## Logic Final

- 1) Import all necessary libraries and functions.
- 2) Define spark context and add .py files required along with csv given in resources list.
- Connect to kafka topic using topic name "transactions-topic-verified" and server 18.211.252.152:9092.
- 4) Read the kafka stream into appropriate schema to make data readable.
- 5) Look Up Table Name: look up table

Card Transaction table Name: card\_transactions

6) Define following user defined functions to perform activities required for rule execution and determine if transaction is fraudulent or genuine.

a. Name of function: ucl\_data

Input: CARD\_ID

Output: UCL from look\_up\_table

b. Name of function: score\_data

Input: CARD\_ID

Output: Credit Score from look up table.

c. Name of function: postcode\_data

Input: card\_id

Output: post code from look up table.

d. Name of function: distance\_calc

Input: post codes from lookup table & kafka stream.

Output: Distance between 2 locations of current transaction and previous

transaction.

e. Name of function: time\_cal

Input: transaction date from lookup table & kafka stream

Output: difference between transaction dates in seconds.

f. Name of function: ITransD\_data

Input: CARD\_ID

Output: transaction date from look up table.

g. Name of function: speed\_calc

Input: Distance & Time calculated from above distance\_calc & time\_cal functions Output: Speed which is mathematically calculated by multiplying distance \* 1000 and dividing by time.

h. Name of function: status res

Input: Amount from current transaction read thru kafka stream, UCL from look up table, Credit\_Score from look up table & Speed calculated from user defined functions.

Output: Status as transaction if its genuine or fraud.

- 7) Execute above user define functions in same order given above. These functions perform us all required logic to deduce if transaction is fraud or genuine. These functions are agents to derive inputs to function status\_res (function H).
- 8) Here are the rules performed on top of inputs supplied to function H.
  - a. If current transaction amount is greater than UCL of look up table for that card\_id, mark transaction as Fraud. Else, proceed to check below:
    - i. If credit score of that card\_id under process is less than 250, reject transaction as FRAUD. Else, proceed to check below:

- 1. If speed calculated is greater than 250, recognize the transaction as "FRAUD". If speed is between 0 and 250, mark the transaction as genuine.
- 9) To summarize, a transaction is qualified to be genuine only when:
  - a. Credit score of member is greater than 200,
  - b. Speed is between 0 & 250
  - c. Amount on current transaction is less than UCL calculated.
- 10) Functions "A", "B", "C", "F" & "H" contact dao.py to call the look up table (table details given in point 5 above) for designated purposes.
  - In process of calling dao.py from this driver.py file, I followed approach called "Import" which loads other .py files in same directory.
  - Establish a spark context to add python files and csv files before we put the command import.
- 11) Function "D" uses geomap.py to calculate distance between last transaction & current transaction locations. This is in turn used in calculating speed which is one of factors for determining status of transaction.
- 12) Function "H" status\_res also calls look\_up\_table using write\_data function when transaction is genuine.
  - Apart from this, it updates card\_transactions table with latest information of posid, amount, transaction date and member ID.

```
atch: 0
card_id
               |member_id
                             |amount |pos_id
                                                      |postcode|transaction_dt_ts |status |
348702330256514 |37495066290 |4380912|248063406800722|96774
                                                                [2017-12-31 08:24:29|GENUINE|
348702330256514 |37495066290
                                                               |2017-12-31 04:15:03|FRAUD
                             |7454328|466952571393508|93645
                                                                |2017-12-31 09:56:42|GENUINE|
348702330256514 |37495066290
                             |4013428|45845320330319 |15868
                                                                |2017-12-31 05:38:54|GENUINE
348702330256514 |37495066290
                             [5495353]545499621965697[79033
                                                               |2017-12-31 21:51:54|GENUINE|
348702330256514 |37495066290
                             |3966214|369266342272501|22832
                                                                |2017-12-31 03:52:51|GENUINE|
                                                                |2017-12-31 00:11:30|FRAUD
348702330256514 |37495066290
                             |1753644|9475029292671 |17923
                                                                |2017-12-31 17:02:39|GENUINE|
                                                                |2017-12-31 20:22:10|GENUINE
                                                                |2017-12-31 01:52:32|FRAUD
                                                                |2017-12-31 05:20:53|FRAUD
                                                                |2017-12-31 14:29:38|GENUINE
5189563368503974|117826301530 |9142237|564240259678903|50635
                                                                |2017-12-31 19:37:19|GENUINE
5407073344486464|1147922084344|6885448|887913906711117|59031
                                                                |2017-12-31 07:53:53|FRAUD
                                                                |2017-12-31 01:06:50|FRAUD
5407073344486464|1147922084344|4028209|116266051118182|80118
5407073344486464|1147922084344|3858369|896105817613325|53820
                                                                |2017-12-31 17:37:26|GENUINE|
                                                                |2017-12-31 04:50:16|FRAUD
                                                                |2017-12-31 13:09:34|GENUINE|
5407073344486464|1147922084344|9492531|211980095659371|49453
                                                                |2017-12-31 14:12:26|GENUINE|
5407073344486464|1147922084344|7550074|345533088112099|15030
                                                               |2017-12-31 02:34:52|FRAUD
nly showing top 20 rows
```





```
Gurrent count: 20000, row: 27999
Gurrent count: 21000, row: 28899
Gurrent count: 22000, row: 29799
 urrent count: 23000, row: 30698
Current count: 24000, row: 31598
Current count: 25000, row: 32498
Current count: 28000, row: 346618652451637.540752175696215.29-04-2018005259.2021-01-04171400.227023
Current count: 29000, row: 35264
Current count: 30000, row: 36164
Current count: 31000, row: 370582035866789.433646648625434.08-07-2018034337.2021-01-04171349.489639
Current count: 32000, row: 375806375521605.880937166605469.26-05-2018130045.2021-01-04171430.733012
Current count: 33000, row: 38176
Current count: 35000, row: 39977
Current count: 36000, row: 40768
urrent count: 37000, row: 41560
 urrent count: 38000, row: 42387
Current count: 40000, row: 43999
Current count: 41000, row: 44784
Current count: 43000, row: 46306
Current count: 44000, row: 47134
 urrent count: 45000, row: 47925
Current count: 46000, row: 48730
Current count: 47000, row: 49500
urrent count: 48000, row: 50351
Durrent count: 51000, row: 5257502990314019.205172644364018.14-07-2018070014.2021-01-04171327.867742
urrent count: 54000, row: 6211
Current count: 55000, row: 6478888441720966.273246841077378.06-10-2018212851.2021-01-04171333.585477
Current count: 55000, row: 64/8
Current count: 56000, row: 7868
Current count: 57000, row: 7868
Current count: 58000, row: 8768
Current count: 59000, row: 9668
  59367
```