Code Logic - Retail Data Analysis

```
#importing necessary libraries and python modules

import sys
import os
from pyspark.sql import SparkSession
from pyspark.sql.functions import *
from pyspark.sql.types import *
from ast import literal_eval
```

Required python modules and libraries are imported here. Literal_eval will be used to convert string from items column into a proper python list of dictionary.

```
# get total cost. It will be arrived at by summing up the unit price and quantity of products.
def get_total_cost(items):
    items = literal_eval(items)
    total cost = 0
    for item in items:
       total_cost += item["unit_price"] * item["quantity"]
   return total cost
# get total items. It will be arrived by summing up the total quantity of the products.
def get_total_items(items):
   items = literal eval(items)
    total items = 0
    for \overline{\text{item}} in \overline{\text{items}}:
        total_items += item["quantity"]
    return total items
# if that order is ORDER or RETURN. Incase the category is ORDER return 1 else i.e., category is RETURN, return 0 for order type.
def type_order(category):
   if category ==
        return 1
    return 0
```

Below are the details of the custom functions:

 get_total_cost(items): This function takes item as an argument and calculate the total cost by summing up the unit price and quantity of products. The formulae to calculate get total cost is:

∑(quantity*unitprice)

2. get_total_items(items): This function takes item as an argument and used to retrieve the total items by summing up total quantity of the products. The formulae to calculate the total item is:

∑(quantity)

- **3.** type_order(category): This function takes category as an argument, and it is used to map type of order. If that order is ORDER or RETURN. In case the category is ORDER return 1 else i.e., category is RETURN, return 0 for order type.
- **4.** type_return(category): This function takes category as an argument, and it is used to map type of order. If that order is ORDER or RETURN. In case the category is ORDER return 0 else i.e., category is RETURN, return 1 for return type.

```
32 # if that order is ORDER or RETURN, Incase the category is ORDER return 0 else i.e., category is RETURN, return 1 for return type.
def type_return(category)
if category == "RETUR"
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            return 1
        return 0
       if len(sys.argv) != 4:
          print("Usage: spark-submit spark-streaming.py <hostname> <port> <topic>")
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            exit(-1)
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       host = sys.argv[1]
        port = sys.argv[2]
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        topic = sys.argv[3]
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        spark = SparkSession \
           .builder \
.appName("RetailDataAnalysis") \
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       .getOrCreate()
spark.sparkContext.setLogLevel('ERROR')
51
52
53
       bootstrap server = host + ":" + port
       lines = spark \
            .readStream
             format("kafka") \
```

In the beginning, the host, port number and topic is received from the command line argument.

host: 18.211.252.152

port: 9092

topic: real-time-project

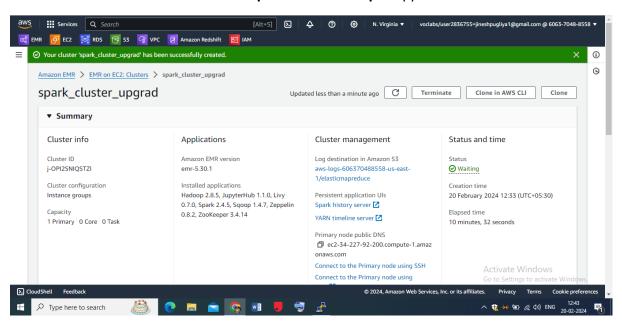
This is used to read the spark stream from kafka bootstrap server.

```
lines = spark \
                   .readStream
                   .format("kafka") \
.option("kafka bootstrap.servers", bootstrap_server) \
.option("subscribe", topic) \
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 60
                   .load()
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            schema = StructType() \
                       = Structtype() \
.add("invoice_no", StringType()) \
.add("country", StringType()) \
.add("timestamp", TimestampType()) \
.add("type", StringType()) \
.add("items", StringType())
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            raw data = lines.selectExpr("cast(value as string)").select(from json("value", schema).alias("temp")).select("temp.*")
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             # create user-defined functions for each
            # Create user-uerthea functions for each
total_cost = udf(lambda items: get_total_cost(items))
total_quantity = udf(lambda items: get_total_items(items))
is_order = udf(lambda types: type_order(types))
is_return = udf(lambda types: type_return(types))
             new_df = raw_data
             new_df = new_df.withColumn("total_cost", total_cost("items")) \
    .withColumn("total_items", total_quantity("items")) \
    .withColumn("is_order", is_order("type")) \
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                                                                                                                                                                                           Go to Settings to activate W
 81
              # create kafka dataframe with invoice number, country, timestamp, total cost, total items, is order and is return
kafkaDF = new_df.select(["invoice_no", "country", "timestamp", "total_cost", "total_items", "is_order", "is_return
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              kafkaDF = kafkaDF.withColumn("total_cost", when(kafkaDF.is_order == 1, kafkaDF.total_cost).otherwise(-kafkaDF.total_cost))
  87
  88
  89
              # streaming raw data
  90
              query0 = kafkaDF.select(["invoice_no", "country", "timestamp", "total_cost", "total_items", "is_order", "is_return"])
 91
 92
              # create time-based KPI with tumbling window of one minute
query1 = kafkaDF.select(["timestamp", "invoice_no", "total_cost", "is_order", "is_return"])
query1 = query1.withWatermark("timestamp", "1 minute").groupBy(window("timestamp", "1 minute")) \
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  95
                            .agg(nound(sum("total_cost"), 2).alias("total_sales_volume"), count("invoice_no").alias("OPM"), \
    round(sum("is_return") / (sum("is_order") + sum("is_return")), 2).alias("rate_of_return"), \
    round(sum("total_cost") / count("invoice_no"), 2).alias("average_transaction_size"))
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              # write stream data to write the time-based KPIs into one minute window each
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              query0 = query0.writeStream \
110
                     .format("console") \
                      .outputMode("append") \
```

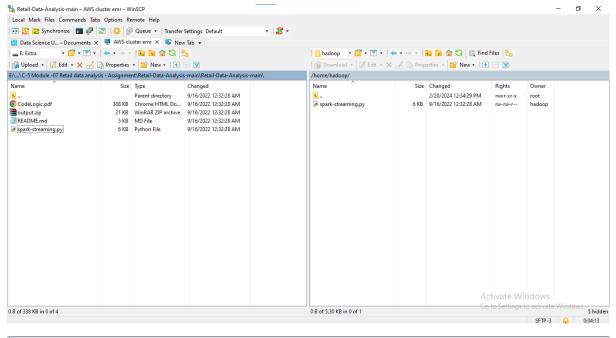
```
| querye = querye.mricestream \
| format("console") \
| outputMode("append") \
| option("truncate", "false") \
| trigger(processingTime="1 minute") \
| start() |
| query1 = query1.writeStream \
| format("json") \
| outputMode("append") \
| option("truncate", "false") \
| option("truncate", "false") \
| option("path", "/user/ec2-user/real-time-project/warehouse/checkpoints1") \
| trigger(processingTime="1 minute") \
| start() |
| query2 = query2.writeStream \
| format("json") \
| outputMode("append") \
| option("truncate", "false") \
| option("truncate", "false") \
| option("path", "/user/ec2-user/real-time-project/warehouse/checkpoints1") \
| trigger(processingTime="1 minute") \
| option("truncate", "false") \
| option("path", "/user/ec2-user/real-time-project/warehouse/checkpoints2") \
| trigger(processingTime="1 minute") \
| start() \
| option("checkpointLocation", "hdfs:///user/ec2-user/real-time-project/warehouse/checkpoints2") \
| trigger(processingTime="1 minute") \
| start() \
| query0.awaitTermination() \
| query0.awaitTermination() \
| query1.awaitTermination() \
| query1.awaitTermination() \
| query2.awaitTermination() \
| query2.awaitTermination() \
| query1.awaitTermination() \
| query2.awaitTermination() \
| query2.awaitTermination() \
| query2.awaitTermination() \
| query1.awaitTermination() \
| query2.awaitTermination() \
| query3.awaitTermination() \
| query3.awaitTermination()
```

Code Deployment and Execution Steps:

1. Create an EMR Instance with Spark and ZooKeeper applications installed in it



2. Copy the python file (spark-streaming.py) under the path: /home/hadoop using WinSCP files transfer application (Windows OS).



- 3. Run the following command to enable Kafka Integration with Apache Spark. export SPARK_KAFKA_VERSION=0.10
- 4. Execute the python file using spark-submit command providing Kafka jar package as an argument. Save the console output in a text file (Console-output.txt).

Bootstrap Server - 18.211.252.152

Port - 9092

Topic - real-time-project

spark-submit --packages org.apache.spark:spark-sql-kafka-0-10_2.11:2.4.5 spark-streaming.py > Console-output.txt

spark-submit --packages org.apache.spark:spark-sql-kafka-0-10_2.11:2.4.5 spark-streaming.py > 18.211.252.152 9092 real-time-project

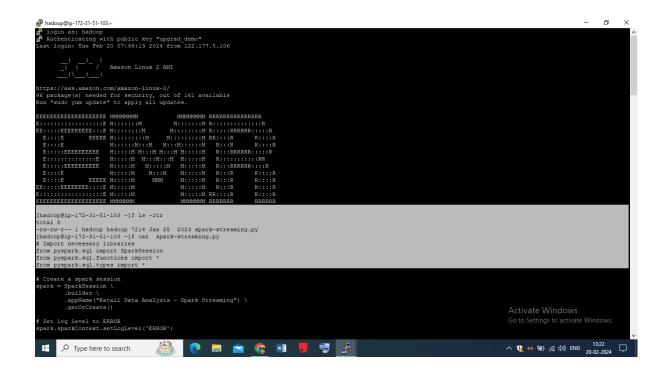
```
hadoop@ip-172-31-60-100:~
   -archives ARCHIVES
                              Comma separated list of archives to be extracted i
nto the
                              working directory of each executor.
                              Principal to be used to login to KDC, while runnin
  --principal PRINCIPAL
g on
                              secure HDFS.
  --keytab KEYTAB
                              The full path to the file that contains the keytab
 for the
                              principal specified above. This keytab will be cop
ied to
                              the node running the Application Master via the Se
cure
                              Distributed Cache, for renewing the login tickets
and the
                              delegation tokens periodically.
[hadoop@ip-172-31-60-100 ~]$ spark-submit --packages org.apache.spark:spark-sql-
kafka-0-10 2.11:2.4.5 spark-streaming.py > Console-output.txt
Ivy Default Cache set to: /home/hadoop/.ivy2/cache
The jars for the packages stored in: /home/hadoop/.ivy2/jars
:: loading settings :: url = jar:file:/usr/lib/spark/jars/ivy-2.4.0.jar!/org/apa
che/ivy/core/settings/ivysettings.xml
org.apache.spark#spark-sql-kafka-0-10 2.11 added as a dependency
:: resolving dependencies :: org.apache.spark#spark-submit-parent-c257e40a-ead7-
4clc-b419-3187d34f6202;1.0
        confs: [default]
        found org.apache.spark#spark-sql-kafka-0-10 2.11;2.4.5 in central
        found org.apache.kafka#kafka-clients;2.0.0 in central
        found org.1z4#1z4-java;1.4.0 in central
        found org.xerial.snappy#snappy-java;1.1.7.3 in central
        found org.slf4j#slf4j-api;1.7.16 in central
        found org.spark-project.spark#unused;1.0.0 in central
:: resolution report :: resolve 412ms :: artifacts dl 11ms
        :: modules in use:
        org.apache.kafka#kafka-clients; 2.0.0 from central in [default]
        org.apache.spark#spark-sql-kafka-0-10 2.11;2.4.5 from central in [defaul
t]
        org.lz4#lz4-java;1.4.0 from central in [default]
        org.slf4j#slf4j-api;1.7.16 from central in [default]
        org.spark-project.spark#unused;1.0.0 from central in [default]
        org.xerial.snappy#snappy-java;1.1.7.3 from central in [default]
                                        modules
                                                           ш
                                                                artifacts
                conf
                           | number| search|dwnlded|evicted|| number|dwnlded|
```

5. Read the console output file using the command:

ls -ltr

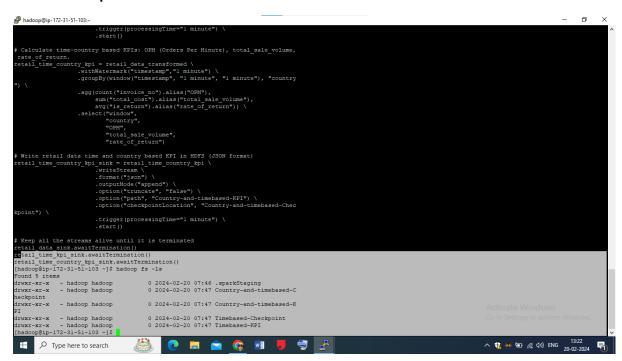
cat Console-output.txt

check whether we can see the transformed data as per our requirement.



6. Now check whether all the JSON files are created for Time based and Time-andCountry based KPIs in HDFS (Path: /user/hadoop/).

hadoop fs -ls



Read Timebased-KPI JSON files:

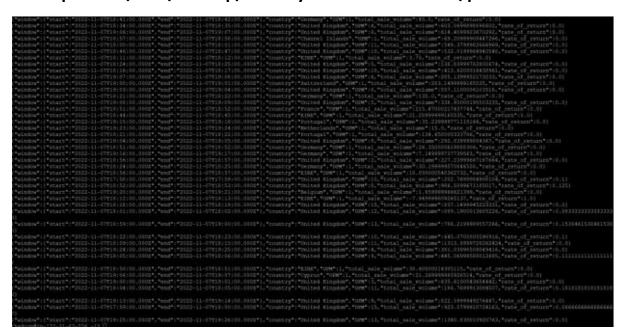
hadoop fs -ls /user/hadoop/Timebased-KPI

hadoop fs -cat /user/hadoop/Timebased-KPI/part*

Read Country-and-timebased-KPI JSON files:

hadoop fs -ls /user/hadoop/Country-and-timebased-KPI

hadoop fs -cat /user/hadoop/Country-and-timebased-KPI/part*



Note: Post completion of all the above steps, terminate the EMR instance from AWS console.