 07:32:23.0,07/17,United States,Reading
6591175617713393,135615957283320,2015-05-16 03:21:15.0,08/16,United States,Arvin
6592184145413632,714853788321318,2010-03-12 08:11:17.0,08/16,United States,Mehlville
6594248319343442,346162645575666,2016-02-15 02:32:58.0,12/16,United States,Lyndhurst
6595638658736751,377918213614639,2015-08-13 10:58:24.0,09/17,United States,Mahwah
6595814135833988,236864426408837,2011-02-18 06:45:46.0,06/14,United States,Johnston
6595928469079750,422173403853722,2012-02-10 02:19:20.0,02/14,United States,Raleigh
6597703848279563,657991752750272,2017-02-16 07:45:34.0,03/17,United States,Pinewood
6598830758632447,660350842993890,2013-07-26 02:22:17.0,12/16,United States,Tampa
6599900931314251,928036864799687,2017-09-09 03:31:54.0,10/17,United States,Loves Park
[hadoop@ip-172-31-60-80 ~]$
```

- Command to load card_transactions.csv to HDFS after moving to EC2-USER:

```
hadoop fs -copyFromLocal /home/hadoop/card_transactions.csv
/user/root/cap_project/card_transactions.csv
```

- Connect to instance over putty and load the jupyter notebook from root user:

```
jupyter notebook --port 7861 --allow-root
```

- Loading member score data stored in central AWS RDS:

```
from pyspark.sql import SparkSession
from pyspark.sql.types import StructType, StructField, StringType, IntegerType
```

```
memberschema = StructType([
    StructField('member_id', StringType(), False),
    StructField('score', IntegerType(), False),
])

memf = spark.read.csv("hdfs:/user/root/cap_project/member_score", header=False,
schema=memberschema)
memf.show()
```

```
memf.show()
```

member_id	score
000037495066290	339
000117826301530	289
001147922084344	393
001314074991813	225
001739553947511	642
003761426295463	413
004494068832701	217
006836124210484	504
006991872634058	697
007955566230397	372
008732267588672	213

- Loading the card holder's data stored in central AWS RDS:

```
from pyspark.sql import SparkSession
from pyspark.sql.types import StructType, StructField, StringType

# Initialize SparkSession
spark = SparkSession.builder \
    .appName("Card Holder Data Load") \
    .getOrCreate()

# Define the schema for card holder's data
cardschema = StructType([
    StructField('card_id', StringType(), False),
    StructField('member_id', StringType(), False),
    StructField('member_joining_dt', StringType(), False),
    StructField('card_purchase_dt', StringType(), False),
    StructField('country', StringType(), False),
    StructField('city', StringType(), False),
])

# Read the data from HDFS
cardf = spark.read.csv("hdfs:/user/root/cap_project/card_member", header=False, schema=cardschema)

# Display the data
cardf.show()
```

```
cardschema = StructType([StructField('card_id', StringType(),False),
                             StructField('member_id', StringType(),False),
                             StructField('member_joining_dt', StringType(),False),
                             StructField('card_purchase_dt', StringType(),False),
                             StructField('country', StringType(),False),
                             StructField('city', StringType(),False),
                             ])
```

```
cardf = spark.read.csv("hdfs://user/root/cap_project/card_member", header = False, schema = cardschema)
```

```
cardf.show()
```

card_id	member_id	member_joining_dt	card_purchase_dt	country	city
340028465709212	009250698176266	2012-02-08 06:04:...	05/13	United States	Barberton
340054675199675	835873341185231	2017-03-10 09:24:...	03/17	United States	Fort Dodge
340082915339645	512969555857346	2014-02-15 06:30:...	07/14	United States	Graham
340134186926007	887711945571282	2012-02-05 01:21:...	02/13	United States	Dix Hills
340265728490548	680324265406190	2014-03-29 07:49:...	11/14	United States	Rancho Cucamonga
340268219434811	929799084911715	2012-07-08 02:46:...	08/12	United States	San Francisco
340379737226464	089615510858348	2010-03-10 00:06:...	09/10	United States	Clinton
340383645652108	181180599313885	2012-02-24 05:32:...	10/16	United States	West New York
340803866934451	417664728506297	2015-05-21 04:30:...	08/17	United States	Beaverton

- Loading card_transaction data from csv:

```
transasction = StructType([StructField('card_id', StringType(),False),
                             StructField('member_id', StringType(),False),
                             StructField('amount', IntegerType(),False),
                             StructField('postcode', StringType(),False),
                             StructField('pos_id', StringType(),False),
                             StructField('transaction_dt', StringType(),False),
                             StructField('status', StringType(),False),
                             ])
```

```
tranf = spark.read.csv("hdfs://user/root/cap_project/card_transactions.csv", header = True, schema = transasction)
```

```
tranf= tranf.filter(tranf.status!='FRAUD')
```

```
tranf.show()
```

card_id	member_id	amount	postcode	pos_id	transaction_dt	status
348702330256514	000037495066290	9084849	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	330148	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	136052	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	4310362	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	9097094	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	2291118	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	4900011	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	633447	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	6259303	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	369067	33946	614677375609919	11-02-2018 00:00:00	GENUINE
348702330256514	000037495066290	1193207	33946	614677375609919	11-02-2018 00:00:00	GENUINE

- Joining the member_score and card_member on member_id to extract credit score of each member and selecting the required fields:

```
# Join member_score and card_member DataFrames on 'member_id'
score = memf.join(cardf, memf.member_id == cardf.member_id, how='left')

# Print the schema to verify the join
score.printSchema()

# Select the relevant columns: 'member_id', 'score', and 'card_id'
score = score.select('member_id', 'score', 'card_id')

# Show a preview of the score DataFrame
score.show()
```



```
score = memf.join(cardf, memf.mem_id == cardf.member_id,how='LEFT')
```

```
score.printSchema()
```

```
root
|-- mem_id: string (nullable = true)
|-- score: integer (nullable = true)
|-- card_id: string (nullable = true)
|-- member_id: string (nullable = true)
|-- member_joining_dt: string (nullable = true)
|-- card_purchase_dt: string (nullable = true)
|-- country: string (nullable = true)
|-- city: string (nullable = true)
```

```
score = score.select('mem_id', 'score', 'card_id')
```

```
score.show()
```

mem_id	score	card_id
000037495066290	339	348702330256514
000117826301530	289	5189563368503974
001147922084344	393	5407073344486464
001314074991813	225	378303738095292
001739553947511	642	348413196172048
003761426295463	413	348536585266345
004494068832701	217	5515987071565183

- Joining both the history transactions CSV and score DF:

```
hist = tranf.join(score, tranf.member_id == score.mem_id,how='outer')
```

```
hist.printSchema()
```

```
root
|-- card_id: string (nullable = true)
|-- member_id: string (nullable = true)
|-- amount: integer (nullable = true)
|-- postcode: string (nullable = true)
|-- pos_id: string (nullable = true)
|-- transaction_dt: string (nullable = true)
|-- status: string (nullable = true)
|-- mem_id: string (nullable = true)
|-- score: integer (nullable = true)
|-- cardid: string (nullable = true)
```

```
hist = hist.select('card_id', 'amount', 'postcode', 'pos_id', 'transaction_dt', 'status', 'score')
```

```
hist.show()
```

card_id	amount	postcode	pos_id	transaction_dt	status	score
340379737226464	6126197	46933	167473544283898	01-05-2016 08:10:50	GENUINE	229
340379737226464	7949232	61840	664980919335952	01-10-2016 10:38:52	GENUINE	229
340379737226464	943839	91743	633038040069180	02-08-2016 00:31:25	GENUINE	229
340379737226464	3764114	91743	633038040069180	02-08-2016 21:35:27	GENUINE	229
340379737226464	6221251	98384	064948657945290	02-10-2016 14:44:14	GENUINE	229
340379737226464	2868312	26032	856772774421259	02-12-2016 21:55:43	GENUINE	229
340379737226464	4418586	20129	390339673634463	02-12-2017 17:05:51	GENUINE	229

- Compute the max transaction date:

```
look_up_table = history.groupBy('card_id').agg(f.max("transaction_date").alias('transaction_date'))
```

```
look_up_table.show()
```

card_id	transaction_date
340379737226464	2018-01-27 00:19:47
377201318164757	2017-11-28 16:32:22
348962542187595	2018-01-29 17:17:14
4389973676463558	2018-01-26 13:47:46
5403923427969691	2018-01-22 23:46:19
345406224887566	2017-12-25 04:03:58
6562510549485881	2018-01-17 08:35:27
5508842242491554	2018-01-31 14:55:58
4407230633003235	2018-01-27 07:21:08
379321864695232	2018-01-03 00:29:37

- Inner join on look up table dataset on card_id:

```
look_up_table = look_up_table.join(score, look_up_table.card_id == score.cardid, how='INNER')
```

```
look_up_table.show()
```

card_id	transaction_date	mem_id	score	cardid
340379737226464	2018-01-27 00:19:47	089615510858348	229	340379737226464
345406224887566	2017-12-25 04:03:58	296206661780881	349	345406224887566
348962542187595	2018-01-29 17:17:14	366246487993992	522	348962542187595
377201318164757	2017-11-28 16:32:22	924475891017022	432	377201318164757

- Calculate the Upper Control Limit (UCL), $UCL = \text{Moving Average} + 3 * (\text{Standard Deviation})$, We shall first calculate the moving average of card amount for last 10 transactions.
- Create a window over existing dataframe and aggregate the same card_id, the dataframe is grouped by card_id and then order by transaction_date.

```
window = Window.partitionBy(history['card_id']).orderBy(history['transaction_date'].desc())
```

```
history_df = history.select('*', f.rank().over(window).alias('rank')).filter(f.col('rank') <= 10)
```

```
history_df.show()
```

card_id	amount	postcode	pos_id	status	score	transaction_date	rank
340379737226464	1784098	26656	000383013889790	GENUINE	229	2018-01-27 00:19:47	1
340379737226464	3759577	61334	016312401940277	GENUINE	229	2018-01-18 14:26:09	2
340379737226464	4080612	51338	562082278231631	GENUINE	229	2018-01-14 20:54:02	3
340379737226464	4242710	96105	285501971776349	GENUINE	229	2018-01-11 19:09:55	4
340379737226464	9061517	40932	232455833079472	GENUINE	229	2018-01-10 20:20:33	5
340379737226464	102248	40932	232455833079472	GENUINE	229	2018-01-10 15:04:33	6
340379737226464	7445128	50455	915439934619047	GENUINE	229	2018-01-07 23:52:27	7
340379737226464	5706163	50455	915439934619047	GENUINE	229	2018-01-07 22:07:07	8
340379737226464	8090127	18626	359283931604637	GENUINE	229	2017-12-29 13:24:07	9
340379737226464	9282351	41859	808326141065551	GENUINE	229	2017-12-28 19:50:46	10

To import all SQL functions to pyspark, we need to import the necessary functions (import pyspark.sql.functions as f)

```
history_df = history_df.groupBy("card_id").agg(f.round(f.avg('amount'),2).alias('moving_avg'), \
                                              f.round(f.stddev('amount'),2).alias('Std_Dev'))
```

card_id	moving_avg	Std_Dev
340379737226464	5355453.1	3107063.55
345406224887566	5488456.5	3252527.52
348962542187595	5735629.0	3089916.54
377201318164757	5742377.7	2768545.84
379321864695232	4713319.1	3203114.94
4389973676463558	4923904.7	2306771.9
4407230633003235	4348891.3	3274883.95

- Calculate UCL from the computed standard deviation and moving average:

```
history_df = history_df.withColumn('UCL',history_df.moving_avg+3*(history_df.Std_Dev))
```

card_id	moving_avg	Std_Dev	UCL
340379737226464	5355453.1	3107063.55	1.4676643749999998E7
345406224887566	5488456.5	3252527.52	1.524603906E7
348962542187595	5735629.0	3089916.54	1.5005378620000001E7
377201318164757	5742377.7	2768545.84	1.4048015219999999E7
379321864695232	4713319.1	3203114.94	1.432266392E7
4389973676463558	4923904.7	2306771.9	1.1844220399999999E7
4407230633003235	4348891.3	3274883.95	1.4173543150000002E7

- Joining the dataframe with previous dataframe on card_id:

```
history_df = history_df.select('card_id','UCL')
```

```
look_up_table = look_up_table.join(history_df,on=['card_id'])
```

```
look_up_table.show()
```

card_id	transaction_date	score	postcode	UCL
340379737226464	2018-01-27 00:19:47	229	26656	1.4676643749999998E7
345406224887566	2017-12-25 04:03:58	349	53034	1.524603906E7
348962542187595	2018-01-29 17:17:14	522	27830	1.5005378620000001E7
377201318164757	2017-11-28 16:32:22	432	84302	1.4048015219999999E7
379321864695232	2018-01-03 00:29:37	297	98837	1.432266392E7
4389973676463558	2018-01-26 13:47:46	400	10985	1.1844220399999999E7
4407230633003235	2018-01-27 07:21:08	567	50167	1.4173543150000002E7
5403923427969691	2018-01-22 23:46:19	324	17350	1.411602776E7

- Remove duplicate on redundant transactions done on card_id, transaction_date and postcode:

```
look_up_table = look_up_table.dropDuplicates(['card_id','transaction_date','postcode'])
```

```
look_up_table.count()
```

```
1000
```


1. Load the dataframe into the look up table, happybase API shall be used.

1st step is to connect with hbase:

```
import happybase

# Establish connection to HBase
connection = happybase.Connection('localhost', port=9090, autoconnect=False)
```

```
# Function to open the connection
def open_connection():
    connection.open()

# Function to close the connection
def close_connection():
    connection.close()

# Function to list all tables in HBase
def list_tables():
    print("Fetching all tables")
    open_connection()
    tables = connection.tables()
    close_connection()
    print("All tables fetched")
    return tables
```

```
# Function to create a new HBase table
def create_table(name, cf):
    print("Creating table " + name)
    tables = list_tables()
    if name not in tables:
        open_connection()
        connection.create_table(name, cf)
        close_connection()
        print("Table created")
    else:
        print("Table already present")

# Function to get a table object
def get_table(name):
    open_connection()
    table = connection.table(name)
    close_connection()
    return table
```

If table does not exist, create the table:

```
# Create the lookup table with the specified column family
create_table('look_up_table', {'info': dict(max_versions=5)})
```

```
Creating table 'look_up_table'...
Fetching all tables...
All tables fetched.
Table 'look_up_table' created successfully.
```

Batch insert data into the table:

```
# Function to batch insert data into the HBase table
def batch_insert_data(df, tableName):
    print("Starting batch insert of events")
    table = get_table(tableName)
    open_connection()
    with table.batch(batch_size=4) as b:
        for row in df.rdd.collect():
            b.put(
                bytes(row.card_id),
                {
                    'info:card_id': bytes(row.card_id),
                    'info:transaction_date': bytes(row.transaction_date),
                    'info:score': bytes(row.score),
                    'info:postcode': bytes(row.postcode),
                    'info:UCL': bytes(row.UCL)
                }
            )
    print("Batch insert done")
    close_connection()

# Insert data into HBase table from the DataFrame
batch_insert_data(look_up_table, 'look_up_table')
```

Once the batch insertion is complete, login to putty as root user and enter Hbase shell

```
5232083808576685 column=info:card_id, timestamp=1607880086427, value=5232083808576685
5232083808576685 column=info:postcode, timestamp=1607880086427, value=17968
5232083808576685 column=info:score, timestamp=1607880086427, value=566
5232083808576685 column=info:transaction_date, timestamp=1607880086427, value=2018-01-09 12:44:31
5232271306465150 column=info:UCL, timestamp=1607880087122, value=10951781.35
5232271306465150 column=info:card_id, timestamp=1607880087122, value=5232271306465150
5232271306465150 column=info:postcode, timestamp=1607880087122, value=12920
5232271306465150 column=info:score, timestamp=1607880087122, value=638
5232271306465150 column=info:transaction_date, timestamp=1607880087122, value=2018-01-22 16:44:59
5232695950818720 column=info:UCL, timestamp=1607880087849, value=15220850.52
5232695950818720 column=info:card_id, timestamp=1607880087849, value=5232695950818720
5232695950818720 column=info:postcode, timestamp=1607880087849, value=79080
5232695950818720 column=info:score, timestamp=1607880087849, value=207
5232695950818720 column=info:transaction_date, timestamp=1607880087849, value=2018-01-29 08:30:32
5239380866598772 column=info:UCL, timestamp=1607880086358, value=12835247.22
5239380866598772 column=info:card_id, timestamp=1607880086358, value=5239380866598772
5239380866598772 column=info:postcode, timestamp=1607880086358, value=72471
5239380866598772 column=info:score, timestamp=1607880086358, value=440
5239380866598772 column=info:transaction_date, timestamp=1607880086358, value=2017-12-07 21:44:43
5242841712000086 column=info:UCL, timestamp=1607880088013, value=15646358.41
5242841712000086 column=info:card_id, timestamp=1607880088013, value=5242841712000086
5242841712000086 column=info:postcode, timestamp=1607880088013, value=48821
5242841712000086 column=info:score, timestamp=1607880088013, value=236
5242841712000086 column=info:transaction_date, timestamp=1607880088013, value=2018-01-27 10:51:48
5249623960609831 column=info:UCL, timestamp=1607880087191, value=12497504.76
5249623960609831 column=info:card_id, timestamp=1607880087191, value=5249623960609831
5249623960609831 column=info:postcode, timestamp=1607880087191, value=16858
5249623960609831 column=info:score, timestamp=1607880087191, value=265
5249623960609831 column=info:transaction_date, timestamp=1607880087191, value=2018-01-28 00:54:29
5252551880815473 column=info:UCL, timestamp=1607880086480, value=11540779.75
5252551880815473 column=info:card_id, timestamp=1607880086480, value=5252551880815473
5252551880815473 column=info:postcode, timestamp=1607880086480, value=39352
5252551880815473 column=info:score, timestamp=1607880086480, value=449
5252551880815473 column=info:transaction_date, timestamp=1607880086480, value=2018-02-01 10:14:39
5253084214148600 column=info:UCL, timestamp=1607880087349, value=13198338.6
5253084214148600 column=info:card_id, timestamp=1607880087349, value=5253084214148600
5253084214148600 column=info:postcode, timestamp=1607880087349, value=78054
5253084214148600 column=info:score, timestamp=1607880087349, value=512
5253084214148600 column=info:transaction_date, timestamp=1607880087349, value=2018-01-27 10:51:49
5254025009868430 column=info:UCL, timestamp=1607880087698, value=14556419.87
```