



CREDIT CARD FRAUD DETECTION – REAL TIME STREAMING

PROBLEM APPROACH

- 1. Read the source card transactions data from Kafka topic (**transactions-topic-verified**) and define the proper data structure / schema on top of the streaming dataframe.
- Derive the column "status" with two distinct values as "GENUINE" and "FRAUD" based on the following three rules:
 - a. <u>Credit Score:</u> Get the credit score from card lookup HBase table based on card id fetched from the source transactions data and validate it. If the score is less than 200 then label it as "FRAUD" and move to next transaction. Otherwise proceed validating the second rule.
 - b. Check Upper Control Limit (UCL): Get the amount from the transaction data and compare it with UCL value present in card lookup HBase table. If the amount is greater than UCL then label it as "FRAUD" and move to next transaction. Otherwise proceed validating the third and final rule.
 - c. <u>Zip code Distance:</u> Here we calculate the distance travelled between two locations of the recent two transactions and its time difference and measure speed in terms of Kilometers/Second. We set the threshold value to 0.25 Kilometers/Second (900 Kilometers/Hour average speed of commercial flights). If the speed measured is greater than threshold value, then label it as "FRAUD". Otherwise, it can be treated as "GENUINE".
- 3. Once the transaction status is derived, we can insert the card transactions source data along with its status in card transactions HBase table.
- 4. Update the postcode and transaction date in card lookup HBase table for the transactions happened only if the status is labelled as "GENUINE".
- 5. After processing all the transactions within the same micro batch, display the card transactions information with its status in the console for live analysis.

CODE IMPLEMENTATION

1. Import the necessary libraries related to PySpark, Spark Streaming, classes and functions definitions provided as boilerplate code (dao.py, geo_map.py and rules.py).

```
from datetime import datetime
from pyspark.sql import SparkSession
from pyspark.sql.functions import col, from_json, to_timestamp, udf
from pyspark.sql.types import StructType, StructField, StringType, IntegerType, LongType, TimestampType
from db.dao import HBaseDao
from db.geo_map import GEO_Map
from rules.rules import is_score_rejected, is_ucl_exceeded, is_zipcode_invalid
```





- 2. Define the required utility functions and UDF on top of it.
 - a. <u>get_distance(): -</u> This function is to calculate the distance between current and previous locations where transactions happened based on its latitude and longitude. It instantiates and calls GEO_Map class (geo_map.py) methods to get the latitude and longitude attributes value and calculate the distance in Kilometers.

geo_map.py

```
import math
import pands as pd

class GEO Map():
    """
    it hold the map for zip code and its latitute and longitude
    """
    inntance = None
    Staticmsthod
    def get_interace():
        if GEO Map_ instance = None:
        GEO Map_ instance = None:
        if GEO Map_ instance

def _init_ (self):
    if GEO Map_ instance

def _init_ (self):
    if GEO Map_ instance = None:
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```

driver.py

b. **<u>get time diff secs(): -</u>** This function calculates the time difference between current and previous transactions timestamp.

driver.py

```
### Utility function to calculate time difference between last two transactions happened

def get_time_diff_secs(trans_transactiondt, lkp_transactiondt):

Calculate the time difference between current and previous transactions timestamp.

Input Arguments: trans_transactiondt (transaction dt from card_transactions real time data - Kafka topic in JSON format)

lkp_transactiondt (transaction_dt from card_lookup table (Hbase) of the same card_id from card_transactions real time data).

return (trans_transactiondt - lkp_transactiondt).total_seconds()
```





c. <u>insert_card_transactions_hbase(): -</u> This function inserts card transactions real time data (Kafka topic in JSON format) into card_transactions HBase table. It first instantiates HbaseDao class object to connect to HBase database and then call required methods to insert the data in card transactions table.

dao.py

```
def write_data(self, key, row, table):
    error = None
    for i in range(2):
        try:
        with self.pool.connection() as connection:
            t = connection.table(table)
            return t.put(key, row)
    except Exception as e:
        error = e
        self.reconnect()
    raise Exception(key + str(e))

def reconnect(self):
    self.connect()
```





driver.py

d. <u>update_card_lookup_hbase(): -</u> This function updates postcode and transaction_dt attributes in card lookup table HBase for which transactions are treated as GENUINE. It first instantiates HbaseDao class object to connect to HBase database and then call required methods to update the card lookup data.

driver.py

e. <u>check_status():</u> This function is to detect the fradulent of the live transactions based on the rules defined (mentioned in the problem approach) and lookup data prepared as part of batch processing.

rules.py

```
This function is to get the credit score of the particular card which transaction happened and see whether it has greater than or equal to 200. If not then it is labelled as Fraud.

Input Aryments: score (credit_score from card_transactions real time data - Kafka topic in JSON format)

if score < 200:
    return True
    return false

def is wcl_exceeded(amount, ucl):

This function is to check whether the amount exceeds the upper control limit of the transaction to be done. If it is exceeded then it is labelled as Fraud.
    Input Aryments: amount (card_transactions real time data - Kafka topic in JSON format)

if amount > ucl. (OCL_attribute value from card_lookup table - HBANE)

if amount > ucl. (OCL_attribute value from card_lookup table - HBANE)

if amount > ucl.

return False

def is ripcode_invalid(distance, time_diff_secs):

This function is to measure the speed in Kilometers/second travelled by the customer between two locations where customer made the recent two transactions (or greats and provived) is to measure the speed in Kilometers/second. If the measure speed between two locations is greater than 0.25 Kilometers/second then it is labelled as Fraud.

Input Aryments: distance (distance convered between two locations and provived)

if time_diff_secs == 0:
    return True

else:

speed_kmps = distance / time_diff_secs

if speed_kmps = distance / time_diff_secs
```





driver.py

f. Define UDF (fraud_detection) on top of check_status() utility function that to be used in spark streaming dataframe for deriving "status" attribute during real time. As part of this UDF, it insert the records into card transactions HBase table and update the postcode and transaction date in card lookup HBase table (as shown in check_status() driver.py screenshot above).

driver.py

```
# Create fraud detection UDF on top of status utility function
fraud_detection = udf(check_status, StringType())
```

Create a Spark Session which is an entry point to spark application and set the log level to 'ERROR'.

driver.py





4. Read the card transactions real time data from Kafka source topic (transactions-topic-verified).

driver.py

5. Define the proper data structure or schema and assign it to the source data read from Kafka topic.

driver.py

6. Add a new column "status" by calling the fraud_detection UDF defined above. This UDF will also take of inserting and updating the records in card transactions and card lookup HBase tables.

driver.py





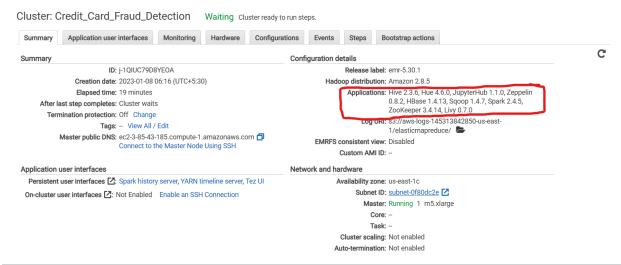
 Write the final transformed card transactions data with its status in the console for live analysis. It runs as a micro-batch with processing time of every "1 Minute" until it is interrupted/terminated.

driver.py

DEPLOYMENT PLAN

PRE-REQUISITES:

 Create an EMR cluster with Hive, HBase, Sqoop, Spark, Zookeeper applications installed in it as shown below:



Applications such as Hue, JupyterHub, Zeppelin, Livy are optional and depends based on your preference.

Download and install MySQL connector in AWS EMR for performing Sqoop data ingestion from AWS RDS to HDFS.

cd /home/hadoop

wget https://de-mysql-connector.s3.amazonaws.com/mysql-connector-java-8.0.25.tar.gz

tar -xvf mysql-connector-java-8.0.25.tar.gz





cd mysql-connector-java-8.0.25/ sudo cp mysql-connector-java-8.0.25.jar /usr/lib/sqoop/lib/

Download and install Hive-HBase handler jar file which will be used for card lookup data
preparation for Hive and Hbase tables using PySpark. and place it in spark libraries
folder. Also place some other required HBase jar files from HBase libraries to spark
libraries folder.

cd /home/hadoop

wget https://repo1.maven.org/maven2/org/apache/hive/hive-hbase-handler/3.1.3/hive-hbase-handler-3.1.3.jar sudo cp hive-hbase-handler-3.1.3.jar /usr/lib/spark/jars/ sudo cp /usr/lib/hbase/*.jar /usr/lib/spark/jars/ sudo cp /usr/lib/hbase/lib/htrace-core-3.1.0-incubating.jar /usr/lib/spark/jars/ sudo cp /usr/lib/hbase/lib/guava-12.0.1.jar /usr/lib/spark/jars/ sudo cp /usr/lib/hbase/lib/metrics-core-2.2.0.jar /usr/lib/spark/jars/

4. Create all the required tables in Hive and HBase such as card_member, member_score, card_transactions (historical data) and card_lookup tables and load the pre-processed data in it as part of batch processing (mid submission).

HBase:

```
hbase(main):001:0> list
TABLE
card_lookup
card_transactions
2 row(s) in 0.1870 seconds

=> ["card_lookup", "card_transactions"]
hbase(main):002:0> [
```

Hive:

```
[hadoop@ip-172-31-84-212 ~]$ hive

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

OK

cred_financials_data
default

Time taken: 0.499 seconds, Fetched: 2 row(s)
hive> use cred_financials_data;

OK

Time taken: 0.019 seconds
hive> show tables;

OK

card_lookup

card_member

card_transactions

member_score

Time taken: 0.024 seconds, Fetched: 4 row(s)
hive> []
```





5. Switch to root user in EMR cluster and install pandas and happybase python packages. sudo su root yum install gcc (Press 'Y' and ENTER after it prompts to download) yum install python3-devel (Press 'Y' and ENTER after it prompts to download)

yum install python3-de pip install cython pip install numpy pip install pandas pip install happybase exit

DEPLOYMENT STEPS:

 Login to AWS EMR and create the python directory under home directory and place all the executable python files inside it (either using scp command or WinSCP tool) and change its permissions.

cd /home/hadoop mkdir -p /home/hadoop/python/src

Place the files using scp command from local to EMR

```
C:\Users\Ganesh Jalakam>scp -r -P 22 -i "C:\Users\Ganesh Jalakam\Documents\AWS\Instance_Keys\RHEL.pem" "D:\IIITB_Data_Science\PG_DS_Course\Capstone_Project\Final_Submission\python\*" hadoop@ec2-3-85-43-185.compute-1.amazonaws.com:/home/hado
op/python/
dao.py
geo_map.py
                                                                                                                         100% 1168
                                                                                                                                             4.9KB/s
                                                                                                                                                          00:00
                                                                                                                         100% 1205
                                                                                                                                             4.3KB/s
                                                                                                                                                          00:00
__init__.py
dao.cpython-37.pyc
                                                                                                                         100%
                                                                                                                                             0.0KB/s
                                                                                                                                                          00:00
                                                                                                                         100% 1732
                                                                                                                                             6.7KB/s
                                                                                                                                                          00:00
                                                                                                                         100% 125 0.4KB/s
100% 10KB 44.5KB/s
  _init__.cpython-37.pyc
                                                                                                                                                          00:00
driver.py
                                                                                                                                                          00:00
rules.py
                                                                                                                         100% 2760
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100% 735KB 474.0KB/s
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uszipsv.csv
                                                                                                                                                          00:01
 _init__.py
                                                                                                                         100%
                                                                                                                                            0.0KB/s
                                                                                                                                                          00:00
C:\Users\Ganesh Jalakam
```

Change the executable files permissions: cd /home/hadoop/python/src chmod 755 *.py chmod 755 db/*.py chmod 755 rules/*.py

driver.py - /home/hadoop/python/src

```
[hadoop@ip-172-31-84-212 src]$ pwd
/home/hadoop/python/src
[hadoop@ip-172-31-84-212 src]$ ls -ltr
total 748
drwxrwxrwx 3 hadoop hadoop 76 Jan 8 01:35 dh
-rwxr-xr-x 1 hadoop hadoop 10319 Jan 8 01:36 driver.py
drwxrwxrwx 2 hadoop hadoop 41 Jan 8 01:36 mules
-rw-rw-r-- 1 hadoop hadoop 752688 Jan 8 01:36 uszipsv.csv
-rwxr-xr-x 1 hadoop hadoop 0 Jan 8 01:36 __init__.py
[hadoop@ip-172-31-84-212 src]$ [
```





dao.py and geo_map.py - /home/hadoop/python/src/db

```
/home/hadoop/python/src/db
[hadoop@ip-172-31-84-212 db]$ ls -ltr
total 8
-rwxr-xr-x 1 hadoop hadoop 1168 Jan 8 01:35 dao.py
-rwxr-xr-x 1 hadoop hadoop 1205 Jan 8 01:35 geo_map.py
-rwxr-xr-x 1 hadoop hadoop 0 Jan 8 01:35 init .py
drwxrwxrwx 2 hadoop hadoop 63 Jan 8 01:35 pycache
[hadoop@ip-172-31-84-212 db]$ [
```

rules.py - /home/hadoop/python/src/rules

```
/home/hadoop/python/src/rules
[hadoop@ip-172-31-84-212 rules]$ ls -ltr
total 4
-rwxr-xr-x 1 hadoop hadoop 2760 Jan 8 01:36 rules.py
-rwxr-xr-x 1 hadoop hadoop 0 Jan 8 01:36 __init__.py
[hadoop@ip-172-31-84-212 rules]$ [
```

2. Zip all the python dependencies into one file with name src.zip (containing dao.py, geo_map.py and rules.py in it).

```
cd /home/hadoop/python/src zip src.zip init .py rules/* db/*
```

```
[hadoop@ip-172-31-84-212 src]$ zip src.zip __init__.py_rules/* db/*
 adding: __init__.py (stored 0%)
 adding: rules/ init .py (stored 0%)
 adding: rules/rules.py (deflated 68%)
 adding: db/dao.py (deflated 61%)
 adding: db/geo map.py (deflated 56%)
 adding: db/ init .py (stored 0%)
 adding: db/ pycache / (stored 0%)
[hadoop@ip-172-31-84-212 src]$ ls -ltr
total 752
drwxrwxrwx 3 hadoop hadoop 76 Jan 8 01:35
-rwxr-xr-x 1 hadoop hadoop 10319 Jan 8 01:36 \overline{\text{driver.py}}
drwxrwxrwx 2 hadoop hadoop 41 Jan 8 01:36 <mark>ru</mark>
-rw-rw-r-- 1 hadoop hadoop 752688 Jan 8 01:36 uszipsv.csv
-rwxr-xr-x 1 hadoop hadoop 0 Jan 8 01:36 init .py
rw-rw-r-- 1 hadoop hadoop 2968 Jan 8 01:53 src.zip
```

 Run the following command to enable Kafka Integration with Apache Spark. export SPARK_KAFKA_VERSION=0.10

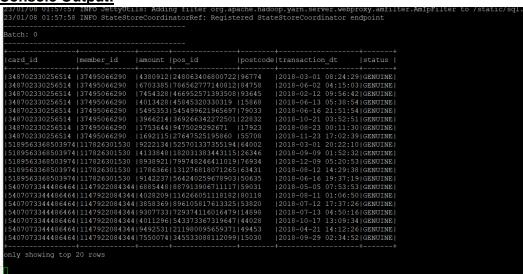




4. Run driver.py script to start spark streaming job which consumes real time transactions data and detects the status (FRAUD or GENUINE) and display the final output of each micro batches in the console.

spark-submit --py-files src.zip --packages org.apache.spark:spark-sql-kafka-0-10_2.11:2.4.5 --files uszipsv.csv driver.py

Console Output:



Job runs continuously as micro batches with 1 Minute time interval until it is terminated/interrupted.

VALIDATION

1. Check the card transactions count before and after streaming the data.

HBase: - count 'card_transactions'

BEFORE STREAMING - 53292 records







AFTER STREAMING - 59367 records

```
ER STREAMING — 59367 records

t count: 19000, row: 4264553579186587|470021900536700|4272858|47578|552792990977827|2017-02-13 04:27:36|GENUINE t count: 2000, row: 4367125470626536|179997837474045|5811475|32535|9451133290339084|2017-02-28 11:06:14|GENUINE t count: 2000, row: 431129964699133|122907349228520|1978460|146986|391311742932204|2018-01-18 16:11:30|GENUINE t count: 2000, row: 4492848090|58931|613838524128130|3001807|71043|366296280375812|2017-0-04 01:34:23|GENUINE t count: 2000, row: 459374417702905|64256046641301|199912|16921|189567397707489|2016-10-14 16:24:41|GENUINE t count: 2000, row: 4585507001604561|44791746765587|11078211|99136|888926523445515|2016-0-15 16:39:06|GENUINE t count: 26000, row: 4692371899772166|30603646617496|4422540|61064|639432391799127|2017-10-07 12:47:53|GENUINE t count: 26000, row: 4719418574936017|500004043659347|1912413|54859|76582549|39249|2016-10-14 19:13:13|6GENUINE t count: 26000, row: 470287944662468|257134899293254|331180|30738|559036431318991|2017-06-22 22:53:28|GENUINE t count: 28000, row: 4805556006187980|4580289293254|331180|30738|559036431318991|2017-06-22 22:53:28|GENUINE t count: 28000, row: 48055556006187980|4582929254|33128189|30738|559036431318991|2017-06-22 22:53:28|GENUINE t count: 30000, row: 498154824137027|9964116335289270|5381248|9377|332218533399343|2017-04-10 03:33:35|GENUINE t count: 30000, row: 498154824137027|9964116335289270|5381248|9377|332218533399343|2017-04-12 03:33:35|GENUINE t count: 30000, row: 51553366114315751|403808871228322|5720647|3847|55900533339132|2016-04-21 00:33:35|GENUINE t count: 30000, row: 5203156036333804|77956663578599|7781043|550450443739|2018-02-11 00:00:00|GENUINE t count: 30000, row: 52041232737297200|133086195760104|9026991|13319|21597263091597|71071-01-10 07:56:45|GENUINE t count: 30000, row: 5204423378793943|191324569045987599|7781451554504443739|2018-02-11 00:00:00|GENUINE t count: 30000, row: 520442336937693893949|19024764595799|789143|5594504443739|2018-02-10 00:00:00|GENUINE t count: 30000, row: 52044233693789859|134
```

<u>Hive: -</u> select status, count(*) from card_transactions group by status;

BEFORE STREAMING

Status _c1 FRAUD 82 GENUINE 53210

```
Query ID = hadoop_20230108012501_db1b7cd1-2a03-468e-820a-1d2ba880cd36
aunching Job 1 out of 1
Status: Running (Executing on YARN cluster with App id application_1673139411187 0014)
      VERTICES
                  MODE
                               STATUS TOTAL COMPLETED RUNNING PENDING FAILED KILLED
Map 1 ..... container
Reducer 2 ..... container
ENUINE 53210
```

AFTER STREAMING

status _c1 212 FRAUD GENUINE 59155





2. Check the latest transaction date available in card_transactions table.

<u>Hive: -</u> select max(transaction_dt) from card_transactions; BEFORE STREAMING - 2018-02-11 00:00:09

AFTER STREAMING - 2018-12-10 09:00:34