

Loading dataset into FTP location and then to HDFS Directory:

```
[therohitsaha08gmail@sl-cdp-prod-en10 ~]$ cd Capstone/ECommerce
[therohitsaha08gmail@sl-cdp-prod-en10 ECommerce]$ ls
olist_public_dataset.csv
[therohitsaha08gmail@sl-cdp-prod-en10 ECommerce]$ hdfs dfs -ls
Found 6 items
drwxrwxr-x - therohitsaha08gmail hadoop 0 2024-01-29 11:42 .sparkStaging
drwxrwxr-x - therohitsaha08gmail hadoop 0 2024-01-25 08:41 ECommerce
drwxrwxr-x - therohitsaha08gmail hadoop 0 2024-01-24 11:31 Weather_Analysis_Project
drwxrwxr-x - therohitsaha08gmail hadoop 0 2024-01-18 12:12 YoutubeDataProject
drwxrwxrwx - therohitsaha08gmail hadoop 0 2023-08-02 09:59 hive_demo
drwxrwxr-x - therohitsaha08gmail hadoop 0 2023-04-22 13:18 simplilearn
[therohitsaha08gmail@sl-cdp-prod-en10 ECommerce]$ hdfs dfs -ls ECommerce
Found 1 items
-rw-rw-r-- 2 therohitsaha08gmail hadoop 10304725 2024-01-25 08:41 ECommerce/olist_public_dataset.csv
[therohitsaha08gmail@sl-cdp-prod-en10 ECommerce]$
```

Load dataset into Spark DataFrame from HDFS:

```
Welcome to
      ____
     / ___/
    / __/
   /___/
  version 3.2.1.7.2.15.1-1

Using Scala version 2.12.10 (OpenJDK 64-Bit Server VM, Java 1.8.0_332)
Type in expressions to have them evaluated.
Type :help for more information.

scala> var df = spark.read.format("csv").option("delimiter", ";").option("header", "false").option("inferSchema", "true").load("ECommerce/olist_public_dataset.csv").toDF("id", "order_status", "order_products_value", "order_freight_value", "order_items_qty", "customer_city", "customer_state", "customer_zip_code_prefix", "product_name_length", "product_description_length", "product_photos_qty", "review_score", "order_purchase_timestamp", "order_approved_at", "order_delivered_customer_date")
df: org.apache.spark.sql.DataFrame = [id: int, order_status: string ... 13 more fields]

scala> df.show()
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|id|order_status|order_products_value|order_freight_value|order_items_qty|customer_city|customer_state|customer_zip_code_prefix|product_name_length|product_description_length|product_photos_qty|review_score|order_purchase_timestamp|order_approved_at|order_delivered_customer_date|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|1|delivered|79.0|17.8|1|Luziania|GO|728|50|201|2|5|02/10/17 10:56|02/10/17 11:07|10/10/17 21:25|
|2|delivered|119.9|27.16|1|Joinville|SC|892|50|511|3|5|24/07/18 20:41|26/07/18 3:24|07/08/18 15:27|
|3|delivered|519.99|41.69|1|Serra|ES|291|48|1156|2|1|08/08/18 8:38|08/08/18 8:55|17/08/18 18:06|
|4|delivered|29.5|17.92|1|RIO DE JANEIRO|RJ|222|21|207|2|4|18/11/17 19:28|18/11/17 19:45|02/12/17 0:28|
|5|delivered|26.77|23.11|1|Sao Paulo|SP|40|41|451|1|5|13/02/18 21:18|13/02/18 22:20|16/02/18 18:17|
|6|delivered|419.9|23.02|1|Santa Adelia|SP|159|46|692|3|5|09/07/17 21:57|09/07/17 22:10|26/07/17 10:57|
|7|delivered|65.0|16.21|1|Varginha|MG|370|47|893|3|5|11/04/17 12:22|13/04/17 13:25|13/04/17 13:25|
|8|delivered|29.99|19.82|1|Sao Paulo|SP|48|55|613|1|4|16/05/17 13:10|16/05/17 13:22|26/05/17 12:55|
|9|delivered|59.99|51.14|1|Carajas|PA|685|28|96|4|5|23/01/17 11:55|29/07/17 12:05|16/08/17 17:14|
|10|delivered|56.99|16.13|1|Resende|RJ|275|59|641|3|5|29/07/17 11:55|29/07/17 12:05|16/08/17 17:14|
|11|delivered|599.0|15.69|1|Guaratininga|SP|125|43|1686|2|4|16/05/17 11:55|29/07/17 12:05|16/08/17 17:14|
```

Extract only 'date' from all 3 timestamp columns respectively and store in a new dataframe:

```
scala> var date_df = df.withColumn("PurchaseDate",to_date(unix_timestamp(col("order_purchase_timestamp"), "dd/MM/yy H:mm").cast("timestamp"))).withColumn("ApproveDate",to_date(unix_timestamp(col("order_approved_at"), "dd/MM/yy H:mm").cast("timestamp"))).withColumn("DeliveryDate",to_date(unix_timestamp(col("order_delivered_customer_date"), "dd/MM/yy H:mm").cast("timestamp")))
date_df: org.apache.spark.sql.DataFrame = [id: int, order_status: string ... 16 more fields]

scala> date_df.show()
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|id|order_status|order_products_value|order_freight_value|order_items_qty|customer_city|customer_state|customer_zip_code_prefix|product_name_length|product_description_length|product_photos_qty|review_score|order_purchase_timestamp|order_approved_at|order_delivered_customer_date|PurchaseDate|ApproveDate|DeliveryDate|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|1|delivered|79.0|17.8|1|Luziania|GO|728|50|201|2|5|02/10/17 10:56|02/10/17 11:07|10/10/17 21:25|2017-10-02|2017-10-02|2017-10-10|
|2|delivered|119.9|27.16|1|Joinville|SC|892|50|511|3|5|24/07/18 20:41|26/07/18 3:24|07/08/18 15:27|2018-07-24|2018-07-26|2018-08-07|
|3|delivered|519.99|41.69|1|Serra|ES|291|48|1156|2|1|08/08/18 8:38|08/08/18 8:55|17/08/18 18:06|2018-08-08|2018-08-08|2018-08-17|
|4|delivered|29.5|17.92|1|RIO DE JANEIRO|RJ|222|21|207|2|4|18/11/17 19:28|18/11/17 19:45|02/12/17 0:28|2017-11-18|2017-11-18|2017-12-02|
|5|delivered|26.77|23.11|1|Sao Paulo|SP|40|41|451|1|5|13/02/18 21:18|13/02/18 22:20|16/02/18 18:17|2018-02-13|2018-02-13|2018-02-15|
|6|delivered|419.9|23.02|1|Santa Adelia|SP|159|46|692|3|5|09/07/17 21:57|09/07/17 22:10|26/07/17 10:57|2017-07-09|2017-07-09|2017-07-26|
|7|delivered|65.0|16.21|1|Varginha|MG|370|47|893|3|5|11/04/17 12:22|13/04/17 13:25|13/04/17 13:25|2017-04-11|2017-04-13|2017-04-13|
|8|delivered|29.99|19.82|1|Sao Paulo|SP|48|55|613|1|4|16/05/17 13:10|16/05/17 13:22|26/05/17 12:55|2017-05-16|2017-05-16|2017-05-26|
|9|delivered|59.99|51.14|1|Carajas|PA|685|28|96|4|5|23/01/17 11:55|29/07/17 12:05|16/08/17 17:14|2017-01-23|2017-01-25|2017-02-02|
|10|delivered|56.99|16.13|1|Resende|RJ|275|59|641|3|5|29/07/17 11:55|29/07/17 12:05|16/08/17 17:14|2017-07-29|2017-07-29|2017-08-16|
|11|delivered|599.0|15.69|1|Guaratininga|SP|125|43|1686|2|4|16/05/17 11:55|29/07/17 12:05|16/08/17 17:14|2017-07-29|2017-07-29|2017-08-16|
```

1. DAILY INSIGHTS

a. Sales

Total sales :

```
scala> var sales_df = date_df.groupBy(col("PurchaseDate")).agg(round(sum(col("order_products_value")),2).as("Total_Sales")).orderBy(asc("PurchaseDate"))
sales_df: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [PurchaseDate: date, Total_Sales: double]

scala> sales_df.show()
+-----+-----+
|PurchaseDate|Total_Sales|
+-----+-----+
| 2016-09-04|      200.0|
| 2016-09-05|       59.0|
| 2016-09-13|       99.0|
| 2016-09-15|       90.0|
| 2016-10-02|      56.99|
| 2016-10-03|      962.3|
| 2016-10-04|     8827.6|
| 2016-10-05|    6264.19|
| 2016-10-06|    8214.07|
| 2016-10-07|    6282.83|
| 2016-10-08|    4575.48|
| 2016-10-09|    2736.38|
| 2016-10-10|    4130.77|
| 2016-10-22|       49.0|
| 2016-12-23|       82.5|
| 2017-01-05|    3891.87|
| 2017-01-06|    391.98|
| 2017-01-07|    357.36|
| 2017-01-08|    552.58|
| 2017-01-09|    559.88|
+-----+-----+
only showing top 20 rows
```

Total sales in each city :

```
scala> var sales_city_df = date_df.groupBy(col("PurchaseDate"),col("customer_city")).agg(round(sum(col("order_products_value")),2).as("Sales_per_City")).orderBy(asc("PurchaseDate"))
sales_city_df: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [PurchaseDate: date, customer_city: string ... 1 more field]

scala> sales_city_df.show()
+-----+-----+-----+
|PurchaseDate|customer_city|Sales_per_City|
+-----+-----+-----+
| 2016-09-04|Foz do Iguaçu|      200.0|
| 2016-09-05|Betim|       59.0|
| 2016-09-13|Feira de Santana|       99.0|
| 2016-09-15|Fortaleza|       90.0|
| 2016-10-02|Montes Claros|      56.99|
| 2016-10-03|Eunapolis|     139.9|
| 2016-10-03|Jundiaí|       45.8|
| 2016-10-03|Belo Horizonte|      81.9|
| 2016-10-03|Taubaté|       58.9|
| 2016-10-03|Bauru|     455.0|
| 2016-10-03|São Paulo|      42.0|
| 2016-10-03|Estrela do Sul|      89.9|
| 2016-10-03|Guacuí|       48.9|
| 2016-10-04|Recife|     267.4|
| 2016-10-04|Vitória da Conquista|    330.0|
| 2016-10-04|Paulo Afonso|      45.9|
| 2016-10-04|São Borja|      18.9|
| 2016-10-04|Canoas|     46.99|
| 2016-10-04|Uberlândia|      30.9|
| 2016-10-04|Campinas|      34.9|
+-----+-----+-----+
only showing top 20 rows
```

Total sales in each state :

```
scala> var sales_state_df = date_df.groupBy(col("PurchaseDate"),col("customer_state")).agg(round(sum(col("order_products_value")),2).as("Sales_per_State")).orderBy(asc("PurchaseDate"))
sales_state_df: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [PurchaseDate: date, customer_state: string ... 1 more field]

scala> sales_state_df.show()
+-----+-----+-----+
|PurchaseDate|customer_state|Sales_per_State|
+-----+-----+-----+
| 2016-09-04|PR|      200.0|
| 2016-09-05|MG|       59.0|
| 2016-09-13|BA|       99.0|
| 2016-09-15|CE|       90.0|
| 2016-10-02|MG|      56.99|
| 2016-10-03|ES|       48.9|
| 2016-10-03|BA|     139.9|
| 2016-10-03|MG|     171.8|
| 2016-10-03|SP|     601.7|
| 2016-10-04|SP|    3100.61|
| 2016-10-04|PE|    1066.4|
| 2016-10-04|RJ|    1287.52|
| 2016-10-04|MG|    1215.49|
| 2016-10-04|RS|     984.39|
| 2016-10-04|GO|     270.69|
| 2016-10-04|SC|     109.9|
| 2016-10-04|DF|       67.9|
| 2016-10-04|PR|       99.0|
| 2016-10-04|ES|       69.9|
| 2016-10-04|BA|     555.8|
+-----+-----+-----+
only showing top 20 rows
```

b. Orders

Total number of orders :

```
scala> var total_order = date_df.groupBy("PurchaseDate").agg(count("*").as("Total_Orders")).orderBy(asc("PurchaseDate"))
total_order: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [PurchaseDate: date, Total_Orders: bigint]

scala> total_order.show()
+-----+-----+
|PurchaseDate|Total_Orders|
+-----+-----+
| 2016-09-04|          1|
| 2016-09-05|          1|
| 2016-09-13|          1|
| 2016-09-15|          1|
| 2016-10-02|          1|
| 2016-10-03|           8|
| 2016-10-04|         63|
| 2016-10-05|         48|
| 2016-10-06|         51|
| 2016-10-07|         46|
| 2016-10-08|         43|
| 2016-10-09|         26|
| 2016-10-10|         39|
| 2016-10-22|          1|
| 2016-12-23|          1|
| 2017-01-05|         32|
| 2017-01-06|          4|
| 2017-01-07|          4|
| 2017-01-08|          6|
| 2017-01-09|          5|
+-----+-----+
only showing top 20 rows
```

City-wise order distribution :

```
scala> var city_order = date_df.groupBy(col("PurchaseDate"),col("customer_city")).agg(count("*").as("Total_Orders")).orderBy(asc("PurchaseDate"))
city_order: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [PurchaseDate: date, customer_city: string ... 1 more field]

scala> city_order.show()
+-----+-----+-----+
|PurchaseDate|customer_city|Total_Orders|
+-----+-----+-----+
| 2016-09-04|Foz do Iguaçu|          1|
| 2016-09-05|Betim|          1|
| 2016-09-13|Feira de Santana|          1|
| 2016-09-15|Fortaleza|          1|
| 2016-10-02|Montes Claros|          1|
| 2016-10-03|Belo Horizonte|          1|
| 2016-10-03|Guacuí|          1|
| 2016-10-03|Eunapolis|          1|
| 2016-10-03|São Paulo|          1|
| 2016-10-03|Estrela do Sul|          1|
| 2016-10-03|Taubaté|          1|
| 2016-10-03|Baurur|          1|
| 2016-10-03|Jundiá|          1|
| 2016-10-04|Serra|          1|
| 2016-10-04|Garça|          1|
| 2016-10-04|Campinas|          1|
| 2016-10-04|Anápolis|          1|
| 2016-10-04|Buritis|          1|
| 2016-10-04|Santana de Parnaíba|          1|
| 2016-10-04|São Paulo|         11|
+-----+-----+-----+
only showing top 20 rows
```

State-wise order distribution :

```
scala> var state_order = date_df.groupBy(col("PurchaseDate"),col("customer_state")).agg(count("*").as("Total_Orders")).orderBy(asc("PurchaseDate"))
state_order: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [PurchaseDate: date, customer_state: string ... 1 more field]

scala> state_order.show()
+-----+-----+-----+
|PurchaseDate|customer_state|Total_Orders|
+-----+-----+-----+
| 2016-09-04|PR|          1|
| 2016-09-05|MG|          1|
| 2016-09-13|BA|          1|
| 2016-09-15|CE|          1|
| 2016-10-02|MG|          1|
| 2016-10-03|ES|          1|
| 2016-10-03|BA|          1|
| 2016-10-03|MG|          2|
| 2016-10-03|SP|          4|
| 2016-10-04|SP|         24|
| 2016-10-04|PE|          4|
| 2016-10-04|RJ|          9|
| 2016-10-04|MG|         11|
| 2016-10-04|RS|          4|
| 2016-10-04|GO|          4|
| 2016-10-04|SC|          1|
| 2016-10-04|DF|          1|
| 2016-10-04|PR|          1|
| 2016-10-04|ES|          1|
| 2016-10-04|BA|          3|
+-----+-----+-----+
only showing top 20 rows
```

Average review score per order :

```
scala> var avg_rev = date_df.groupBy(col("PurchaseDate")).agg(round(sum(col("review_score"))/count("*"),3).as("Avg_Rev_per_Order")).orderBy(asc("PurchaseDate"))
avg_rev: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [PurchaseDate: date, Avg_Rev_per_Order: double]

scala> avg_rev.show()
+-----+-----+
|PurchaseDate|Avg_Rev_per_Order|
+-----+-----+
| 2016-09-04|          3.0|
| 2016-09-05|          5.0|
| 2016-09-13|          4.0|
| 2016-09-15|          5.0|
| 2016-10-02|          5.0|
| 2016-10-03|         3.625|
| 2016-10-04|         3.857|
| 2016-10-05|         4.229|
| 2016-10-06|          4.0|
| 2016-10-07|         3.739|
| 2016-10-08|         4.233|
| 2016-10-09|         3.923|
| 2016-10-10|         4.308|
| 2016-10-22|          1.0|
| 2016-12-23|          5.0|
| 2017-01-05|         4.125|
| 2017-01-06|          3.25|
| 2017-01-07|          3.0|
| 2017-01-08|         4.333|
| 2017-01-09|          4.8|
+-----+-----+
only showing top 20 rows
```

Average freight charges per order :

```
scala> var avg_fre = date_df.groupBy(col("PurchaseDate")).agg(round(sum(col("order_freight_value"))/count("*"),3).as("Avg_Frei_per_Order")).orderBy(asc("PurchaseDate"))
avg_fre: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [PurchaseDate: date, Avg_Frei_per_Order: double]

scala> avg_fre.show()
+-----+-----+
|PurchaseDate|Avg_Frei_per_Order|
+-----+-----+
| 2016-09-04|         39.88|
| 2016-09-05|         14.16|
| 2016-09-13|         19.93|
| 2016-09-15|         26.89|
| 2016-10-02|         15.15|
| 2016-10-03|        15.705|
| 2016-10-04|        20.357|
| 2016-10-05|        23.849|
| 2016-10-06|        21.093|
| 2016-10-07|        19.256|
| 2016-10-08|        24.621|
| 2016-10-09|        23.502|
| 2016-10-10|        21.252|
| 2016-10-22|        37.97|
| 2016-12-23|        44.16|
| 2017-01-05|        20.032|
| 2017-01-06|        15.883|
| 2017-01-07|        17.373|
| 2017-01-08|        47.932|
| 2017-01-09|        16.624|
+-----+-----+
only showing top 20 rows
```

Average time taken to approve the orders (order approved – order purchased) :

```
scala> import org.apache.spark.sql.types.DoubleType
import org.apache.spark.sql.types.DoubleType

scala> var approval_diff = df.withColumn("order_purchase_timestamp", to_timestamp(col("order_purchase_timestamp"), "dd/MM/yy H:mm")).withColumn("order_approved_at", to_timestamp(col("order_approved_at"), "dd/MM/yy H:mm")).withColumn
("ApprovalDifference(InMinutes)", ((col("order_approved_at").cast("long") - col("order_purchase_timestamp").cast("long")) / 60).cast(DoubleType)).select("id", "order_purchase_timestamp", "order_approved_at", "ApprovalDifference(InMi
nutes)")
approval_diff: org.apache.spark.sql.DataFrame = [id: int, order_purchase_timestamp: timestamp ... 2 more fields]

scala> approval_diff.show()
+-----+-----+-----+-----+
|id|order_purchase_timestamp|order_approved_at|ApprovalDifference(InMinutes)|
+-----+-----+-----+-----+
|1| 2017-10-02 10:56:00|2017-10-02 11:07:00|          11.0|
|2| 2018-07-24 20:41:00|2018-07-26 03:24:00|        1843.0|
|3| 2018-08-08 00:30:00|2018-08-08 08:55:00|          17.0|
|4| 2017-11-18 19:28:00|2017-11-18 19:45:00|           9.0|
|5| 2018-02-13 21:18:00|2018-02-13 22:20:00|          62.0|
|6| 2017-07-09 21:57:00|2017-07-09 22:10:00|          13.0|
|7| 2017-04-11 12:22:00|2017-04-13 13:25:00|        2943.0|
|8| 2017-05-16 13:10:00|2017-05-16 13:22:00|           9.0|
|9| 2017-01-23 18:29:00|2017-01-25 02:50:00|        1941.0|
|10| 2017-07-29 11:55:00|2017-07-29 12:05:00|           9.0|
|11| 2017-05-16 19:41:00|2017-05-16 19:50:00|           9.0|
|12| 2017-07-13 19:58:00|2017-07-13 20:10:00|          12.0|
|13| 2018-06-07 10:06:00|2018-06-09 03:13:00|        2467.0|
|14| 2018-07-25 17:44:00|2018-07-25 17:55:00|          11.0|
|15| 2018-03-01 14:14:00|2018-03-01 15:10:00|          56.0|
|16| 2018-06-07 19:03:00|2018-06-12 23:31:00|       7468.0|
|17| 2018-01-02 19:00:00|2018-01-02 19:09:00|           9.0|
|18| 2017-12-26 23:41:00|2017-12-26 23:50:00|           9.0|
|19| 2017-11-21 00:03:00|2017-11-21 00:14:00|          11.0|
|20| 2017-10-26 15:54:00|2017-10-26 16:08:00|          14.0|
+-----+-----+-----+-----+
only showing top 20 rows

scala> var avg_approve = approval_diff.agg(round(avg("ApprovalDifference(InMinutes)"),3).as("Average Time To Approve(Minutes))).show()
+-----+
|Average Time To Approve(Minutes)|
+-----+
|                               625.149|
+-----+

avg_approve: Unit = ()
```

```

scala> var delivery_diff = df.withColumn("order_purchase_timestamp", to_timestamp(col("order_purchase_timestamp"), "dd/MM/yyyy HH:mm"))
withColumn("order_delivered_customer_date", to_timestamp(col("order_delivered_customer_date"), "dd/MM/yyyy HH:mm"))
withColumn("DeliveryDifference(InHours)", ((col("order_delivered_customer_date").cast("long") - col("order_purchase_timestamp").cast("long")) / 3600).cast(DoubleType)).select("id", "order_purchase_timestamp", "order_delivered_customer_date", "DeliveryDifference(InHours)")
delivery_diff: org.apache.spark.sql.DataFrame = [id: int, order_purchase_timestamp: timestamp ... 2 more fields]

scala> delivery_diff.show()
+-----+-----+-----+
|id|order_purchase_timestamp|order_delivered_customer_date|DeliveryDifference(InHours)|
+-----+-----+-----+
1 | 2017-10-02 10:56:00 | 2017-10-10 21:25:00 | 202.48333333333332 |
2 | 2018-07-24 20:41:00 | 2018-08-07 15:27:00 | 330.76666666666665 |
3 | 2018-08-08 08:38:00 | 2018-08-17 18:06:00 | 225.46666666666667 |
4 | 2017-11-18 19:28:00 | 2017-12-02 00:28:00 | 317.0 |
5 | 2018-02-13 21:18:00 | 2018-02-16 18:17:00 | 68.98333333333333 |
6 | 2017-07-09 21:57:00 | 2017-07-26 10:57:00 | 397.8 |
7 | 2017-04-11 12:22:00 | 2017-04-13 13:25:00 | 49.08 |
8 | 2017-05-16 13:10:00 | 2017-05-26 12:55:00 | 239.75 |
9 | 2017-01-23 19:29:00 | 2017-02-02 14:08:00 | 235.65 |
10 | 2017-07-29 11:55:00 | 2017-08-16 17:14:00 | 437.31666666666666 |
11 | 2017-05-16 19:41:00 | 2017-05-29 11:18:00 | 303.61666666666667 |
12 | 2017-07-13 19:58:00 | 2017-07-19 14:04:00 | 138.1 |
13 | 2018-06-07 10:06:00 | 2018-06-19 12:05:00 | 209.98333333333333 |
14 | 2018-07-25 17:44:00 | 2018-07-30 15:52:00 | 118.13333333333334 |
15 | 2018-03-01 14:14:00 | 2018-03-12 23:36:00 | 273.36666666666667 |
16 | 2018-06-07 19:03:00 | 2018-06-21 15:34:00 | 332.51666666666665 |
17 | 2018-01-02 19:00:00 | 2018-01-20 01:38:00 | 414.6333333333333 |
18 | 2017-12-26 23:41:00 | 2018-01-08 22:36:00 | 310.91666666666667 |
19 | 2017-11-21 00:03:00 | 2017-11-27 18:28:00 | 162.41666666666666 |
20 | 2017-10-26 15:54:00 | 2017-11-08 22:22:00 | 318.46666666666666 |
+-----+-----+-----+

only showing top 20 rows

scala> var avg_delivery = delivery_diff.agg(round(avg("DeliveryDifference(InHours)"),3).as("Average Time To Deliver(Hours)")).show()
+-----+
|Average Time To Deliver(Hours)|
+-----+
| 301.39 |
+-----+

avg_delivery: Unit = ()

```

new dataframe having conversion of 'order purchase date' to 'week number' :

[illegible]

Total sales :

```
scala> var weeklysales_df = weekly_df.groupBy("WeekNumber").agg(round(sum(col("order_products_value")),2).as("Total_Weekly_Sales")).orderBy("WeekNumber")
weeklysales_df: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, Total_Weekly_Sales: double]

scala> weeklysales_df.show()
+-----+-----+
|WeekNumber|Total_Weekly_Sales|
+-----+-----+
|1|173944.85|
|2|236810.92|
|3|255674.81|
|4|256544.96|
|5|265820.53|
|6|268544.45|
|7|256302.52|
|8|283705.55|
|9|308138.64|
|10|293062.7|
|11|288693.72|
|12|305988.97|
|13|258064.74|
|14|280830.14|
|15|258832.68|
|16|295689.26|
|17|293175.73|
|18|308817.36|
|19|363995.98|
|20|364922.4|
+-----+-----+

only showing top 20 rows
```

Total sales in each city :

```
scala> var weeklysales_city_df = weekly_df.groupBy(col("WeekNumber"),col("customer_city")).agg(round(sum(col("order_products_value")),2).as("Total_Weekly_Sales")).orderBy("WeekNumber")
weeklysales_city_df: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, customer_city: string ... 1 more field]

scala> weeklysales_city_df.show()
+-----+-----+
|WeekNumber|customer_city|Total_Weekly_Sales|
+-----+-----+
|1|Rancharia|19.9|
|1|Carapicuíba|173.38|
|1|Balneario Camboriu|101.8|
|1|RIO DE JANEIRO|12889.46|
|1|Ferraz de Vasconcelos|94.9|
|1|Areado|89.9|
|1|Luz|30.0|
|1|Amparo|400.69|
|1|Belem|1043.29|
|1|Diadema|81.56|
|1|Itaquaquecetuba|227.49|
|1|Louveira|35.0|
|1|Altinópolis|58.99|
|1|Sao Miguel Do Oeste|203.0|
|1|Analandia|27.9|
|1|Capitao Leonidas Marques|53.99|
|1|Penedo|149.99|
|1|Resende|219.0|
|1|Santa Maria|369.84|
|1|Sao Jose do Vale do Rio Preto|98.99|
+-----+-----+
only showing top 20 rows
```

Total sales in each state :

```
scala> var weeklysales_state_df = weekly_df.groupBy(col("WeekNumber"),col("customer_state")).agg(round(sum(col("order_products_value")),2).as("Total_Weekly_Sales")).orderBy("WeekNumber")
weeklysales_state_df: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, customer_state: string ... 1 more field]

scala> weeklysales_state_df.show()
+-----+-----+
|WeekNumber|customer_state|Total_Weekly_Sales|
+-----+-----+
|1|TO|1074.0|
|1|SE|40.89|
|1|RO|979.86|
|1|SC|4246.08|
|1|PI|438.7|
|1|RN|862.58|
|1|AM|133.79|
|1|PE|3388.4|
|1|PR|7578.1|
|1|MT|3468.43|
|1|BA|6745.95|
|1|MS|2744.11|
|1|MG|21649.59|
|1|AC|288.49|
|1|SP|69261.83|
|1|ES|2640.4|
|1|PB|1062.59|
|1|CE|2723.16|
|1|RJ|22600.0|
|1|PA|1501.98|
+-----+-----+
only showing top 20 rows
```

b. Orders

Total number of orders :

```
scala> var totalweekly_order = weekly_df.groupBy("WeekNumber").agg(count("*").as("Total_Weekly_Orders")).orderBy("WeekNumber")
totalweekly_order: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, Total_Weekly_Orders: bigint]

scala> totalweekly_order.show()
+-----+-----+
|WeekNumber|Total_Weekly_Orders|
+-----+-----+
|1|1439|
|2|1875|
|3|1964|
|4|1956|
|5|2077|
|6|2148|
|7|2060|
|8|2145|
|9|2412|
|10|2246|
|11|2221|
|12|2381|
|13|2047|
|14|2190|
|15|2005|
|16|2258|
|17|2359|
|18|2570|
|19|2802|
|20|2797|
+-----+-----+
only showing top 20 rows
```

City-wise order distribution :

```
scala> var weekly_city_order = weekly_df.groupBy(col("WeekNumber"),col("customer_city")).agg(count("*").as("Total_Weekly_Orders")).orderBy("WeekNumber")
weekly_city_order: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, customer_city: string ... 1 more field]

scala> weekly_city_order.show()
+-----+-----+-----+
|WeekNumber|customer_city|Total_Weekly_Orders|
+-----+-----+-----+
|1|Ferraz de Vasconc...|2|
|1|Americana|6|
|1|Balneario Camboriu|2|
|1|RIO DE JANEIRO|113|
|1|Belem|8|
|1|Santa Maria|4|
|1|Itaquaquecetuba|3|
|1|Sao Miguel Do Oeste|2|
|1|Cachoeira Alta|1|
|1|Sao Jose do Egito|1|
|1|Vitoria da Conquista|2|
|1|Papucaia|1|
|1|Fernando Prestes|1|
|1|Pontes e Lacerda|1|
|1|Itaocara|1|
|1|Rondonopolis|1|
|1|Bataguassu|1|
|1|Sao Jose dos Campos|5|
|1|Ilha Solteira|1|
|1|Cajamar|2|
+-----+-----+-----+
only showing top 20 rows
```

State-wise order distribution :

```
scala> var weekly_state_order = weekly_df.groupBy(col("WeekNumber"),col("customer_state")).agg(count("*").as("Total_Weekly_Orders")).orderBy("WeekNumber")
weekly_state_order: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, customer_state: string ... 1 more field]

scala> weekly_state_order.show()
+-----+-----+-----+
|WeekNumber|customer_state|Total_Weekly_Orders|
+-----+-----+-----+
|1|TO|5|
|1|SE|2|
|1|SC|42|
|1|PI|4|
|1|RO|6|
|1|RN|6|
|1|AM|3|
|1|PE|28|
|1|PR|78|
|1|MT|20|
|1|MS|20|
|1|MG|170|
|1|AC|2|
|1|SP|592|
|1|ES|27|
|1|PB|7|
|1|CE|19|
|1|RJ|195|
|1|PA|13|
|1|BA|52|
+-----+-----+-----+
only showing top 20 rows
```

Average review score per order :

```
scala> var weekly_avg_rev = weekly_df.groupBy(col("WeekNumber")).agg(round(sum(col("review_score"))/count("*"),3).as("Avg_Rev_per_Order")).orderBy("WeekNumber")
weekly_avg_rev: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, Avg_Rev_per_Order: double]

scala> weekly_avg_rev.show()
+-----+-----+
|WeekNumber|Avg_Rev_per_Order|
+-----+-----+
|1|4.054|
|2|4.06|
|3|4.03|
|4|4.076|
|5|4.09|
|6|4.041|
|7|4.057|
|8|4.042|
|9|4.053|
|10|4.054|
|11|4.059|
|12|4.034|
|13|4.071|
|14|4.08|
|15|4.039|
|16|4.029|
|17|4.029|
|18|4.105|
|19|4.043|
|20|4.044|
+-----+-----+
only showing top 20 rows
```

Average freight charges per order :

```
scala> var weekly_avg_fre = weekly_df.groupBy(col("WeekNumber")).agg(round(sum(col("order_freight_value"))/count("*"),3).as("Avg_Rev_per_Order")).orderBy("WeekNumber")
weekly_avg_fre: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, Avg_Rev_per_Order: double]

scala> weekly_avg_fre.show()
+-----+-----+
|WeekNumber|Avg_Rev_per_Order|
+-----+-----+
|1|21.869|
|2|22.24|
|3|21.588|
|4|21.808|
|5|21.738|
|6|21.306|
|7|21.871|
|8|22.863|
|9|21.608|
|10|21.881|
|11|21.646|
|12|22.257|
|13|21.792|
|14|21.503|
|15|21.064|
|16|21.736|
|17|21.162|
|18|21.233|
|19|21.614|
|20|22.053|
+-----+-----+
only showing top 20 rows
```

converting the three 'timestamp' columns to unified format:

```
scala> var df2 = date_df.withColumn("WeekNumber", weekofyear(col("PurchaseDate"))).withColumn("order_purchase_timestamp", to_timestamp(col("order_purchase_timestamp"), "dd/MM/yy H:mm")).withColumn("order_approved_at", to_timestamp(col("order_approved_at"), "dd/MM/yy H:mm")).withColumn("order_delivered_customer_date", to_timestamp(col("order_delivered_customer_date"), "dd/MM/yy H:mm"))
df2: org.apache.spark.sql.DataFrame = [id: int, order_status: string ... 17 more fields]

scala> df2.show()
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|id|order_status|order_products_value|order_freight_value|order_items_qty|customer_city|customer_state|customer_zip_code_prefix|product_name_length|product_description_length|product_photos_qty|review_score|order_purchase_t|
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
|1|delivered|79.0|17.8|1|Luziania|GO|728|50|201|2|5|2017-10-02|
|10:56:00|2017-10-02 11:07:00|2017-10-10 21:25:00|2017-10-02|2017-10-02|2017-10-10|40|
|2|delivered|119.9|27.16|1|Joinville|SC|892|50|511|3|5|2018-07-24|
|20:41:00|2018-07-26 03:24:00|2018-08-07 15:27:00|2018-07-24|2018-07-26|2018-08-07|30|
|3|delivered|519.99|41.69|1|Serra|ES|291|48|1156|2|1|2018-08-08|
|08:38:00|2018-08-08 08:55:00|2018-08-17 18:06:00|2018-08-08|2018-08-08|2018-08-17|32|
|4|delivered|29.5|17.92|1|RIO DE JANEIRO|RJ|222|21|207|2|4|2017-11-18|
|19:28:00|2017-11-18 19:45:00|2017-12-02 00:28:00|2017-11-18|2017-11-18|2017-12-02|46|
|5|delivered|26.77|23.11|1|Sao Paulo|SP|40|41|451|1|5|2018-02-13|
|21:18:00|2018-02-13 22:20:00|2018-02-16 18:17:00|2018-02-13|2018-02-13|2018-02-16|7|
|6|delivered|419.9|23.02|1|Santa Adelia|SP|159|46|692|3|5|2017-07-09|
|21:57:00|2017-07-09 22:10:00|2017-07-26 10:57:00|2017-07-09|2017-07-09|2017-07-26|27|
|7|delivered|65.0|16.21|1|Varginha|MG|370|47|893|3|5|2017-04-11|
|12:22:00|2017-04-13 13:25:00|2017-04-13 13:25:00|2017-04-11|2017-04-13|2017-04-13|15|
|8|delivered|29.99|19.82|1|Sao Paulo|SP|48|55|613|1|4|2017-05-16|
|13:10:00|2017-05-16 13:22:00|2017-05-26 12:55:00|2017-05-16|2017-05-16|2017-05-26|20|
|9|delivered|59.99|51.14|1|Carajas|PA|685|28|96|4|5|2017-01-23|
|18:29:00|2017-01-25 02:50:00|2017-02-02 14:08:00|2017-01-23|2017-01-25|2017-02-02|4|
|10|delivered|56.99|16.13|1|Resende|RJ|275|59|641|3|5|2017-07-29|
|11:55:00|2017-07-29 12:05:00|2017-08-16 17:14:00|2017-07-29|2017-07-29|2017-08-16|30|
|11|delivered|599.0|15.69|1|Guaratininga|SP|125|43|1686|2|4|2017-05-16|
|19:41:00|2017-05-16 19:50:00|2017-05-29 11:18:00|2017-05-16|2017-05-16|2017-05-29|20|
|12|delivered|62.0|16.19|1|Tijucas|SC|882|50|285|1|4|2017-07-13|
|19:58:00|2017-07-13 20:10:00|2017-07-19 14:04:00|2017-07-13|2017-07-13|2017-07-19|28|
|13|delivered|250.0|35.02|1|Mirassol|SP|151|57|771|1|1|2018-06-07|
```

Average time taken to approve the orders (order approved – order purchased) :

```
scala> var weekly_approval_diff = df2.withColumn("TimeDifference(InMinutes)", ((col("order_approved_at").cast("long") - col("order_purchase_timestamp").cast("long")) / 60).cast(DoubleType)).groupBy("WeekNumber").agg(avg("TimeDifference(InMinutes)").as("WeeklyAverageTimeToApprove(Minutes)")).orderBy("WeekNumber")
weekly_approval_diff: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, WeeklyAverageTimeToApprove(Minutes): double]

scala> weekly_approval_diff.show()
+-----+-----+
|WeekNumber|WeeklyAverageTimeToApprove(Minutes)|
+-----+-----+
|1|645.6681006916295|
|2|646.3057630732992|
|3|762.2146863844977|
|4|498.90439672801637|
|5|664.701686746988|
|6|600.4473438956197|
|7|559.6588465298142|
|8|574.5076923076923|
|9|557.215326970955|
|10|401.4129175946548|
|11|386.83520936515083|
|12|492.9088235294118|
|13|590.9213098729227|
|14|546.5223744292238|
|15|619.062874251497|
|16|1503.6458333333333|
|17|802.8473282442748|
|18|716.4544037412315|
|19|516.1725|
|20|532.5956381837683|
+-----+-----+
only showing top 20 rows

scala> var weekly_avg_approve = weekly_approval_diff.agg(round(avg("WeeklyAverageTimeToApprove(Minutes)"),3).as("Weekly AVG Approval Time(InMinutes)")).show()
+-----+
|Weekly AVG Approval Time(InMinutes)|
+-----+
|629.027|
+-----+

weekly_avg_approve: Unit = ()
```


Average order delivery time :

```
scala> var weekly_delivery_diff = df2.withColumn("TimeDifference(InHours)", ((col("order_delivered_customer_date").cast("long") - col("order_purchase_timestamp").cast("long")) / 3600).cast(DoubleType)).groupBy("WeekNumber").agg(avg("TimeDifference(InHours)").as("WeeklyAverageTimeToDeliver(Hours)")).orderBy("WeekNumber")
weekly_delivery_diff: org.apache.spark.sql.Dataset[org.apache.spark.sql.Row] = [WeekNumber: int, WeeklyAverageTimeToDeliver(Hours): double]

scala> weekly_delivery_diff.show()
+-----+-----+
|WeekNumber|WeeklyAverageTimeToDeliver(Hours)|
+-----+-----+
|1|328.11188406212657|
|2|333.7524894746478|
|3|328.2062041884815|
|4|335.1090483757061|
|5|358.5847804391221|
|6|371.29964521851343|
|7|372.8246552304069|
|8|409.42559897089575|
|9|427.27816314886836|
|10|403.0652822642667|
|11|383.0805335199878|
|12|323.54522983521264|
|13|328.86308505284353|
|14|299.66362570356443|
|15|296.0802390710383|
|16|301.6511227431345|
|17|296.3451316550927|
|18|277.55423344017095|
|19|257.7608514292731|
|20|279.1148763711543|
+-----+-----+
only showing top 20 rows

scala> var weekly_avg_delivery = weekly_delivery_diff.agg(round(avg("WeeklyAverageTimeToDeliver(Hours)"),3).as("Weekly AVG Delivery Time(InHours)")).show()
+-----+
|Weekly AVG Delivery Time(InHours)|
+-----+
|303.926|
+-----+
weekly_avg_delivery: Unit = ()
```

c. Total freight charges :

```
scala> var weeklyFreight_df = weekly_df.groupBy("WeekNumber").agg(round(sum(col("order_freight_value")),2).as("Total_Freight_Charges")).orderBy("WeekNumber").show()
+-----+-----+
|WeekNumber|Total_Freight_Charges|
+-----+-----+
|1|31469.52|
|2|41699.72|
|3|42398.0|
|4|42656.18|
|5|45150.2|
|6|45765.64|
|7|45054.13|
|8|49041.98|
|9|52119.31|
|10|49145.77|
|11|48075.61|
|12|52992.88|
|13|44608.61|
|14|47091.58|
|15|42233.19|
|16|49080.8|
|17|49920.09|
|18|54567.61|
|19|60562.31|
|20|61682.45|
+-----+-----+
only showing top 20 rows

weeklyFreight_df: Unit = ()
```

d. Distribution of freight charges in each city :

```
scala> var weeklyfreight_city_df = weekly_df.groupBy(col("WeekNumber"),col("customer_city")).agg(round(sum(col("order_freight_value")),2).as("City_Freight_Charges")).orderBy("WeekNumber").show()
+-----+-----+-----+
|WeekNumber|customer_city|City_Freight_Charges|
+-----+-----+-----+
|1|Ferraz de Vasconc...|36.81|
|1|Americana|93.16|
|1|Balneario Camboriu|42.56|
|1|RIO DE JANEIRO|2637.07|
|1|Bolem|176.16|
|1|Santa Maria|149.99|
|1|Itaquaquecetuba|80.58|
|1|Sao Miguel Do Oeste|45.63|
|1|Cachoeira Alta|15.66|
|1|Sao Jose do Egito|24.84|
|1|Vitoria da Conquista|39.54|
|1|Papuaca|16.21|
|1|Fernando Prestes|10.96|
|1|Pontes e Lacerda|20.01|
|1|Itaocara|18.09|
|1|Rondonopolis|16.79|
|1|Bataguassu|18.23|
|1|Sao Jose dos Campos|73.56|
|1|Ilha Solteira|49.3|
|1|Cajamar|26.65|
+-----+-----+-----+
only showing top 20 rows

weeklyfreight_city_df: Unit = ()
```