Transaction Analysis Script for Hadoop and NoSQL

1. Retrieve the most recent 10 transactions for each card.

• First step: Extract the latest 10 transactions for each card from tablee_10transact.

```
df_10trans = spark.sql("\
SELECT card_id, amount, postcode, transaction_dt, status, rn \FROM (\
SELECT card_id, amount, postcode, transaction_dt, status,
ROW_NUMBER() OVER (PARTITION BY card_id ORDER BY \
unix_timestamp(transaction_dt,'dd-MM-yyyy hh:mm:ss') DESC) AS rn \
FROM table_10transact \
WHERE status = 'GENUINE') a \
WHERE a.rn <= 10")</pre>
```

- Next, transform the data into the tb_10trans table.
- >>> df_10trans.createOrReplaceTempView('table_10transact')
- >>> spark.sql('SELECT * FROM table_10transact LIMIT 20').show()

```
>>> spark.sql('SELECT * FROM table 10transact LIMIT 20').show();
         card id| amount|postcode|
                                       transaction dt| status| rn|
|340028465709212|8696557|
                            24658|02-01-2018 03:25:35|GENUINE|
                            58270|15-11-2017 01:59:54|GENUINE|
|340028465709212| 430409|
|340028465709212|6503191|
                            84776|09-11-2017 07:18:21|GENUINE|
|340028465709212|8884049|
                            25537|07-10-2017 09:17:12|GENUINE|
|340028465709212|9291309|
                            31322|12-08-2017 08:29:54|GENUINE|
|340028465709212|8370505|
                            84056|12-07-2017 02:51:29|GENUINE|
|340028465709212|9687739|
                            51542|05-07-2017 11:05:55|GENUINE|
                                                                  71
|340028465709212|6500086|
                            25040|24-06-2017 01:13:31|GENUINE|
                                                                  81
|340028465709212| 581323|
                            46182|17-05-2017 12:36:12|GENUINE|
|340028465709212|5118701|
                            12045|30-03-2017 04:09:10|GENUINE| 10|
|340054675199675|
                            50140|15-01-2018 10:56:43|GENUINE|
|340054675199675|9728785|
                            77373|10-01-2018 02:47:11|GENUINE|
                                                                  21
|340054675199675|2223104|
                            35973|09-01-2018 10:59:10|GENUINE|
|340054675199675|1201277|
                            84530|28-12-2017 05:48:04|GENUINE|
                            40023|18-12-2017 10:33:04|GENUINE|
                                                                  51
|340054675199675|6140357|
|340054675199675|7914699|
                            41844|12-12-2017 07:04:51|GENUINE|
                                                                  61
|340054675199675|7573707|
                            12024|06-12-2017 08:52:38|GENUINE|
|340054675199675|2797924|
                            54141|04-12-2017 12:59:15|GENUINE|
|340054675199675|7876899|
                            71047|27-11-2017 01:54:59|GENUINE|
340054675199675|5418389|
                            21084|05-11-2017 12:00:53|GENUINE|
```

2. Calculate UCL

Following that, we'll compute the Upper Control Limit (UCL) for each card.

```
df_ucl = spark.sql("\
SELECT a.card_id, (a.avge + (3 * a.std)) as UCL \
FROM (\
SELECT t.card_id, AVG(t.amount) AS avge, STDDEV(t.amount) as std \
FROM table_10transact t \
GROUP BY t.card_id) a")
```

• Then, the data is transformed into the UCL_table table.

```
>>> df_ucl.createOrReplaceTempView('UCL_table')
>>> spark.sql('SELECT * FROM UCL_table LIMIT 3').show()
```

```
>>> df_ucl = spark.sql("\
... SELECT a.card_id, (a.avge + (3 * a.std)) as UCL \
... FROM (\
... SELECT t.card_id, AVG(t.amount) AS avge, STDDEV(t.amount) as std \
... FROM table_10transact t \
... GROUP BY t.card_id) a")
>>> df_ucl.createOrReplaceTempView('UCL_table')
>>> spark.sql('SELECT * FROM UCL_table LIMIT 3').show()
+------+
| card_id| UCL|
+-----+
| 340028465709212|1.6685076623853374E7|
| 340054675199675|1.5032693399975928E7|
| 340082915339645|1.5323729774843596E7|
+------+
```

3. Insert data for look-up table

• Finally, join those 2 tables with tb_card and tb_score to insert data into look-uptable

```
spark.sql("INSERT INTO TABLE lookup_table \
SELECT trans.card_id, ucl.ucl, trans.postcode,
trans.transaction_dt, CAST(cdsc.score as double)\
FROM table_10transact trans \
JOIN UCL_table ucl \
ON ucl.card_id = trans.card_id \
JOIN (\
SELECT DISTINCT crd.card_id, scr.score \
FROM card_mem crd \
JOIN mem_score scr \
ON crd.member_id = scr.member_id) AS cdsc \
ON trans.card_id = cdsc.card_id \
WHERE trans.rn = 1")
```

Here main reason for type casting the score was by default the score was in the string format.

```
>>> spark.sql("INSERT INTO TABLE lookup table \
... SELECT trans.card id, ucl.ucl, trans.postcode, trans.transaction_dt, cdsc.score \
... FROM table | Oltransact trans \
... JOIN UCL table ucl \
... On ucl.card id = trans.card id \
... JOIN (CL table ucl \
... JOIN (C. SELECT DISTINCT crd.card_id, scr.score \
... FROM card mem crd \
... JOIN (C. SELECT DISTINCT crd.card_id, scr.score \
... FROM card mem crd \
... JOIN mem score scr \
... ON ord.member id = scr.member id) AS cdsc \
... ON ord.member id = scr.member id) AS cdsc \
... ON trans.card_id = cdsc.card_id \
... WHERE trans.rm = !")
Traceback (most recent call last):
File "\stdin>", line 1, in \text{module} \
File "\stdin>", line 1, in \text{module} \
File "\stdin>", service \
File "\stdin>", line 1, in \text{module} \
File "\stdin>", service \
File "\stdin>", line 1, in \text{module} \
File "\stdin>", service \
File "\stdin>", line 1, in \text{module} \
File "\stdin>", line 1, in \text{module} \
File "\stdin>", service \
File "\stdin>", line 1, in \text{module} \
File "\stdin>", line 1, in \tex
```

>>> spark.sql('SELECT * FROM lookup_table LIMIT 3').show()

```
SELECT trans.card_id, ucl.ucl, trans.postcode, trans.transaction_dt, CAST(cdsc.score as double)
   FROM table_10transact trans \
   JOIN UCL_table ucl \
ON ucl.card_id = trans.card_id \
   JOIN (\
SELECT DISTINCT crd.card_id, scr.score \
   FROM card mem crd \
   JOIN mem score scr
... ON crd.member id = scr.member id) AS cdsc \
... ON trans.card_id = cdsc.card_id \
   WHERE trans.\overline{rn} = 1")
>>> spark.sql('SELECT * FROM lookup_table LIMIT 3').show()
         card id|
                                      ucl|postcode|
                                                          transaction dt|score|
340028465709212|1.6685076623853374E7|
                                              24658|02-01-2018 03:25:35|
                                                                              2331
340054675199675|1.5032693399975928E7|
340082915339645|1.5323729774843596E7|
                                              41754|03-11-2017 09:06:10|
```

4. Saving lookup table

Save the lookup table into MongoDB

```
lookup_df = sqlContext.sql("SELECT * FROM lookup_table")
lookup_df.coalesce(1).write.format('csv').options(header='Tru
e',delimiter=',').mode('overwrite').save("s3://creditcardcaps
tone/lookup_table")
```

<pre>>>> lookup df = sqlContext.sql(": >>> lookup_df.coalesce(1).write</pre>	SELECT * FROM lookup_table") format('csv').options(header='Tr	ue',delimiter=',').mode('ove	rwrite').save("s3://creditcardcaps	tone/lookup_table")
	X	X-X-X		