



Code Logic - Retail Data Analysis

Below is the step by step explanation of the code logic and flow:

1. Imported all the required libraries for this project.

```
# Importing required libraries

from pyspark.sql import SparkSession
from pyspark.sql.functions import *
from pyspark.sql.types import *
```

- 2. Defined separate utility functions to determine total items in an order, total cost of an order and whether the order is a new buy order or a return order.
 - a. Utility function to determine total items in an order i.e. sum of the quantities of each item in the order.

```
8
    def get_total_item_count(items):
10
11
        This method is used to calculate the toal number of items in an order.
12
        :param items: The list of items in an order.
13
        :return total count: The sum of quantities of each item in the order.
14
15
        total count = 0
       for item in items:
         total count = total count + int(item['quantity'])
18
        return total count
```

b. Utility function to determine total cost of an order i.e. sum of quantity multiplied by unit price for each item in the order. If the order is a new buy order, cost is returned else if the order is a return order, then cost is multiplied by -1 and then returned.





c. Utility function to determine the value if 'is_order' column. If the order is a new buy order, the value is set to 1 else 0.

```
def get_is_order(order):
36
37
         This method is to determine whether a the order is a new buy order or not.
         :param order: Detrmines the type of order on which the value of 'is_order' column is set.
38
39
         :return 0/1: Return 1 if order is a new buy order else return 0.
40
41 =
42 -
        if order == 'ORDER':
        return 1
43 E
        else:
44
        return 0
45
```

d. Utility function to determine the value if 'is_return' column. If the order is a return order, the value is set to 1 else 0.

3. Establishing the spark session and setting the spark log level to 'ERROR'.

```
# Setting up the spark session
58
59

☐spark = SparkSession \

              .builder \
60
61
              .appName('RetailDataAnalysis') \
62
              .getOrCreate()
63
64
      # Setting the spark log level to ERROR.
65
66
     spark.sparkContext.setLogLevel('ERROR')
```





4. Connect to the given kafka server and topic using the spark session created above from where the data is to be read.

5. Defining the JSON schema for the reading the data from kafka topic

```
# Defining the json Schema of the data to read it properly.
77
78
79
    .add("invoice no", StringType()) \
80
81
                  .add("country", StringType()) \
82
                  .add("timestamp", TimestampType()) \
                  .add("type", StringType()) \
83
84
                  .add("items", ArrayType(StructType([
                      StructField("SKU", StringType()),
85
86
                      StructField("title", StringType()),
87
                      StructField("unit price", StringType()),
                      StructField("quantity", StringType())])))
88
```

6. Casting the raw data as a string and aliasing it to data. In addition to this, defining the UDF's (user defined functions) to calculate additional columns 'total_items', 'total_cost', 'is_order' and 'is_return' using the utility functions defined in the starting.

```
# Coverting raw data to a string and aliasing it as data.
92
      orderStream = orderRaw.select(from json(col('value').cast('string'), jsonSchema).alias('data')).select('data.*')
93
94
      # Defining a UDF's with utility functions for calculating different column values.
95
96
      # UDF to get to total items in an order.
      add_total_item_count = udf(get_total_item_count, IntegerType())
97
98
99
      # UDF to get total cost of an order.
      add_total_order_cost = udf(get_total_order_cost, FloatType())
      # UDF to determine if an order is an order is a new buy order.
      add_is_order = udf(get_is_order, IntegerType())
104
      # UDF to determine if an order is a return order.
105
      add_is_return = udf(get_is_return, IntegerType())
```





7. Calculating the additional columns using UDF's defined above.

```
# Calculating additional columns using the UDF's defined above.

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| Calculating additional columns using the
```

8. Writing the information collected for every 1 minutes time interval to the console.

```
# Writing the summarized information to console at a processing interval of 1 minute.

| PextendedOrderQuery = expandedOrderStream \
| select("invoice_no", "country", "timestamp", "total_cost", "total_items", "is_order", "is_return") \
| writeStream \
| coutputMode("append") \
| format("console") \
| coption("truncate", "false") \
| trigger(processingTime="l minute") \
| start()
```

- 9. The next step is to calculate the time based KPI's (Key Performance Indicators) in a window of 1 minute. To calculate the below KPI's I'm grouping the data based on 1 minute time interval on timestamp column. Also, watermark is set to 1 minute so that the data coming late up to 1 minute is also processed.
 - a. Orders Per Minute (OPM): It is calculated as the count of total invoice numbers in that batch.
 - b. Total Sale Volume: It is defined as the sum of 'total cost' column for that batch.
 - c. Rate of Return: It is defined as the average of 'is return' column for that batch.
 - d. Average Transaction Size: it is calculated as the average 'total_cost' column for that batch.

```
127
       # Calculating Time Based Key Performance Indicators
128
129

    aggStreamByTime = expandedOrderStream \
                          .withWatermark("timestamp", "l minute") \
130
                          .groupBy(window("timestamp", "1 minute", "1 minute")) \
131
132
                          .agg(sum("total cost").alias("total sale volume"),
                                  count ("invoice no") .alias ("OPM") ,
133
134
                                  avg("is_return").alias('rate_of_return'),
135
                                 avg("total cost").alias('average transaction size'))
136
```





10. Following this, the next step is to write the time based KPI's output to a JSON file on HDFS.

```
137
    # Writing Time Based KPI values to a json file.
138
139 =queryByTime = aggStreamByTime \
                    .select("window", "OPM", 'total sale volume', 'rate of return', 'average transaction size')
140
141
                    .writeStream \
                    .outputMode('append') \
142
143
                    .format('json') \
144
                    .option("truncate", "false") \
                    .option('path', '/output/timeBasedKPI') \
145
146
                    .option("checkpointLocation", "/output/timeBasedKPI") \
                    .trigger(processingTime="1 minute") \
147
148
                    .start()
```

11. The next step is to calculate KPI's based on time and country. To calculate these, we're grouping the data based on 1 minute interval on timestamp column and then the country column. Again, watermark is set to 1minute so that the data coming late up to 1 minute is also processed.

12. Finally, we need to write the time and country based API's to JSON files as well and then await for the termination of above three queries.

```
161

queryByTimeCountry = aggStreamByTimeCountry \
162
                            .select("window", 'country', "OPM", 'total sale volume', 'rate of return') \
163
                            .writeStream \
164
                           .outputMode('append') \
165
                            .format('json') \
166
                            .option('truncate', 'false') \
                            .option('path', '/output/timeCountryBasedKPI') \
167
168
                            .option("checkpointLocation", "/output/timeCountryBasedKPI") \
                            .trigger(processingTime = 'l minute') \
169
170
                           .start()
171
172
      # Awaiting termination of the above written queries.
173
174
      extendedOrderQuery.awaitTermination()
175
      queryByTime.awaitTermination()
176
      quervBvTimeCountry.awaittermination()
177
```

13. The next step is to run the code. The below command is to be executed on spark EMR cluster.

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



