

Code Logic - Retail Data Analysis

In this document, you will describe the code and the overall steps taken to solve the project.

logic for python script file named "spark-streaming.py"

Setting up the important libraries for the module

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

Writing the python function which contain logic for the UDFs

UDF1 to determine Is_Order in case of Order.

```
def udf1(a):
    if a == 'ORDER':
        return 1
    else:
        return 0
```

UDF2 to determine Is_Return in case of Return.

```
def udf2(a):
    if a == 'RETURN':
        return 1
    else:
        return 0
```

UDF3 to determine Total Order Cost.

```
def udf3(a, b, c):
    if a == 'ORDER':
        return b * c
    else:
        return b * c * -1
```

Initialising the spark session and setting up the log level to error as a good practice

Establishing Spark Session

```
spark = SparkSession.builder.appName('StructuredSocketRead')
    .getOrCreate()
spark.sparkContext.setLogLevel('ERROR')
```

Reading the input data from kafka mentioning the details of kafka broker such as bootstrap server, port and topic name

```
# Reading data from Kafka Server & Topic given
```

```
lines = spark.readStream.format('kafka'
                                ).option('kafka.bootstrap.servers',
                                '18.211.252.152:9092').option('subscribe', 'real-time-project'
                                ).option('failOnDataLoss', 'false').option('startingOffsets',
                                'earliest').load()
```

Defining schema for each order , using appropriate datatypes and StructField in case of item attribute

```
schema = StructType([StructField('country', StringType()),
                        StructField('invoice_no', LongType()),
                        StructField('items',
                        ArrayType(StructType([StructField('SKU',
                        StringType()), StructField('title', StringType()),
                        StructField('unit_price', FloatType()),
                        StructField('quantity', IntegerType())])),
                        StructField('timestamp', TimestampType()),
                        StructField('type', StringType())])
```

Reading the raw json data from kafka as 'casted' by casting it into string and storing it into alias data

```
# Casting raw data as string and aliasing
```

```
Casted = lines.select(from_json(col('value').cast('string'),
                                schema).alias('parsed'))
```

```
# Parsed DF
```

```
new_df = Casted.select('parsed.*')
```

```
# unveiling items array to derive new columns & KPIs
```

```
df1 = new_df.select(col('type'), col('country'), col('invoice_no'),
                    col('timestamp'), explode(col('items')))
```

Selecting appropriate columns and renaming few

```
# Columns from Items array displayed
```

```
df2 = df1.select(
    'type',
    'country',
    'invoice_no',
    'timestamp',
    'col.SKU',
    'col.title',
```

```
        'col.unit_price',
        'col.quantity',
    )
```

```
# Array Columns renamed
```

```
df2 = df2.withColumnRenamed('col.SKU', 'SKU')
df2 = df2.withColumnRenamed('col.title', 'TITLE')
df2 = df2.withColumnRenamed('col.unit_price', 'unit_price')
df2 = df2.withColumnRenamed('col.quantity', 'quantity')
```

Declaring the udfs

```
# Declare user def function for total cost
```

```
Cost_value = udf(udf3, FloatType())
# Declare user def function Is Order
```

```
Det_Order = udf(udf1, IntegerType())
```

```
# Declare user def function for Is Return
```

```
Det_Return = udf(udf2, IntegerType())
```

Adding the respective variable to the dataframe

```
# Adding Total Cost to data frame
```

```
df2 = df2.withColumn('Cost', Cost_value(df2.type, df2.unit_price,
                                         df2.quantity))
```

```
# Adding Is Order flag to existing DF
```

```
df2 = df2.withColumn('is_Order', Det_Order(df2.type))
```

```
# Adding Is Return flag to existing DF
```

```
df2 = df2.withColumn('is_Return', Det_Return(df2.type))
```

Pumping the raw data to the output for each order for window function for 1 minute

```
df3 = df2.withWatermark('timestamp', '10 minutes')
        .groupby(window('timestamp', '1 minute'),
                  'invoice_no', 'country', 'is_Order',
                  'is_Return').sum('Cost', 'quantity')
    )
```

Function for timebased KPIs

```
df4 = df2.select(
    'invoice_no',
    'timestamp',
```

```

        'Cost',
        'quantity',
        'is_Order',
        'is_Return',
    )
Final_time = df4.withWatermark('timestamp', '10 minutes'
                               ).groupby(window('timestamp', '1
minute'
        )).agg(sum('Cost').alias('Total_sales_vol'),
                F.approx_count_distinct('invoice_no').alias('OPM'),
                sum('is_Order').alias('total_Order'), sum('is_Return'
                ).alias('total_return'), sum('quantity'
                ).alias('total_items'))

```

KPI for rate of return

```
# KPI for rate of return
```

```

Final = Final.withColumn('rate_of_return', Final.total_return
                        / (Final.total_Order + Final.total_return))
Final_country_time = Final.select('window', 'country', 'OPM',
                                  'Total_sales_vol',

```

'rate_of_return') # KPI for average transaction size

```
# KPI for average transaction size....
```

```

Final_time = Final_time.withColumn('Avg_trans_size',
                                   Final_time.Total_sales_vol
                                   / (Final_time.total_Order
                                   + Final_time.total_return))
Final_time = Final_time.select('window', 'OPM', 'Total_sales_vol',
                               'Avg_trans_size', 'rate_of_return')

```

Time and country based KPIs

```

df3_time_country = df2.select(
    'country',
    'invoice_no',
    'timestamp',
    'Cost',
    'quantity',
    'is_Order',
    'is_Return',
)
Final = df3_time_country.withWatermark('timestamp', '10 minutes'
    ).groupby(window('timestamp', '1 minute'), 'country').agg(
    sum('Cost').alias('Total_sales_vol'),
    F.approx_count_distinct('invoice_no').alias('OPM'),
    sum('invoice_no').alias('sum_invoice'),
    sum('is_Order').alias('total_Order'),
    sum('is_Return').alias('total_return'),
    sum('quantity').alias('total_items'),
)

```

Printing the KPIs to HDFS as json file

```
query_2 = Final_time.writeStream.outputMode('Append').format('json')
    .option('format', 'append').option('truncate', 'false')
    .option('path', 'time_KPI').option('checkpointLocation',
    'time_KPI_json').trigger(processingTime='1 minute').start()
```

Printing the output on console

printing output on console

```
query_1 = df3.writeStream.outputMode('complete').format('console')
    .option('truncate', 'False').start()
```

Printing Time and Country KPI to HDFS as Json file

Printing time and country KPIs to HDFS as json file

```
query_3 = Final_country_time.writeStream.outputMode('Append')
    .format('json').option('format', 'append').option('truncate',
    'false').option('path', 'time_country_KPI')
    .option('checkpointLocation',
    'time_country_KPI_json')
    .trigger(processingTime='1 minute')
    .start()
```

Query termination command

```
query_1.awaitTermination()
query_2.awaitTermination()
query_3.awaitTermination()
```

Console Command:

- Setup EMR Cluster for spark streaming and login as hadoop
- go to root user by sudo -i
- install pip install kafka-python
- Next, open another console and login as hadoop and create spark-streaming.py file by "vi spark-streaming.py"
- Next, i set the kafka version using the following command: "export PYSPARK_KAFKA_VERSION=0.10."
- After writing the code in the spark-streaming.py we can run the sparkjon by the following command: spark-submit --packages org.apache.spark:spark-sql-kafka-0-10_2.11:2.4.5 spark-streaming.py.

Final summarised Input Values

Batch: 0

Window	Invoice_No	Country	Is_Order	Is_Return	Sum(Cost)	Sum(Quantity)
[2022-04-11 00:50:00, 2022-04-11 00:51:00]	154132548937038	United Kingdom	1	0	18.149999618530273	15
[2022-04-10 20:52:00, 2022-04-10 20:53:00]	154132548934614	United Kingdom	1	0	14.090000033378601	12
[2022-04-08 20:22:00, 2022-04-08 20:23:00]	154132548905259	United Kingdom	1	0	188.91999816894531	160
[2022-04-10 03:36:00, 2022-04-10 03:37:00]	154132548924190	United Kingdom	1	0	12.339999914169312	14
[2022-04-11 09:43:00, 2022-04-11 09:44:00]	154132548942537	United Kingdom	1	0	137.41999530792236	140
[2022-04-13 01:06:00, 2022-04-13 01:07:00]	154132548966337	United Kingdom	1	0	124.9399995803833	114
[2022-04-09 15:05:00, 2022-04-09 15:06:00]	154132548916479	United Kingdom	1	0	16.28000032901764	18
[2022-04-08 22:57:00, 2022-04-08 22:58:00]	154132548906837	United Kingdom	1	0	128.090000867843628	117
[2022-04-10 19:18:00, 2022-04-10 19:19:00]	154132548933741	United Kingdom	1	0	16.6299999523163	15
[2022-04-10 09:31:00, 2022-04-10 09:32:00]	154132548927860	United Kingdom	1	0	142.049999833106995	117
[2022-04-11 03:43:00, 2022-04-11 03:44:00]	154132548938808	United Kingdom	1	0	102.09000205993652	113
[2022-04-08 22:44:00, 2022-04-08 22:45:00]	154132548906719	United Kingdom	1	0	1515.8799985647202	123
[2022-04-10 12:57:00, 2022-04-10 12:58:00]	154132548929977	United Kingdom	1	0	135.62999987602234	124
[2022-04-10 17:52:00, 2022-04-10 17:53:00]	154132548932853	United Kingdom	1	0	146.88999938964844	116
[2022-04-10 18:09:00, 2022-04-10 18:10:00]	154132548933046	United Kingdom	1	0	1297.310001373291	157
[2022-04-12 20:01:00, 2022-04-12 20:02:00]	154132548963298	EIRE	1	0	128.34999942779541	17
[2022-04-13 00:16:00, 2022-04-13 00:17:00]	154132548965849	United Kingdom	1	0	10.470000088214874	13
[2022-04-10 20:02:00, 2022-04-10 20:03:00]	154132548934159	United Kingdom	1	0	11.930000185966492	16
[2022-04-09 02:13:00, 2022-04-09 02:14:00]	154132548908847	United Kingdom	1	0	161.87999820709229	137
[2022-04-11 05:38:00, 2022-04-11 05:39:00]	154132548939969	United Kingdom	1	0	175.29999732971191	114

only showing top 20 rows

Check in HDFS that all the required KPIs are present

- hadoop fs -ls

```
[hadoop@ip-172-31-87-221 ~]$ hadoop fs -ls
Found 5 items
drwxr-xr-x - hadoop hadoop 0 2022-04-13 09:55 .sparkStaging
drwxr-xr-x - hadoop hadoop 0 2022-04-13 09:55 time_KPI
drwxr-xr-x - hadoop hadoop 0 2022-04-13 09:30 time_KPI_json
drwxr-xr-x - hadoop hadoop 0 2022-04-13 09:55 time_country_KPI
drwxr-xr-x - hadoop hadoop 0 2022-04-13 09:30 time_country_KPI_json
```

Also checked folders to see json files

-hadoop fs -ls time_KPI

```
[hadoop@ip-172-31-87-221 ~]$ hadoop fs -ls time_KPI
Found 231 items
drwxr-xr-x - hadoop hadoop 0 2022-04-13 09:55 time_KPI/_spark_metadata
-rw-r--r-- 1 hadoop hadoop 7144 2022-04-13 09:31 time_KPI/part-00000-0dfcd44d-ef6b-4c1d-8203-3c1e23ac6d3a-c000.json
-rw-r--r-- 1 hadoop hadoop 0 2022-04-13 09:30 time_KPI/part-00000-12521846-2577-43f6-b382-be72eb26593f-c000.json
-rw-r--r-- 1 hadoop hadoop 0 2022-04-13 09:40 time_KPI/part-00000-1d58138c-f1b8-4262-99bc-df29bb107c4c-c000.json
-rw-r--r-- 1 hadoop hadoop 0 2022-04-13 09:55 time_KPI/part-00000-45ee51e1-201c-4fda-996a-a723fcaaeac7-c000.json
-rw-r--r-- 1 hadoop hadoop 0 2022-04-13 09:54 time_KPI/part-00000-6da32e0a-4c5f-4d20-8768-aa2b7de3b6af-c000.json
-rw-r--r-- 1 hadoop hadoop 0 2022-04-13 09:32 time_KPI/part-00000-acalbf7-d3dc-4003-bd38-586fd9767066-c000.json
-rw-r--r-- 1 hadoop hadoop 0 2022-04-13 09:41 time_KPI/part-00000-c9163fc0-3f05-4c9b-8670-a2dce8e2b203-c000.json
-rw-r--r-- 1 hadoop hadoop 8385 2022-04-13 09:31 time_KPI/part-00001-a507e139-4610-43f1-a281-1e82ebc5b634-c000.json
-rw-r--r-- 1 hadoop hadoop 7747 2022-04-13 09:31 time_KPI/part-00002-48a9e60d-be0c-427f-956b-45d39c245890-c000.json
-rw-r--r-- 1 hadoop hadoop 8632 2022-04-13 09:31 time_KPI/part-00003-a5b101e3-4321-45eb-8aed-0837a7e09fb9-c000.json
-rw-r--r-- 1 hadoop hadoop 184 2022-04-13 09:55 time_KPI/part-00004-993acb72-1d27-4f11-a808-9c9ece6a6af8-c000.json
-rw-r--r-- 1 hadoop hadoop 6160 2022-04-13 09:31 time_KPI/part-00004-ca49a5b8-76b8-48d4-baa3-fe4e1786e1b4-c000.json
-rw-r--r-- 1 hadoop hadoop 7599 2022-04-13 09:31 time_KPI/part-00005-7ef16c7a-7bee-4036-9507-1d84b2ffac69-c000.json
-rw-r--r-- 1 hadoop hadoop 7941 2022-04-13 09:31 time_KPI/part-00006-05748603-0670-4c79-8e6a-e0263b1bda7f-c000.json
-rw-r--r-- 1 hadoop hadoop 9091 2022-04-13 09:31 time_KPI/part-00007-0ca5d872-41ec-45e9-9d62-89e103f23c99-c000.json
-rw-r--r-- 1 hadoop hadoop 8512 2022-04-13 09:31 time_KPI/part-00008-8bc36165-a6d4-44c7-8b71-025fed0cf3d6-c000.json
-rw-r--r-- 1 hadoop hadoop 7978 2022-04-13 09:31 time_KPI/part-00009-cdef628-1342-4f5b-b6f9-fb839c8207d5-c000.json
-rw-r--r-- 1 hadoop hadoop 8474 2022-04-13 09:31 time_KPI/part-00010-71b5a582-e8ac-48fe-9ad6-dfb3c88adb9b-c000.json
```


-hadoop fs -ls time_country_KPI

```
[hadoop@ip-172-31-87-221 ~]$ hadoop fs -ls time_country_KPI
Found 248 items
drwxr-xr-x  - hadoop hadoop          0 2022-04-13 09:55 time_country_KPI/ spark_metadata
-rw-r--r--  1 hadoop hadoop          0 2022-04-13 09:40 time_country_KPI/part-00000-110748a9-7809-4349-bac2-00461468b406-c000.json
-rw-r--r--  1 hadoop hadoop          0 2022-04-13 09:32 time_country_KPI/part-00000-294a7043-f98d-4967-9a24-334dc898780c-c000.json
-rw-r--r--  1 hadoop hadoop          0 2022-04-13 09:54 time_country_KPI/part-00000-5482c10e-55b6-4e8d-b1a4-aa9fe3d611e0-c000.json
-rw-r--r--  1 hadoop hadoop    14936 2022-04-13 09:31 time_country_KPI/part-00000-88644f7a-f8db-4199-9246-210fbf5dd8b9-c000.json
-rw-r--r--  1 hadoop hadoop          0 2022-04-13 09:41 time_country_KPI/part-00000-8872f346-fdc9-435e-bb47-9e6bf9f7fc73-c000.json
-rw-r--r--  1 hadoop hadoop          0 2022-04-13 09:30 time_country_KPI/part-00000-a709906f-bd1a-48c1-ae0f-3f98789f6486-c000.json
-rw-r--r--  1 hadoop hadoop          0 2022-04-13 09:55 time_country_KPI/part-00000-dd86dec3-f82e-4131-8673-6e9e4a023973-c000.json
-rw-r--r--  1 hadoop hadoop    12172 2022-04-13 09:31 time_country_KPI/part-00001-a3e07a13-ba30-4611-87a9-893b0e64dfd9-c000.json
-rw-r--r--  1 hadoop hadoop    12885 2022-04-13 09:31 time_country_KPI/part-00002-7a623264-2e3c-4d59-b648-3c0bc261d2b0-c000.json
-rw-r--r--  1 hadoop hadoop    15101 2022-04-13 09:31 time_country_KPI/part-00003-df1bc123-0aa3-499f-a787-faab7db64049-c000.json
-rw-r--r--  1 hadoop hadoop    13861 2022-04-13 09:31 time_country_KPI/part-00004-d8b07ec6-3039-4ffd-bc70-c4496a2f5b6e-c000.json
-rw-r--r--  1 hadoop hadoop    12531 2022-04-13 09:31 time_country_KPI/part-00005-16a9c314-075b-4635-9818-8aa9f48029d1-c000.json
-rw-r--r--  1 hadoop hadoop      173 2022-04-13 09:41 time_country_KPI/part-00005-6f06b466-4929-4e24-8f63-428e0d40737e-c000.json
-rw-r--r--  1 hadoop hadoop    14908 2022-04-13 09:31 time_country_KPI/part-00006-c5e5b5ad-6441-4090-ad6a-064f16050985-c000.json
-rw-r--r--  1 hadoop hadoop    10731 2022-04-13 09:31 time_country_KPI/part-00007-51f2899a-99d2-4541-8230-7e1e2d8db8cb-c000.json
-rw-r--r--  1 hadoop hadoop      165 2022-04-13 09:55 time_country_KPI/part-00007-ed4ff1ca-b02b-4bf1-bc77-0a38351ba68d-c000.json
-rw-r--r--  1 hadoop hadoop    14333 2022-04-13 09:31 time_country_KPI/part-00008-4be5bd10-2ca3-4094-b5ff-fa3e17810abf-c000.json
-rw-r--r--  1 hadoop hadoop    11497 2022-04-13 09:31 time_country_KPI/part-00009-64274eb8-c25b-4894-b5e9-8ba6bc7454d4-c000.json
-rw-r--r--  1 hadoop hadoop    13684 2022-04-13 09:31 time_country_KPI/part-00010-d37c76f9-7fb7-48c5-9025-410c5a12e42f-c000.json
```

Used cat command to take a look at the data:

-hadoop fs -cat time_KPI/part*

```
[hadoop@ip-172-31-87-221 ~]$ hadoop fs -cat time_KPI/part*
{"window":{"start":"2022-04-13T08:19:00.000Z","end":"2022-04-13T08:20:00.000Z"},"OPM":8,"Total_sales_vol":442.4299959242344,"Avg_trans_size":15.80107128300837,"rate_of_return":0.0}
{"window":{"start":"2022-04-08T22:05:00.000Z","end":"2022-04-08T22:06:00.000Z"},"OPM":6,"Total_sales_vol":393.3500111103058,"Avg_trans_size":49.16875138878822,"rate_of_return":0.0}
{"window":{"start":"2022-04-10T23:02:00.000Z","end":"2022-04-10T23:03:00.000Z"},"OPM":12,"Total_sales_vol":2034.7099537849426,"Avg_trans_size":58.13457010814122,"rate_of_return":0.0}
{"window":{"start":"2022-04-11T00:29:00.000Z","end":"2022-04-11T00:30:00.000Z"},"OPM":10,"Total_sales_vol":365.55999717116356,"Avg_trans_size":11.423749911598861,"rate_of_return":0.0}
{"window":{"start":"2022-04-09T01:43:00.000Z","end":"2022-04-09T01:44:00.000Z"},"OPM":9,"Total_sales_vol":941.9599961340427,"Avg_trans_size":24.152820413693405,"rate_of_return":0.0}
{"window":{"start":"2022-04-08T10:12:00.000Z","end":"2022-04-08T10:13:00.000Z"},"OPM":5,"Total_sales_vol":258.51001274585724,"Avg_trans_size":19.88538559583517,"rate_of_return":0.3076923076923077}
{"window":{"start":"2022-04-12T10:29:00.000Z","end":"2022-04-12T10:30:00.000Z"},"OPM":10,"Total_sales_vol":565.409888503551,"Avg_trans_size":16.154571110010146,"rate_of_return":0.0}
{"window":{"start":"2022-04-11T17:17:00.000Z","end":"2022-04-11T17:18:00.000Z"},"OPM":7,"Total_sales_vol":366.25000166893005,"Avg_trans_size":12.208333388964336,"rate_of_return":0.0}
{"window":{"start":"2022-04-11T14:18:00.000Z","end":"2022-04-11T14:19:00.000Z"},"OPM":6,"Total_sales_vol":372.35000535845757,"Avg_trans_size":20.6861114088032,"rate_of_return":0.0}
{"window":{"start":"2022-04-11T08:16:00.000Z","end":"2022-04-11T08:17:00.000Z"},"OPM":4,"Total_sales_vol":208.3400073349476,"Avg_trans_size":13.021250458434224,"rate_of_return":0.1875}
{"window":{"start":"2022-04-11T00:21:00.000Z","end":"2022-04-11T00:22:00.000Z"},"OPM":9,"Total_sales_vol":1.5199961066246033,"Avg_trans_size":0.042222114072905645,"rate_of_return":0.08333333333333333}
{"window":{"start":"2022-04-11T14:07:00.000Z","end":"2022-04-11T14:08:00.000Z"},"OPM":15,"Total_sales_vol":584.2900012135506,"Avg_trans_size":15.791621654420286,"rate_of_return":0.0}
{"window":{"start":"2022-04-12T04:03:00.000Z","end":"2022-04-12T04:04:00.000Z"},"OPM":3,"Total_sales_vol":27.15999972820282,"Avg_trans_size":5.431999945640564,"rate_of_return":0.0}
{"window":{"start":"2022-04-13T01:10:00.000Z","end":"2022-04-13T01:11:00.000Z"},"OPM":5,"Total_sales_vol":270.8599934875965,"Avg_trans_size":13.542999674379825,"rate_of_return":0.0}
```

-hadoop fs -cat time_country_KPI/part*

```
[hadoop@ip-172-31-87-221 ~]$ hadoop fs -cat time_country_KPI/part*
{"window":{"start":"2022-04-12T20:38:00.000Z","end":"2022-04-12T20:39:00.000Z"},"country":"Italy","OPM":1,"Total_sales_vol":49.86000299453735,"rate_of_return":0.0}
{"window":{"start":"2022-04-10T13:13:00.000Z","end":"2022-04-10T13:14:00.000Z"},"country":"United Kingdom","OPM":15,"Total_sales_vol":1328.3400115966797,"rate_of_return":0.0}
{"window":{"start":"2022-04-09T21:05:00.000Z","end":"2022-04-09T21:06:00.000Z"},"country":"EIRE","OPM":1,"Total_sales_vol":15.710000038146973,"rate_of_return":0.0}
{"window":{"start":"2022-04-11T12:34:00.000Z","end":"2022-04-11T12:35:00.000Z"},"country":"United Kingdom","OPM":11,"Total_sales_vol":720.9799802899361,"rate_of_return":0.02777777777777776}
{"window":{"start":"2022-04-10T03:43:00.000Z","end":"2022-04-10T03:44:00.000Z"},"country":"United Kingdom","OPM":6,"Total_sales_vol":313.71000149846077,"rate_of_return":0.0}
{"window":{"start":"2022-04-10T16:35:00.000Z","end":"2022-04-10T16:36:00.000Z"},"country":"EIRE","OPM":1,"Total_sales_vol":4345.100000143051,"rate_of_return":0.0}
{"window":{"start":"2022-04-10T04:26:00.000Z","end":"2022-04-10T04:27:00.000Z"},"country":"France","OPM":2,"Total_sales_vol":493.93999660015106,"rate_of_return":0.0}
{"window":{"start":"2022-04-12T15:59:00.000Z","end":"2022-04-12T16:00:00.000Z"},"country":"United Kingdom","OPM":11,"Total_sales_vol":359.8899979889393,"rate_of_return":0.13793103448275862}
{"window":{"start":"2022-04-08T14:03:00.000Z","end":"2022-04-08T14:04:00.000Z"},"country":"United Kingdom","OPM":13,"Total_sales_vol":549.6299954652786,"rate_of_return":0.0975609756097561}
{"window":{"start":"2022-04-11T04:05:00.000Z","end":"2022-04-11T04:06:00.000Z"},"country":"United Kingdom","OPM":12,"Total_sales_vol":436.4900006055832,"rate_of_return":0.10344827586206896}
{"window":{"start":"2022-04-08T17:51:00.000Z","end":"2022-04-08T17:52:00.000Z"},"country":"France","OPM":1,"Total_sales_vol":47.03999996185303,"rate_of_return":0.0}
{"window":{"start":"2022-04-08T00:27:00.000Z","end":"2022-04-08T00:28:00.000Z"},"country":"Sweden","OPM":1,"Total_sales_vol":5.53000009059906,"rate_of_return":0.0}
{"window":{"start":"2022-04-10T12:41:00.000Z","end":"2022-04-10T12:42:00.000Z"},"country":"Germany","OPM":1,"Total_sales_vol":60.77999997138977,"rate_of_return":0.0}
{"window":{"start":"2022-04-13T01:17:00.000Z","end":"2022-04-13T01:18:00.000Z"},"country":"United Kingdom","OPM":6,"Total_sales_vol":402.13999915122986,"rate_of_return":0.0}
{"window":{"start":"2022-04-09T06:11:00.000Z","end":"2022-04-09T06:12:00.000Z"},"country":"United Kingdom","OPM":5,"Total_sales_vol":300.8199962377548,"rate_of_return":0.0}
{"window":{"start":"2022-04-12T19:03:00.000Z","end":"2022-04-12T19:04:00.000Z"},"country":"EIRE","OPM":1,"Total_sales_vol":10.819999933242798,"rate_of_return":0.0}
{"window":{"start":"2022-04-13T06:20:00.000Z","end":"2022-04-13T06:21:00.000Z"},"country":"United Kingdom","OPM":9,"Total_sales_vol":353.1799958348274,"rate_of_return":0.0}
{"window":{"start":"2022-04-08T00:04:00.000Z","end":"2022-04-08T00:05:00.000Z"},"country":"United Kingdom","OPM":2,"Total_sales_vol":209.52999806404114,"rate_of_return":0.0}
{"window":{"start":"2022-04-12T00:25:00.000Z","end":"2022-04-12T00:26:00.000Z"},"country":"United Kingdom","OPM":5,"Total_sales_vol":136.3699984550476,"rate_of_return":0.15384615384615385}
```

Transfer of file from HDFS to local machine using WINSXP

-i created the directories timebased-KPI and country-and-timebased-KPI . using the get command i copied the output the the following directories

```
[hadoop@ip-172-31-87-221 ~]$ mkdir timebased-KPI
[hadoop@ip-172-31-87-221 ~]$ hadoop fs -get time_KPI timebased_KPI
[hadoop@ip-172-31-87-221 ~]$ mkdir country-and-timebased-KPI
[hadoop@ip-172-31-87-221 ~]$ hadoop fs -get time_country_KPI country-and-timebased_KPI
[hadoop@ip-172-31-87-221 ~]$
```

Thereafter i used winscp to establish connection between the cluster and my local machine and copied them into the local machine.