Module 3 - Data Integration and Aggregation

March 7, 2025

[1]: from pyspark.sql import SparkSession

```
spark = SparkSession.builder \
     .appName('OlistData') \
     .getOrCreate()
    25/02/28 09:03:58 WARN SparkSession: Using an existing Spark session; only
    runtime SQL configurations will take effect.
[2]: hdfs_path = '/olist/'
[3]: customers_df = spark.read.csv(hdfs_path + 'olist_customers_dataset.
     ⇔csv',header=True,inferSchema=True)
    orders_df = spark.read.csv(hdfs_path + 'olist_orders_dataset.
     ⇔csv', header=True, inferSchema=True)
    order_item_df = spark.read.csv(hdfs_path + 'olist_order_items_dataset.
      ⇒csv',header=True,inferSchema=True)
    payments_df = spark.read.csv(hdfs_path + 'olist_order_payments_dataset.
      →csv',header=True,inferSchema=True)
    reviews_df = spark.read.csv(hdfs_path + 'olist_order_reviews_dataset.
     →csv',header=True,inferSchema=True)
    products_df = spark.read.csv(hdfs_path + 'olist_products_dataset.
      ⇒csv',header=True,inferSchema=True)
    sellers_df = spark.read.csv(hdfs_path + 'olist_sellers_dataset.
     geolocation_df = spark.read.csv(hdfs_path + 'olist_geolocation_dataset.
      ⇔csv',header=True,inferSchema=True)
    category_translation_df = spark.read.csv(hdfs_path +_
      → 'product_category_name_translation.csv',header=True,inferSchema=True)
[4]: # Cache Frequently used Data for Better Performance
    orders_df.cache()
    customers_df.cache()
    order_item_df.cache()
```

```
[4]: DataFrame[order_id: string, order_item_id: int, product_id: string, seller_id:
      string, shipping_limit_date: timestamp, price: double, freight_value: double]
 []:
 [5]:
      orders_items_joined_df = orders_df.join(order_item_df,'order_id','inner')
 [6]: orders_items_products_df = orders_items_joined_df.

→join(products_df,'product_id','inner')
 [7]: orders_items_products_sellers_df = orders_items_products_df.
       [8]: full_orders_df = orders_items_products_sellers_df.
       →join(customers df, 'customer id', 'inner')
 [9]: # GEolocation Data
      full_orders df = full_orders_df.join(geolocation_df,full_orders_df.
       ⇒customer_zip_code_prefix == geolocation_df.

¬geolocation_zip_code_prefix, 'left')
[10]: |full_orders_df = full_orders_df.join(reviews_df,'order_id','left')
[11]: | full_orders_df = full_orders_df.join(payments_df,'order_id','left')
[12]: full orders df.cache()
     25/02/28 08:37:30 WARN package: Truncated the string representation of a plan
     since it was too large. This behavior can be adjusted by setting
     'spark.sql.debug.maxToStringFields'.
[12]: DataFrame[order id: string, customer id: string, seller id: string, product id:
      string, order_status: string, order_purchase_timestamp: timestamp,
      order_approved_at: timestamp, order_delivered_carrier_date: timestamp,
      order_delivered_customer_date: timestamp, order_estimated_delivery_date:
      timestamp, order_item_id: int, shipping_limit_date: timestamp, price: double,
      freight_value: double, product_category_name: string, product_name_lenght: int,
     product_description_lenght: int, product_photos_qty: int, product_weight_g: int,
     product_length_cm: int, product_height_cm: int, product_width_cm: int,
      seller zip code prefix: int, seller city: string, seller state: string,
      customer_unique_id: string, customer_zip_code_prefix: int, customer_city:
      string, customer_state: string, geolocation_zip_code_prefix: int,
      geolocation_lat: double, geolocation_lng: double, geolocation_city: string,
      geolocation_state: string, review_id: string, review_score: string,
      review_comment_title: string, review_comment_message: string,
      review_creation_date: string, review_answer_timestamp: string,
     payment_sequential: int, payment_type: string, payment_installments: int,
```

```
payment_value: double]
[18]: from pyspark.sql.functions import *
[14]: full orders df.printSchema()
     root
      |-- order_id: string (nullable = true)
      |-- customer_id: string (nullable = true)
      |-- seller_id: string (nullable = true)
      |-- product_id: string (nullable = true)
      |-- order_status: string (nullable = true)
      |-- order purchase timestamp: timestamp (nullable = true)
      |-- order_approved_at: timestamp (nullable = true)
      |-- order delivered carrier date: timestamp (nullable = true)
      |-- order_delivered_customer_date: timestamp (nullable = true)
      |-- order estimated delivery date: timestamp (nullable = true)
      |-- order_item_id: integer (nullable = true)
      |-- shipping_limit_date: timestamp (nullable = true)
      |-- price: double (nullable = true)
      |-- freight_value: double (nullable = true)
      |-- product_category_name: string (nullable = true)
      |-- product_name_lenght: integer (nullable = true)
      |-- product_description_lenght: integer (nullable = true)
      |-- product_photos_qty: integer (nullable = true)
      |-- product_weight_g: integer (nullable = true)
      |-- product_length_cm: integer (nullable = true)
      |-- product_height_cm: integer (nullable = true)
      |-- product_width_cm: integer (nullable = true)
      |-- seller zip code prefix: integer (nullable = true)
      |-- seller_city: string (nullable = true)
      |-- seller state: string (nullable = true)
      |-- customer_unique_id: string (nullable = true)
      |-- customer_zip_code_prefix: integer (nullable = true)
      |-- customer_city: string (nullable = true)
      |-- customer_state: string (nullable = true)
      |-- geolocation_zip_code_prefix: integer (nullable = true)
      |-- geolocation_lat: double (nullable = true)
      |-- geolocation_lng: double (nullable = true)
      |-- geolocation_city: string (nullable = true)
      |-- geolocation state: string (nullable = true)
      |-- review_id: string (nullable = true)
      |-- review score: string (nullable = true)
      |-- review_comment_title: string (nullable = true)
      |-- review comment message: string (nullable = true)
      |-- review_creation_date: string (nullable = true)
```

|-- review_answer_timestamp: string (nullable = true)
|-- payment_sequential: integer (nullable = true)

```
|-- payment_type: string (nullable = true)
     |-- payment_installments: integer (nullable = true)
     |-- payment_value: double (nullable = true)
[19]: # Total Revenues Per Seller
     seller_revenue_df = full_orders_df.groupBy('seller_id').agg(sum('price'))
[20]: seller_revenue_df.show(5)
     [Stage 26:======>>(199 + 1) / 200]
    +----+
               seller_id|
                               sum(price) |
    +----+
     |e63e8bfa530fb1691...|219481.0000000035|
    |ff063b022a9a0aab9...|
     |a49928bcdf77c55c6...|1220624.6000000054|
    |33ac3e28642ab8bda...| 615628.8499999995|
    |7aa4334be125fcdd2...| 2509294.489999999|
    +----+
    only showing top 5 rows
[]:
[]: # Total Orders Per Custoemr
     # Average Review Score Per Seller
     # Most Sold Products ( Top 10 )
     # Top Custoemrs By Spending
[]:
        Optimized Joins For Data integration
[8]: from pyspark.sql.functions import *
     orders_items_joined_df = orders_df.join(order_item_df,'order_id','inner')
[6]: orders_items_products_df = orders_items_joined_df.

→join(products_df, 'product_id', 'inner')
[9]: orders_items_products_sellers_df = orders_items_products_df.

→join(broadcast(sellers_df), 'seller_id', 'inner')
```

```
[10]: full_orders_df = orders_items_products_sellers_df.

→join(customers_df, 'customer_id', 'inner')
[11]: # GEolocation Data
      full orders df = full orders df.join(broadcast(geolocation df),full orders df.
       ⇒customer zip code prefix == geolocation df.

¬geolocation_zip_code_prefix, 'left')
[12]: full_orders_df = full_orders_df.join(broadcast(reviews_df), 'order_id', 'left')
[13]: full_orders_df = full_orders_df.join(payments_df,'order_id','left')
[15]: full_orders_df.cache()
     25/02/28 09:07:07 WARN package: Truncated the string representation of a plan
     since it was too large. This behavior can be adjusted by setting
     'spark.sql.debug.maxToStringFields'.
[15]: DataFrame[order_id: string, customer_id: string, seller_id: string, product_id:
      string, order_status: string, order_purchase_timestamp: timestamp,
      order_approved_at: timestamp, order_delivered_carrier_date: timestamp,
      order_delivered_customer_date: timestamp, order_estimated_delivery_date:
      timestamp, order item id: int, shipping limit date: timestamp, price: double,
      freight_value: double, product_category_name: string, product_name_lenght: int,
      product_description_lenght: int, product_photos_qty: int, product_weight_g: int,
     product_length_cm: int, product_height_cm: int, product_width_cm: int,
      seller zip code prefix: int, seller city: string, seller state: string,
      customer_unique_id: string, customer_zip_code_prefix: int, customer_city:
      string, customer_state: string, geolocation_zip_code_prefix: int,
      geolocation_lat: double, geolocation_lng: double, geolocation_city: string,
      geolocation_state: string, review_id: string, review_score: string,
      review_comment_title: string, review_comment_message: string,
      review_creation_date: string, review_answer_timestamp: string,
     payment_sequential: int, payment_type: string, payment_installments: int,
     payment_value: double]
```

2 Aggregation

```
[18]: # Total Orders Per Customer

customer_order_count_df = full_orders_df.groupBy('customer_id')\
    .agg(count('order_id').alias('total_orders'))\
    .orderBy(desc('total_orders'))

customer_order_count_df.show(5)
```

[Stage 25:======>

(1 + 1) / 2

```
+-----+
| customer_id|total_orders|
+-----+
|351e40989da90e704...| 11427|
|50920f8cd0681fd86...| 10752|
|9b43e2a62de9bab3a...| 8556|
|270c23a11d024a44c...| 8001|
|d3e82ccec3cb5f956...| 6876|
+------+
only showing top 5 rows
```

```
[20]: #Average Review Score Per Seller
seller_review_df = full_orders_df.groupBy('seller_id')\
.agg(avg('review_score').alias('avg_review_score'))\
.orderBy(desc('avg_review_score'))
seller_review_df.show()
```

(1 + 1) / 2

```
[Stage 28:========>
+----+
          seller_id|avg_review_score|
+----+
|a353b1083c9863d75...|
                             5.01
|bd43e172d599bed47...|
                             5.01
|a61cc04793308395a...|
                             5.01
|1f2eebc0e970fd3c4...|
                             5.01
|7ad41305e96a6cab8...|
                             5.0|
|64c9a1db4e73e19aa...|
                             5.01
|89757206b887aed36...|
                             5.0|
|a56a8043ebf66e421...|
                             5.0
|f1fdf2d1318657575...|
                             5.0
1929f342384a6607af...
                             5.01
|05a48cc8859962767...|
                             5.0
|7238f877570096ae4...|
                             5.0
|1cd9e0cc1839d5551...|
                             5.0
                             5.0
|b5b800c4065bebf4d...|
|392f7f2c797e4dc07...|
                             5.0
|05ca864204d09595a...|
                             5.01
|94d76e96eedd97625...|
                             5.0
|9d213f303afae4983...|
                             5.01
|d598f929fc44e1e38...|
                             5.0|
|0ad80de75c8113263...|
                             5.01
+----+
```

only showing top 20 rows

```
[21]: # Top 10 Most Sold Products
     top_products_df = full_orders_df.groupBy('product_id')\
      .agg(count('order_id').alias('total_sold'))\
      .orderBy(desc('total_sold'))\
      .limit(10)
     top_products_df.show()
     (1 + 1) / 2
               product_id|total_sold|
     |aca2eb7d00ea1a7b8...|
                             86740|
     |422879e10f4668299...|
                             81110
     |99a4788cb24856965...|
                             78775
     |389d119b48cf3043d...|
                             602481
     |d1c427060a0f73f6b...|
                             592741
     |368c6c730842d7801...|
                             58358 l
     |53759a2ecddad2bb8...|
                             52654
     |53b36df67ebb7c415...|
                             52105
     |154e7e31ebfa09220...|
                             427001
     |3dd2a17168ec895c7...|
                             40787
 []: # Top 10 Customer By Spending
 []:
        Window Function and Ranking
[23]: from pyspark.sql.window import Window
[24]: # Dense Rank for Sellers Based on Revenue
     window_spec = Window.partitionBy('seller_id').orderBy(desc('price'))
[26]: # Rank Top Selling Products Per seller
     top_seller_products_df = full_orders_df.withColumn('rank',rank().
       ⇔over(window_spec)).filter(col('rank')<=5)</pre>
```

```
top_seller_products_df.select('seller_id','price','rank').show()
    [Stage 39:>
                                                                             (0 + 1) / 1]
                 seller_id|price|rank|
    |0015a82c2db000af6...|895.0|
                                   11
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1 |
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   11
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   11
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1|
    |0015a82c2db000af6...|895.0|
                                   1 |
    +----+
    only showing top 20 rows
[]:
[]:
     # Dense Rank for Sellers Based on Revenue
[]:
[]:
[]:
[]:
[]:
```

4 Advance Aggregation and Enrichment

```
[]: full_orders_df
[28]: # Total Revenue & Average Order Value (AOV) per Customer
     customer_spending_df = full_orders_df.groupBy('customer_id')\
      .agg(
         count('order_id').alias('total_orders'),
         sum('price').alias('total_spent'),
         round(avg('price'),2).alias('AOV')
     )\
      .orderBy(desc('total_spent'))
     customer_spending_df.show()
                                                                       (1 + 1) / 2
     -----+
              customer_id|total_orders|
                                             total_spent|
         ------
     |d3e82ccec3cb5f956...|
                                6876
                                             6662844.0| 969.0|
     |df55c14d1476a9a34...|
                                 743|
                                             3565657.0| 4799.0|
     |fe5113a38e3575c04...|
                                2292|
                                             3293604.0 | 1437.0 |
     |ec5b2ba62e5743423...|
                                1428|
                                             2556120.0 | 1790.0 |
     |63b964e79dee32a35...|
                                6072
                                              2501664.0 | 412.0 |
     |46bb3c0b1a65c8399...|
                                 748
                                             2336752.0 | 3124.0 |
     105455dfa7cd02f13d...|
                                2184 | 2160194.400000087 | 989.1
     |3690e975641f01bd0...|
                                             2124498.0 | 2649.0 |
                                 802
     |349509b216bd5ec11...|
                                 743 l
                                             1923627.0 | 2589.0 |
     |695476b5848d64ba0...|
                                 687 | 1820543.1299999943 | 2649.99 |
                                             1755520.0 | 2110.0 |
     |73236a0796f53d60d...|
                                 832 l
                                             1676400.0| 2200.0|
     cc803a2c412833101...|
                                 762
                                5820 | 1658641.7999999512 | 284.99 |
     |1ff773612ab8934db...|
     |fced842c7dad61e8c...|
                                              1654898.0 | 2749.0 |
```

only showing top 20 rows

|1ecb47d23dc8203cd...|

|de832e8dbb1f588a4...|

|803cd9b04f9cd252c...|

|d72181923840c8895...|

|06d478ba352a27a51...|

|0049e8442c2a3e4a8...|

```
[29]: 6662844/6876
```

1146|

1204

1164 | 1629588.3599999903 | 1399.99 |

2190 | 1584990.5999999817 | 723.74 |

2721 | 1488114.8999999566 | 546.9 |

1512312.0 | 3099.0 |

1461150.0 | 1275.0 |

1444800.0 | 1200.0 |

```
[29]: 969.0
[]:
[31]: # Seller Performance Metrics ( Revenue, Average Review, Order Count)
     seller_performance_df = full_orders_df.groupBy('seller_id') \
     .agg(
        count('order_id').alias('total_orders'),
        sum('price').alias('total revenue'),
        round(avg('review_score'),2).alias('avg_review_score'),
        round(stddev('price'),2).alias('price variability')
     )\
     .orderBy(desc('total revenue'))
[32]: seller_performance_df.show()
     [Stage 43:=========>
                                                                 (1 + 1) / 2
     +-----
               seller_id|total_orders|
    total_revenue|avg_review_score|price_variability|
    +----+
    |4869f7a5dfa277a7d...| 184587|3.6138717319998816E7|
                                                               4.091
    111.65
    |53243585a1d6dc264...|
                           54514 | 3.429159295000016E7 |
                                                               4.12
    499.651
    |4a3ca9315b744ce9f...| 330661| 3.375957084003399E7|
                                                               3.77
    59.37
    |7c67e1448b00f6e96...|
                          233306 | 3.2282321790014144E7 |
                                                               3.42
    50.39|
    |fa1c13f2614d7b5c4...|
                          87686 | 3.013938631000357E7
                                                               4.38
    307.71
    |da8622b14eb17ae28...| 264433| 2.98576697300434E7|
                                                               3.981
    72.921
    |7e93a43ef30c4f03f...|
                          50226 | 2.631570630000493E7 |
                                                               4.15
    377.241
    |1025f0e2d44d7041d...| 229587|2.2937518520012498E7|
                                                               3.891
    84.31
    |46dc3b2cc0980fb8e...|
                           90426 | 2.179177329001596E7 |
                                                               4.18
    187.49
    |955fee9216a65b617...| 232364|2.0964410670014285E7|
                                                               4.04|
    84.94
    |7a67c85e85bb2ce85...|
                          167231|2.0312794890029624E7|
                                                               4.26|
    56.231
     |620c87c171fb2a6dd...|
                           142232 | 2.011983960002556E7 |
                                                               4.36
```

```
100.45
    |7d13fca1522535862...| 88807| 1.815688191000456E7|
                                                        4.07|
    151.18
    |a1043bafd471dff53...| 132672|1.7662675980011847E7|
                                                        4.25
    37.19
    |6560211a19b47992c...|
                       286539|1.7315932900000416E7|
                                                        3.861
    35.04
    |edb1ef5e36e0c8cd8...| 38945|1.6624835150005734E7|
                                                        4.431
    460.851
    |1f50f920176fa81da...| 297292|1.6497454440035844E7|
                                                        4.041
    7.39|
    |5dceca129747e92ff...| 50420|1.4910548340005763E7|
                                                        4.17|
    299.84
    |cc419e0650a3c5ba7...| 256032|1.4751464500039315E7|
                                                        4.07
    22.67|
    |3d871de0142ce09b7...| 175876|1.4184525300005388E7|
                                                        4.15
    38.14
    +-----
    only showing top 20 rows
[34]: # Product Popularity Metrics
    product_metrics_df = full_orders_df.groupBy('product_id')\
     .agg(
       count('order_id').alias('total_sales'),
       sum('price').alias('total_revenue'),
       round(avg('price'),2).alias('avg_price'),
       round(stddev('price'),2).alias('price_volatility'),\
        collect_set('seller_id').alias('unique_sellers')
    ) \
     .orderBy(desc('total_sales'))
[35]: product_metrics_df.show()
    (1 + 1) / 2
    product_id|total_sales| total_revenue|avg_price|price_volatility|
    unique_sellers|
    +-----
    +----+
    |aca2eb7d00ea1a7b8...|
                      86740 | 6164630 . 2999962345 |
    3.17 | [955fee9216a65b61... |
```

81110 | 4442791.509997333 |

54.77

|422879e10f4668299...|

4.46 [1f50f920176fa81d		
99a4788cb24856965	78775 6921762.709995905	87.87
4.08 [4a3ca9315b744ce9		
389d119b48cf3043d	60248 3280533.129998912	54.45
4.37 [1f50f920176fa81d		
d1c427060a0f73f6b	59274 8220103.330002628	138.68
16.58 [a1043bafd471dff5	·	·
368c6c730842d7801	58358 3181698.899999065	54.52
4.59 [1f50f920176fa81d		
53759a2ecddad2bb8	52654 2893017.499999481	54.94
4.52 [1f50f920176fa81d		
53b36df67ebb7c415	52105 6159887.409998229	118.22
20.13 [7d13fca152253586		
154e7e31ebfa09220	42700 962160.9999997382	22.53
1.92 [cc419e0650a3c5ba		
3dd2a17168ec895c7	40787 6116941.299997734	149.97
0.85 [de722cd6dad950a9		
e53e557d5a159f5aa	39516 3329353.9499996677	84.25
11.32 [6973a06f484aacf4		
2b4609f8948be1887	36179 3171618.7699996904	87.66
4.22 [cc419e0650a3c5ba		
35afc973633aaeb6b	31206 2735668.999999728	87.66
3.32 [d20b021d3efdf267		
e0d64dcfaa3b6db5c	31153 5226407.629999666	167.77
30.9 [7d13fca152253586		
42a2c92a0979a949c	30486 1810926.0000002119	59.4
0.64 [813348c996469b40	0004014700000 04000765	F0 011
7c1bd920dbdf22470	29018 1739338.8199997821	59.94
2.77 [cc419e0650a3c5ba	2070000	106 501
a62e25e09e05e6faf	28898 3079869.0	106.58
1.5 [634964b17796e643 5a848e4ab52fd5445	28737 3534363.6299997577	122.99
0.0 [c826c40d7b19f62a	20131 3034303.0299991311	122.33
c4baedd846ed09b85	28166 2802044.6499998467	99.48
11.9 [a1043bafd471dff5]	20100 2002044.0433330407	<i>33</i> .±0
b532349fe46b38fbc	27176 993089.5700001451	36.54
1.92 [1025f0e2d44d7041	2.1.0 00000.0100001401	00.01
		+
++		
only showing top 20 rows		
· • •		

[]: # Monthly Revenue and Order Count Trend ----> HW

order_purchase_timestamp ---> month

```
total_orders
     total_revenue
     avg_order_value
     min_order_value
     max_orderValues
[]:
[]:
[39]: # Customer Retention Analysis (First & Last Order)
     customer_retention_df = full_orders_df.groupBy('customer_id')\
     .agg(
         first('order_purchase_timestamp').alias('first_order_date'),
         last('order_purchase_timestamp').alias('last_order_date'),
         count('order_id').alias('total_orders'),
         round(avg('price'),2).alias('aov')
     )\
     .orderBy(desc('total_orders'))
[40]: customer_retention_df.show()
                                                                   (1 + 1) / 2
     -+
     customer_id| first_order_date| last_order_date|total_orders|
    aov
     |351e40989da90e704...|2017-07-13 10:42:37|2017-07-13 10:42:37|
                                                                  11427
     |50920f8cd0681fd86...|2018-01-27 11:28:32|2018-01-27 11:28:32|
                                                                  10752
    43.821
     |9b43e2a62de9bab3a...|2017-05-25 22:27:50|2017-05-25 22:27:50|
                                                                   8556
    26.4
     |270c23a11d024a44c...|2017-08-08 20:26:31|2017-08-08 20:26:31|
                                                                   8001
     |d3e82ccec3cb5f956...|2017-03-18 14:28:34|2017-03-18 14:28:34|
                                                                   6876
     |5c87184371002d49e...|2018-01-05 19:15:37|2018-01-05 19:15:37|
                                                                   6876
    12.49
     |d5f2b3f597c7ccafb...|2017-12-13 14:21:15|2017-12-13 14:21:15|
                                                                   6706
     c2f18647725395af4...|2018-03-06 19:21:47|2018-03-06 19:21:47|
                                                                   66121
     |24e7dc2ff8c071263...|2017-11-24 16:16:45|2017-11-24 16:16:45|
                                                                   6597|
```

```
59.21
    |7bb57d182bdc11653...|2018-04-02 17:11:30|2018-04-02 17:11:30|
                                                                         62581
    |d22f25a9fadfb1abb...|2018-05-12 12:28:58|2018-05-12 12:28:58|
                                                                         6072
    |63b964e79dee32a35...|2018-02-14 16:34:27|2018-02-14 16:34:27|
                                                                         6072
    |1ff773612ab8934db...|2018-04-19 13:54:06|2018-04-19 13:54:06|
    5820 | 284.99 |
    |13aa59158da63ba0e...|2017-09-23 14:56:45|2017-09-23 14:56:45|
                                                                         52061
    79.99
    |78fc46047c4a639e8...|2017-11-28 22:24:18|2017-11-28 22:24:18|
    5200 | 109.97 |
    |dd3f1762eb601f41c...|2018-07-18 12:59:21|2018-07-18 12:59:21|
    4992 | 179.99 |
    |a193aa8d905b8e246...|2018-02-12 18:04:28|2018-02-12 18:04:28|
                                                                         48961
    9.991
    |9eb3d566e87289dcb...|2018-06-08 16:42:11|2018-06-08 16:42:11|
                                                                         4872
    5.11
    |2ba91e12e5e4c9f56...|2017-11-25 13:54:39|2017-11-25 13:54:39|
                                                                        47521
    |1b2ab6eda1946a6ff...|2017-11-24 10:41:43|2017-11-24 10:41:43|
                                                                        4728
    only showing top 20 rows
[]: # HW - Correct the last order date
[]:
[]:
[]:
      Extended Enrichment
[]:
```

```
14
```

full_orders_df = full_orders_df.withColumn('is_delivered',__

when(col('order status') == 'delivered', lit(1)).otherwise(lit(0)))\

[43]: # Order Status Flags

[46]: full_orders_df.where(full_orders_df['order_status'] == 'canceled').

select('order_status', 'is_delivered', 'is_canceled').show(100)

+			+
order_	_status is +	_delivered	is_canceled
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
ca	anceled	0	1
ca	anceled	0	1
ca	anceled	0	1
ca	anceled	0	1
l ca	anceled	0	1
ca	anceled	0	1
ca	anceled	0	1
ca	anceled	0	1
ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
l ca	anceled	0	1
•	anceled	0	•
l ca	anceled	0	•
l ca	anceled	0	
l ca	anceled	0	•
	anceled	0	
l ca	anceled	0	
	anceled	0	
l ca	anceled	0	1

1	canceled	01	1
1	canceled	01	1
1	canceled	0	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	0	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
1	canceled	01	1
	canceled	0	1

```
1|
canceled
                      01
canceled
                      01
                                    1|
                      01
canceled
                                    1 |
canceled|
                      0|
                                    1|
canceled|
                      01
                                    1 l
canceled|
                      0|
                                    1|
canceled|
                      0|
                                    1|
canceled
                      01
                                    1 l
canceled|
                      0|
                                    11
canceled|
                      0|
                                    1|
canceled|
                      0|
                                    1|
canceled|
                      01
                                    1|
                      0|
canceled|
                                    1|
```

only showing top 100 rows

```
[47]: # Order Revenue Calcualtion

full_orders_df = full_orders_df.

withColumn('order_revenue',col('price')+col('freight_value'))
```

[48]: full_orders_df.select('price','freight_value','order_revenue').show()

```
|price|freight_value|order_revenue|
+----+
|29.99|
               8.72
                            38.71
[29.99]
                            38.71|
               8.72
129.991
               8.72
                            38.71
[29.99]
               8.72|
                            38.71
129.991
               8.72
                            38.71
[29.99]
               8.72|
                            38.71
[29.99]
               8.72
                            38.71
[29.99]
               8.72
                            38.71
                            38.71
[29.99]
               8.72
[29.99]
               8.72
                            38.71
                            38.71
[29.99]
               8.72
[29.99]
               8.72
                            38.71
|29.99|
               8.72|
                            38.71|
[29.99]
               8.72
                            38.71
[29.99]
               8.72|
                            38.71|
[29.99]
               8.72
                            38.71
               8.72|
[29.99]
                            38.71
129.991
               8.72
                            38.71
[29.99]
               8.72
                            38.71
129.991
               8.72|
                            38.71
```

```
[52]: customer_spending_df.printSchema()
     root
      |-- customer_id: string (nullable = true)
      |-- total_orders: long (nullable = false)
      |-- total_spent: double (nullable = true)
      |-- AOV: double (nullable = true)
[57]: # Customer Segmentation based on spending
     customer_spending_df = customer_spending_df.withColumn(
          'customer_segment',
         when(col('AOV') >=1200,"High-Value")
          .when( (col('AOV')<1200) & (col('AOV') >=700), 'Medium_Value')
          .otherwise('Low-Value'))
[58]: customer_spending_df.show()
     (1 + 1) / 2
     +----+
              customer_id|total_orders|
                                             total_spent|
                                                             AOV | customer_segment |
     |d3e82ccec3cb5f956...|
                                6876
                                             6662844.0| 969.0|
                                                                   Medium_Value
     |df55c14d1476a9a34...|
                                743|
                                             3565657.0| 4799.0|
                                                                     High-Value
     |fe5113a38e3575c04...|
                                2292|
                                             3293604.0 | 1437.0 |
                                                                     High-Value|
     ec5b2ba62e5743423...|
                                             2556120.0 | 1790.0 |
                                                                     High-Value |
                                1428
     |63b964e79dee32a35...|
                                             2501664.0| 412.0|
                                                                      Low-Value|
                                6072
     46bb3c0b1a65c8399...|
                                 748
                                             2336752.0 | 3124.0 |
                                                                     High-Value
     105455dfa7cd02f13d...|
                                2184 | 2160194.400000087 | 989.1
                                                                   Medium_Value |
                                             2124498.0| 2649.0|
                                                                     High-Value
     |3690e975641f01bd0...|
                                 802|
     |349509b216bd5ec11...|
                                 743|
                                             1923627.0 | 2589.0
                                                                     High-Value
                                                                     High-Value|
     |695476b5848d64ba0...|
                                 687 | 1820543.1299999943 | 2649.99 |
     |73236a0796f53d60d...|
                                 832 l
                                             1755520.0 | 2110.0
                                                                     High-Value
                                                                     High-Value |
     cc803a2c412833101...
                                 762
                                             1676400.0| 2200.0|
     |1ff773612ab8934db...|
                                5820 | 1658641.7999999512 | 284.99 |
                                                                      Low-Value
     |fced842c7dad61e8c...|
                                             1654898.0 | 2749.0 |
                                                                     High-Value |
                                1164 | 1629588.3599999903 | 1399.99 |
     |1ecb47d23dc8203cd...|
                                                                     High-Value
     |de832e8dbb1f588a4...|
                                2190 | 1584990.5999999817 | 723.74 |
                                                                   Medium_Value |
     |803cd9b04f9cd252c...|
                                 488
                                             1512312.0 | 3099.0 |
                                                                     High-Value |
     |d72181923840c8895...|
                                2721 | 1488114.8999999566 | 546.9 |
                                                                     Low-Value
     |06d478ba352a27a51...|
                                1146|
                                             1461150.0 | 1275.0 |
                                                                     High-Value |
     |0049e8442c2a3e4a8...|
                                1204
                                             1444800.0 | 1200.0 |
                                                                     High-Value
     +----+
     only showing top 20 rows
```

```
[59]: full_orders_df = full_orders_df.join(customer_spending_df.
       select('customer_id','customer_segment'),'customer_id',how='left')
[62]: full_orders_df.select('customer_id','customer_segment').show()
       -----+
               customer_id|customer_segment|
     +----+
     |9ef432eb625129730...|
                                Low-Value|
                                Low-Value |
     |9ef432eb625129730...|
     |9ef432eb625129730...|
                                Low-Value |
     |9ef432eb625129730...|
                                Low-Value
     |9ef432eb625129730...|
                                Low-Value
     |9ef432eb625129730...|
                                Low-Value
     |9ef432eb625129730...|
                                Low-Value
                                Low-Value
     |9ef432eb625129730...|
     |9ef432eb625129730...|
                                Low-Value
                                Low-Value
     |9ef432eb625129730...|
     |9ef432eb625129730...|
                                Low-Value
     |9ef432eb625129730...|
                                Low-Value
     |9ef432eb625129730...|
                                Low-Value
     |9ef432eb625129730...|
                                Low-Value |
     |9ef432eb625129730...|
                                Low-Value
                                Low-Value |
     |9ef432eb625129730...|
     |9ef432eb625129730...|
                                Low-Value
                                Low-Value |
     |9ef432eb625129730...|
     |9ef432eb625129730...|
                                Low-Value |
     |9ef432eb625129730...|
                                Low-Value|
     only showing top 20 rows
[63]: full_orders_df.select('order_purchase_timestamp').show()
     |order_purchase_timestamp|
     +----+
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33
```

```
2017-10-02 10:56:33
           2017-10-02 10:56:33|
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33|
           2017-10-02 10:56:33
           2017-10-02 10:56:33|
           2017-10-02 10:56:33|
           2017-10-02 10:56:33
           2017-10-02 10:56:33
           2017-10-02 10:56:33
         -----+
     only showing top 20 rows
[65]: #Hourly Order Distribution
     full_orders_df = full_orders_df.
       →withColumn('hour_of_day',expr('hour(order_purchase_timestamp)'))
[67]: full_orders_df.select('order_purchase_timestamp', 'hour_of_day').show()
```

```
|order_purchase_timestamp|hour_of_day|
     2017-10-02 10:56:33
                                   10 l
                                   10|
     2017-10-02 10:56:33
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10 l
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10 l
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   101
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33|
                                   10|
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33
                                   10|
     2017-10-02 10:56:33|
                                   10|
```

```
+----+
     only showing top 20 rows
[72]: # Weekday vs Weekend Order
     full_orders_df = full_orders_df.withColumn('order_day_type',\
      →when(dayofweek('order_purchase_timestamp').isin(1,7),lit('Weekend')).
      ⇔otherwise(lit('weekday')))
[73]: full_orders_df.select('order_purchase_timestamp','order_day_type').show()
        ----+
     |order_purchase_timestamp|order_day_type|
          2017-10-02 10:56:33
                                   weekday|
          2017-10-02 10:56:33
                                   weekday|
          2017-10-02 10:56:33
                                   weekday|
          2017-10-02 10:56:33|
                                   weekday|
          2017-10-02 10:56:33
                                   weekday|
          2017-10-02 10:56:33|
                                   weekday|
          2017-10-02 10:56:33
                                   weekday|
                                   weekday|
          2017-10-02 10:56:33
          2017-10-02 10:56:33|
                                   weekday|
          2017-10-02 10:56:33
                                   weekday|
         ----+
     only showing top 20 rows
[74]: full_orders_df.printSchema()
    root
      |-- customer_id: string (nullable = true)
      |-- order_id: string (nullable = true)
      |-- seller_id: string (nullable = true)
```

2017-10-02 10:56:33

```
|-- product_id: string (nullable = true)
|-- order_status: string (nullable = true)
|-- order_purchase_timestamp: timestamp (nullable = true)
|-- order_approved_at: timestamp (nullable = true)
|-- order delivered carrier date: timestamp (nullable = true)
|-- order_delivered_customer_date: timestamp (nullable = true)
|-- order estimated delivery date: timestamp (nullable = true)
|-- order_item_id: integer (nullable = true)
|-- shipping_limit_date: timestamp (nullable = true)
|-- price: double (nullable = true)
|-- freight_value: double (nullable = true)
|-- product_category_name: string (nullable = true)
|-- product_name_lenght: integer (nullable = true)
|-- product_description_lenght: integer (nullable = true)
|-- product_photos_qty: integer (nullable = true)
|-- product_weight_g: integer (nullable = true)
|-- product_length_cm: integer (nullable = true)
|-- product_height_cm: integer (nullable = true)
|-- product_width_cm: integer (nullable = true)
|-- seller zip code prefix: integer (nullable = true)
|-- seller_city: string (nullable = true)
|-- seller state: string (nullable = true)
|-- customer_unique_id: string (nullable = true)
|-- customer zip code prefix: integer (nullable = true)
|-- customer_city: string (nullable = true)
|-- customer_state: string (nullable = true)
|-- geolocation_zip_code_prefix: integer (nullable = true)
|-- geolocation_lat: double (nullable = true)
|-- geolocation_lng: double (nullable = true)
|-- geolocation_city: string (nullable = true)
|-- geolocation_state: string (nullable = true)
|-- review_id: string (nullable = true)
|-- review_score: string (nullable = true)
|-- review_comment_title: string (nullable = true)
|-- review comment message: string (nullable = true)
|-- review creation date: string (nullable = true)
|-- review_answer_timestamp: string (nullable = true)
|-- payment_sequential: integer (nullable = true)
|-- payment_type: string (nullable = true)
|-- payment_installments: integer (nullable = true)
|-- payment_value: double (nullable = true)
|-- is_delivered: integer (nullable = false)
|-- is_canceled: integer (nullable = false)
|-- order_revenue: double (nullable = true)
|-- customer_segment: string (nullable = true)
|-- hour_of_day: integer (nullable = true)
|-- order_day_type: string (nullable = false)
```

```
[]: # a new column frieght category based on freight value --> low, med or high
 []: # Order Volume by Customer State
 []:
[75]:
      !hadoop fs -mkdir /olist/processed/
[79]: full_orders_df.write.mode('overwrite').parquet('/olist/processed')
[78]:
     !hadoop fs -ls -h /olist/processed/
     Found 11 items
                  2 root hadoop
                                         0 2025-02-28 14:57 /olist/processed/ SUCCESS
     -rw-r--r--
                                    30.8 M 2025-02-28 14:57 /olist/processed/part-
     -rw-r--r--
                  2 root hadoop
     00000-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
     -rw-r--r--
                  2 root hadoop
                                    31.1 M 2025-02-28 14:57 /olist/processed/part-
     00001-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
     -rw-r--r--
                  2 root hadoop
                                    30.3 M 2025-02-28 14:57 /olist/processed/part-
     \tt 00002-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
                                    31.1 M 2025-02-28 14:57 /olist/processed/part-
     -rw-r--r--
                  2 root hadoop
     00003-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
                                    31.1 M 2025-02-28 14:57 /olist/processed/part-
     -rw-r--r--
                  2 root hadoop
     00004-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
     -rw-r--r--
                  2 root hadoop
                                    18.6 M 2025-02-28 14:57 /olist/processed/part-
     00005-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
                  2 root hadoop
                                    19.2 M 2025-02-28 14:57 /olist/processed/part-
     -rw-r--r--
     00006-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
     -rw-r--r--
                  2 root hadoop
                                    18.9 M 2025-02-28 14:57 /olist/processed/part-
     00007-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
                                    18.6 M 2025-02-28 14:57 /olist/processed/part-
     -rw-r--r--
                  2 root hadoop
     00008-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet
                  2 root hadoop
                                    18.8 M 2025-02-28 14:57 /olist/processed/part-
     -rw-r--r--
```

00009-5303d6eb-c739-4c66-9f5f-d566a672389c-c000.snappy.parquet