<u>Credit Card Fraud Detection System – Solution Approach</u>

As per the guidelines provided in the Problem Statement, the solution is broken down into 07 tasks numbered 1 to 7.

Tasks 1 to 4 are designed using the batch-processing approach and Tasks 5 to 7 are designed using the real-time stream processing approach.

Let's walkthrough the solution.

Task 1: Load the transactions history data (card_transactions.csv) in a NoSQL database and create a look-up table with columns specified earlier in the problem statement in it.

Solution 1:

- 1. Using Hbase Shell, create an Hbase table named **card_transactions** with a column family named **cf1** (See, hbase/createTableForCardTransactions.sql)
- 2. Using Hive Editor in Hue, create a Hive External table named **card_transactions** in the **default** Hive database. This Hive table is mapped with the Hbase table **card_transactions**. (See, hive/createTableForCardTransactions.hql)
- 3. Using Hive Editor in Hue, create a Hive External table named **stagging** in the **default** Hive database. This is a stagging table to load data from the **card_transactions.csv** file. (See, hive/createTableForCardTransactions.hql)

Pre-requisites to step 4: Copy the card_transactions.csv to a location on HDFS (See, hdfs/copyCard_TransactionsCSV.txt)

- 4. Load the data of **card_transactions.csv** in the Hive table named **stagging**. (See, hive/loadCardTransactions.hql)
- 5. Load the entire data from the Hive table **stagging** into the Hive table **card_transactions** (See, hive/loadCardTransactions.hql)

Task 2: Write a script to ingest the relevant data from AWS RDS to Hadoop.

Solution 2:

Pre-requisites to step 1: Setup some HDFS directories with relevant permissions. (See, hdfs/setupDirForDataIngestion.txt)

- 1. *Create an External Hive table named* **card_member** in the **default** Hive database. (See, hive/createTableCardMember.hql)
- 2. Create a Sqoop job named **GetCardMember** to ingest the data from the AWS RDS table **card_member** to the Hive table **card_member**. This is an incremental sqoop import job. (See, sqoop/ingestCardMemberData.sqoop)
- 3. *Create an External Hive table named* **member_score** in the **default** Hive database. (See, hive/createTableMemberScore.hql)
- 4. Create a Sqoop job named **GetMemberScore** to ingest the data from the AWS RDS table **member_score** to the Hive table **member_score**. (See, sqoop/ingestMemberScoreData.sqoop)

Note: These scoop jobs will be executed as part of oozie workflow.

Task 3: Write a script to calculate the moving average and standard deviation of the last 10 transactions for each card_id for the data present in Hadoop and NoSQL database. If the total number of transactions for a particular card_id is less than 10, then calculate the parameters based on the total number of records available for that card_id. The script should be able to extract and feed the other relevant data ('postcode', 'transaction_dt', 'score', etc.) for the look-up table along with card_id and UCL.

Solution 3:

- 1. Using Hbase Shell, create an Hbase table named **look_up** with a column family **cf1** (See, hbase/createTableForLookUp.sql)
- 2. Using Hive Editor in Hue, create a Hive External table named **look_up** in the **default** Hive database. This Hive table is mapped with the Hbase table **look_up**. (See, hive/createTableLookUp.hql)
- 3. Using Hive Editor in Hue, create a Hive External table named **query_stagging** in the **default** Hive database. This is a stagging table which stores the calculated **UCL** and associated **card_id**. (See, hive/createTableQueryStagging.hql)
- 4. Insert data into the Hive table named **query_stagging** using the Hive query which calculates **UCL**. (See, hive/loadQueryStagging.hql)
- 5. Load data into the Hive table **look_up** by performing a join on the following tables:
- a) card_transactions
- b) query_stagging
- c) *member_score*

(See, hive/loadLookUp.hql)

Task 4: Set up a job scheduler to schedule the scripts run after every 4 hours. The job should take the data from the NoSQL database and AWS RDS and perform the relevant analyses as per the rules and should feed the data in the look-up table.

Solution 4:

- 1. In order to perform the analyis i.e. calculate UCL and update the look_up table, an oozie scheduler is written with the following components:
- a) Workflow
- b) Coordinator
- c) Job properties
- 2. The workflow spawns 02 parallel sqoop actions to fetch data from card_member and member_score AWS RDS tables respectively:
- a) sqoopjob_qetCardMember
- b) sqoopjob_getMemberScore
- 3. The above 02 scoop actions are joined into a Hive query action **hivejob_calculateUCL** which calculates the UCL. Once this scoop action is completed, the Hive query action **hivejob_loadLookupTable** is spawned.
- 4. The Hive query action **hivejob_loadLookupTable** updates the **look_up** table with UCL and other details for a card_id.
- 5. The workflow is scheduled to execute after every 04 hours. The coordinator is configured for this purpose.

(See, oozie/)

For **Tasks 5, 6** and **7**, a Spark streaming application is designed and developed which will consume real-time streaming data from a Kafka Server. This transaction data will be consumed and processed to validate if a particular transaction is **GENUINE** or **FRAUD** based on **03** parameters defined in the problem statement. The categorised transaction data will then be written to the **card_transactions** table available in a NOSQL database.

Spark Streaming Application Structure

- 1. RealtimeFraudDetectionApplication.java
 - ➤ This is the main class from where the Spark Streaming application is launched.
- 2. TransactionPOJO.java
 - ➤ This is a POJO (Plain Old Java Object) class which holds transaction data received from the Kafka server. It contains private member variables which correspond to the transaction details like card_id, member_id, amount etc. and a set of public getter/setter methods for accessing/modifying the transaction details.
- 3. TransactionDAO.java
 - ➤ This is a DAO (Data Access Object) class which contains methods for connecting to the NOSQL database and other DML (Data Manipulation Language) methods for reading and writing data to the database
- 4. ZipCodeDistance.java
 - ➤ This is a utility to calculate the distance in Kilometers between 02 post codes i.e. post code of the current transaction and the last transaction of a card holder
- 5. ZipCode.java
 - ➤ This is a POJO which holds zip code data information for a post code. It contains private member variables which correspond to the post code details like latitude, longitude etc. and a set of public getter/setter methods for accessing/modifying the post code details.

Spark Streaming Application Logic Flow

1. Initialize The Spark Streaming Application

- Create and initialize a Spark Configuration object to run the application in local mode
- ➤ Create a HashMap object to store NOSQL database parameters received from command line arguments. This information is passed to the **TransactionDAO** class via the **initializeTransactionDAO()** method.
- > Create a **JavaStreamingContext** object to create DStreams with an interval of 1 second.
- ➤ Create a HashMap object to store Kafka parameters to be used for connecting to the Kafka Server
- ➤ Using the **createDirectStream()** method of the **KafkaUtil** class, create a JavaInputDStream which will contain transactions data (as JSON strings) as RDDs within DStreams

2. Consume And Process The Transaction Data

- Create a FlatMapFunction() to process the transactions data received as JSON strings in RDDs of Dstreams
- ➤ Using the **toJSON()** method of the **JSONSerializer** class, parse the JSON strings as JSON objects.
- > Retrieve and store the transaction data from the JSON objects using the **get()** method.
- ➤ For each record (transaction), create an object of **TransactionPOJO** class and initialise the transaction data (retrived from the JSON object) using the setter methods of the class.

3. Evaluate The Transaction Based On 03 Parameters

- ➤ Rule1: Transaction Amount should be less than or equal to UCL
 - Using the getUCL() method of the TransactionDAO class, retrieve the UCL from the look_up table for the card holder
 - Using the **getAmount()** method of the **TransactionPOJO** class, retrieve the amount from the TransactionPOJO object containing transaction data.
 - Check if transaction amount is less than or equal to UCL. If yes, set **ruleUCL** boolean variable to true.

- ➤ Rule2: Member Score should greater than or equal to 200
 - Using the **getScore()** method of the **TransactionDAO** class, retrieve the member score from the look_up table for the card holder.
 - Check if score is greater than or equal to 200. If yes, set **ruleScore** boolean variable to true.
- ➤ Rule3: Distance travelled between 02 post codes should be greater than 0.25 KM/sec
 - Using the getPostCode() method of the TransactionDAO class, retrieve the zip code of the last transaction from the look_up table for the card holder
 - Using the **getPostCode()** method of the **TransactionPOJO** class, retrieve the zip code of current transaction from the TransactionPOJO object containing transaction data.
 - Using the **getDistanceViaZipCode()** method of the **ZipCodeDistance** class, calculate the distance in Kilometers between the 02 zip codes.
 - Using the getTransactionDate() method of the TransactionDAO class, retrieve the transaction date of the last transaction from the look_up table for the card holder
 - Using the **getTransactionDate()** method of the **TransactionPOJO** class, retrieve the transaction date of current transaction from the TransactionPOJO object containing transaction data.
 - Calculate the difference in time between the last transaction date and current transaction date.
 - Calculate the distance covered in secs using the below formula:

$distance Covered In Secs \ = distance In KM/date Diffence In Secs$

- Check if distanceCoveredInSecs is less than **0.25**. If yes, set **ruleZipCode** boolean variable to true.
- **Note:** In order to deal with incorrect transaction date data i.e. the last transaction date in the look_up table is greater than or newer than the current transaction date, it was proposed to take an absolute value of transaction dates difference to make sure a lot of data does not get labeled as FRAUD. I have taken the suggested and most popular approach.

4. Update The NOSQL Database

- ➤ If all the rules are met, then classify the transaction to be **GENUINE**
 - Using the **setStatus()** method of the **TransactionPOJO** class, set the status as GENUINE.
 - Using the **updateLookUp()** method of the **TransactionDAO** class, update the new post code and transaction date in the look_up table.
- > Else classify it as **FRAUD**
 - Using the **setStatus()** method of the **TransactionPOJO** class, set the status as **FRAUD**.
- ➤ Using the **insertTransaction()** method of the **TransactionDAO** class, insert the new transaction details in the card_transactions table.

Execution Environment Details For Spark Streaming Project:

- 1. This project contains the following JAVA classes:
 - RealtimeFraudDetectionApplication.java
 - > TransactionPOJO.java
 - ➤ TransactionDAO.java
 - ➤ ZipCodeDistance.java
 - ➤ ZipCode.java
- 2. Please make sure JDK1.8 is installed on the machine on which the programs are executed.
- 3. All of these JAVA classes (and a few more supporting classes) will be bundled in the CreditCardFraudDetection-0.0.1-SNAPSHOT-jar-with-dependencies.jar
- 4. Note: This project is tested ONLY
 - ➤ In Local Mode
 - > On LINUX
 - ➤ Please note this program is Not Tested On Windows OS.
- 5. Create a folder named **data** and place the **zipCodePosId.csv** file in it. This folder should be placed at the same location from where the program is executed.
- 6. Below is the syntax of the command for running the programs via command-prompt:

```
java -cp <Absolute file system path of the JAR file> <Fully-qualified Java Class Name> <HOST-NAME> <HBASE_MASTER_PORT> <CLIENT_PORT>
```

➤ Below is a sample example, in this case the JAR file is present on the same location where the command prompt is launched:

java -cp CreditCardFraudDetection-0.0.1-SNAPSHOT-jar-with-dependencies.jar com.upgrad.bigdata.project.RealtimeFraudDetectionApplication "quickstart.cloudera" "60000" " 2181"