



RETAIL GROCERY ANALYSIS

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RETAIL GROCERY ANALYSIS

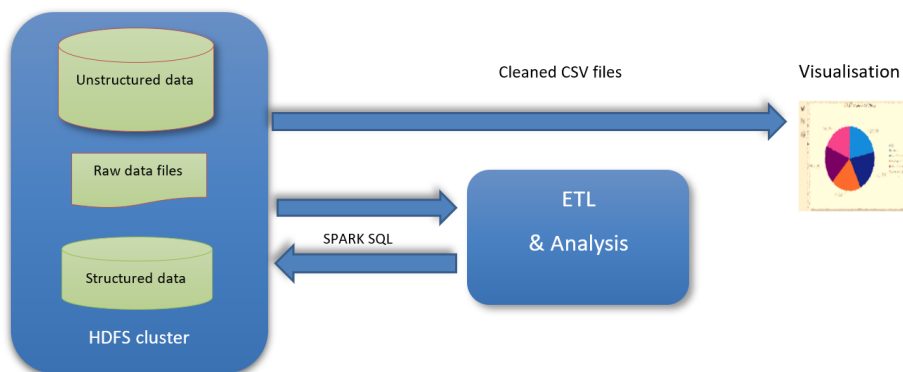
Objective:

- To perform analysis on a retail grocery company.
- To extract and build ETL pipeline and produce aggregated data into data lake.
- To analyse and provide business insights on improving sales volume and customer retention.

Solution:

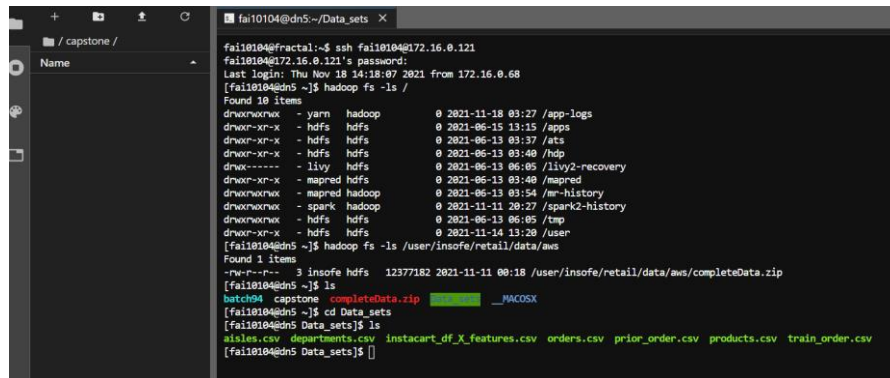
- Extract unstructured raw data from HDFS located in: /user/insofe/retail/data/aws
- Use SparkSQL to read CSV file by specifying path location.
- Create schema and dataframes for each schema.
- Check for null values and redundancies.
- Merge dataframes and perform aggregations. One final dataframe for cleaned data.
- Remove duplicate columns
- Write dataframes into CSV file and load it back on HDFS. ETL pipelining is now achieved.
- Load the output CSV file into local machine and import to Power BI for visualisation.
- Remove duplicate rows and columns, null values, perform necessary transformation for analysis.
- Analyse for product-order insights by creating visuals.

Solution Architecture:



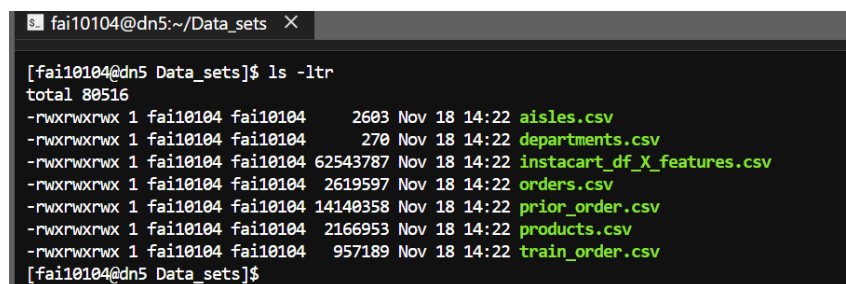
Steps:

1. Open Terminal in JupyterLab
2. Connect to Hadoop by using ssh connection:
 - a. Provide IP address and password.
 - b. Connect to HDFS by using `hadoop fs -ls /path location`



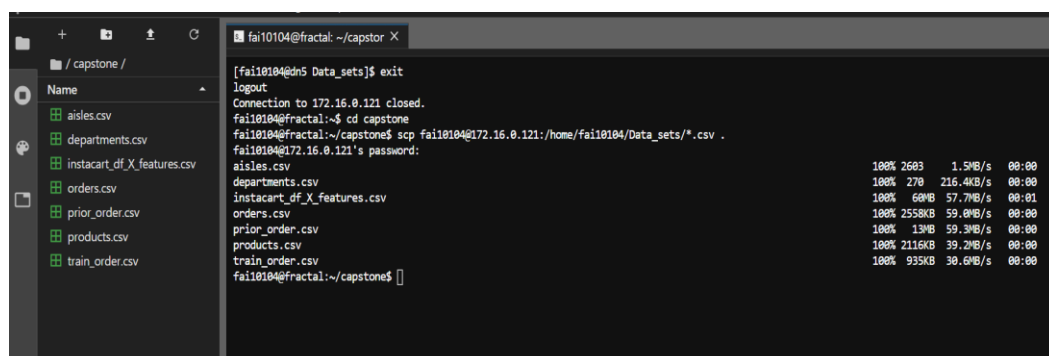
```
fai10104@fractal:~$ ssh fai10104@172.16.0.121
fai10104@172.16.0.121's password:
Last login: Thu Nov 18 14:18:07 2021 from 172.16.0.68
[fai10104@dn5 ~]$ hadoop fs -ls /
Found 10 items
drwxr-xr-x   - yarn      hadoop          0 2021-11-18 03:27 /app-logs
drwxr-xr-x   - hdfs      hdfs          0 2021-06-15 13:15 /apps
drwxr-xr-x   - hdfs      hdfs          0 2021-06-13 03:37 /ats
drwxr-xr-x   - hdfs      hdfs          0 2021-06-13 03:40 /hdp
drwxr-xr-x   - livy      hdfs          0 2021-06-13 06:05 /livy2-recovery
drwxr-xr-x   - mapred    hdfs          0 2021-06-13 03:40 /mapred
drwxr-xr-x   - mapred    hadoop        0 2021-06-13 03:54 /mr-history
drwxr-xr-x   - spark     hadoop        0 2021-11-11 20:27 /spark2-history
drwxr-xr-x   - hdfs      hdfs          0 2021-06-13 06:05 /tmp
drwxr-xr-x   - hdfs      hdfs          0 2021-11-14 13:20 /user
[fai10104@dn5 ~]$ hadoop fs -ls /user/insofe/retail/data/aws
Found 1 items
-rw-r--r--   3 insofe  hdfs    12377182 2021-11-11 00:18 /user/insofe/retail/data/aws/completeData.zip
[fai10104@dn5 ~]$ ls
batch04  capstone  completeData.zip  _MACOSX
[fai10104@dn5 ~]$ cd Data_sets
[fai10104@dn5 Data_sets]$ ls
aisles.csv  departments.csv  instacart_df_X_features.csv  orders.csv  prior_order.csv  products.csv  train_order.csv
[fai10104@dn5 Data_sets]$
```

- c. Use `ls -ltr` for listing all files and directories in the cluster



```
[fai10104@dn5 Data_sets]$ ls -ltr
total 80516
-rwxrwxrwx 1 fai10104 fai10104 2603 Nov 18 14:22 aisles.csv
-rwxrwxrwx 1 fai10104 fai10104 270 Nov 18 14:22 departments.csv
-rwxrwxrwx 1 fai10104 fai10104 62543787 Nov 18 14:22 instacart_df_X_features.csv
-rwxrwxrwx 1 fai10104 fai10104 2619597 Nov 18 14:22 orders.csv
-rwxrwxrwx 1 fai10104 fai10104 14140358 Nov 18 14:22 prior_order.csv
-rwxrwxrwx 1 fai10104 fai10104 2166953 Nov 18 14:22 products.csv
-rwxrwxrwx 1 fai10104 fai10104 957189 Nov 18 14:22 train_order.csv
[fai10104@dn5 Data_sets]$
```

- d. Exit the connection.
- e. Use `scp cluster_location local_path .` for loading all CSV files into local machine (Optional)



```
[fai10104@dn5 Data_sets]$ exit
logout
Connection to 172.16.0.121 closed.
fai10104@fractal:~$ cd capstone
fai10104@fractal:~/capstone$ scp fai10104@172.16.0.121:/home/fai10104/Data_sets/*.csv .
fai10104@172.16.0.121's password:
aisles.csv          100% 2603 1.5kB/s 00:00
departments.csv     100% 270 216.4kB/s 00:01
instacart_df_X_features.csv 100% 60MB 57.7kB/s 00:01
orders.csv          100% 2558KB 59.0kB/s 00:00
prior_order.csv     100% 13MB 59.3kB/s 00:00
products.csv        100% 2116KB 39.2kB/s 00:00
train_order.csv     100% 935KB 30.6kB/s 00:00
fai10104@fractal:~/capstone$
```

- Open another terminal to start connection and use Pyspark.
(Note: PySpark is used inside the connection throughout the ETL pipeline).

```
fail10104@172.16.0.121's password:
Last login: Fri Nov 19 17:56:31 2021 from 172.16.0.68
[fail10104@dn5 ~]$ pyspark
Python 3.8.5 (default, Jun 2 2021, 00:05:57)
[GCC 4.8.5 20150623 (Red Hat 4.8.5-44)] on linux
Type "help", "copyright", "credits" or "license()" for more information.
WARNING: An illegal reflective access operation has occurred
WARNING: Illegal reflective access by org.apache.spark.unsafe.Platform (file:/usr/share/spark-3.1.1-bin-hadoop3.2/jar
o DirectByteBuffer(long,int)
WARNING: Please consider reporting this to the maintainers of org.apache.spark.unsafe.Platform
WARNING: Use --illegal-access=warn to enable warnings of further illegal reflective access operations
WARNING: All illegal access operations will be denied in a future release
21/11/19 18:55:45 WARN NativeCodeLoader: Unable to load native-hadoop library for your platform... using builtin-jav
Using Spark's default log4j profile: org/apache/spark/log4j-defaults.properties
Setting default log level to "WARN".
To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4040. Attempting port 4041.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4041. Attempting port 4042.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4042. Attempting port 4043.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4043. Attempting port 4044.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4044. Attempting port 4045.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4045. Attempting port 4046.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4046. Attempting port 4047.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4047. Attempting port 4048.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4048. Attempting port 4049.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4049. Attempting port 4050.
21/11/19 18:55:46 WARN Utils: Service 'SparkUI' could not bind on port 4050. Attempting port 4051.
Welcome to
      ____
     / ___/
    / __/   version 3.1.1
   /___/

Using Python version 3.8.5 (default, Jun 2 2021 00:05:57)
Spark context Web UI available at http://dn5.insofe.edu.in:4051
Spark context available as 'sc' (master = local[*], app id = local-1637328346729).
SparkSession available as 'spark'.
>>>
```

- Start a spark session by importing necessary packages.
- Use `spark.read.csv` for reading the loaded CSV file:

```
fail10104@fail10104:~/capstone X fail10104@dn5:~ X fail10104@dn5:~ X
Name
aisles.csv
departments.csv
instacart_df_X_features.csv
orders.csv
prior_order.csv
products.csv
train_order.csv

>>> from pyspark.sql import SparkSession
>>> from pyspark.sql import functions as f
>>> from pyspark.sql.types import StructType, StructField, StringType, IntegerType, BooleanType
>>> spark = SparkSession.builder.appName('pyspark - example read csv').getOrCreate()
>>> sc = spark.sparkContext
>>> df = spark.read.csv("/home/fail10104/Data_sets/aisles.csv")
>>> print(df)
DataFrame[_c0: string, _c1: string]
>>> df.show()
+-----+-----+
|_c0|_c1|
+-----+-----+
|aisle_id|aisle|
|1|prepared soups sa...|
|2|specialty cheeses|
|3|energy granola bars|
|4|instant foods|
|5|marinades meat pr...|
|6|other|
|7|packaged meat|
|8|bakery desserts|
|9|pasta sauce|
|10|kitchen supplies|
|11|cold flu allergy|
|12|fresh pasta|
|13|prepared meals|
|14|tofu meat alterna...|
|15|packaged seafood|
|16|fresh herbs|
|17|baking ingredients|
|18|bulk dried fruits...|
```

Fig.1: `df.show()` is used to display data in aisles.csv

6. Schema Creation:

- `StructType()` is used to create schema by defining fields to allow nullable(True) or not (False)
- To convert schema to dataframe, use

```
df=spark.read.csv("path",header=True,schema=schema)
```

This dataframe df reads file located in "path", defines schema as mentioned in "schema" variable above and keeps the first row as the header.

a. aisle.csv

```

>>> schema_ais = StructType()
>>> .add("aisle_id", IntegerType(), False).add("aisle", StringType(), True)...
>>> df_ais = spark.read.csv("/home/fai10104/Data_sets/aisles.csv", header=True, schema=schema_ais)
>>> df_ais.printSchema()
root
 |-- aisle_id: integer (nullable = true)
 |-- aisle: string (nullable = true)

>>> df_ais.show()
+-----+-----+
|aisle_id|aisle|
+-----+-----+
|1|prepared soups sa...|
|2|specialty cheeses|
|3|energy granola bars|
|4|instant foods|
|5|marinades meat pr...|
|6|other|
|7|packaged meat|
|8|bakery desserts|
|9|pasta sauce|
|10|kitchen supplies|
|11|cold flu allergy|
|12|fresh pasta|
|13|prepared meals|
|14|tofu meat alterna...|
|15|packaged seafood|
|16|fresh herbs|
|17|baking ingredients|
|18|bulk dried fruits...|
|19|oils vinegars|

```

Fig.2: `schema_ais` = schema for aisle; `df_ais` = dataframe for aisle schema.

b. department.csv

```

>>> schema_dept = StructType()
>>> .add("department_id", IntegerType(), False).add("department", StringType(), True)...
>>> df_dept = spark.read.csv("/home/fai10104/Data_sets/departments.csv", header=True, schema=schema_dept)
>>> df_dept.printSchema()
root
 |-- department_id: integer (nullable = true)
 |-- department: string (nullable = true)

>>> df_dept.show()
+-----+-----+
|department_id|department|
+-----+-----+
|1|frozen|
|2|other|
|3|bakery|
|4|produce|
|5|alcohol|
|6|international|
|7|beverages|
|8|pets|
|9|dry goods pasta|
|10|bulk|
|11|personal care|
|12|meat seafood|
|13|pantry|
|14|breakfast|
|15|canned goods|
|16|dairy eggs|
|17|household|
|18|babies|
|19|snacks|

```

Fig.3: `schema_dept` = schema for dept data; `df_dept` = dataframe for dept schema.

c. order.csv

```

>>> schema_orders = StructType()
>>> .add("order_id", IntegerType(), False)\
>>> .add("user_id", IntegerType(), True)\
>>> .add("eval_set", StringType(), True)\
>>> .add("order_number", IntegerType(), True)\
>>> .add("order_dow", IntegerType(), True)\
>>> .add("order_hour_of_day", IntegerType(), True)\
>>> .add("days_since_prior_order", IntegerType(), True)...
>>> df_order = spark.read.csv("/home/fai10104/Data_sets/orders.csv", header=True, schema=schema_orders)
>>> df_order.printSchema()
root
 |-- order_id: integer (nullable = true)
 |-- user_id: integer (nullable = true)
 |-- eval_set: string (nullable = true)
 |-- order_number: integer (nullable = true)
 |-- order_dow: integer (nullable = true)
 |-- order_hour_of_day: integer (nullable = true)
 |-- days_since_prior_order: integer (nullable = true)

>>> df_order.show()
+-----+-----+-----+-----+-----+-----+
|order_id|user_id|eval_set|order_number|order_dow|order_hour_of_day|days_since_prior_order|
+-----+-----+-----+-----+-----+-----+
|1363380|50|prior|1|3|9|null|
|3131183|50|prior|2|6|12|null|
|2197065|50|prior|3|1|13|null|
|3201600|50|prior|4|0|11|null|
|2756886|50|prior|5|4|14|null|
|1905727|50|prior|6|2|10|null|
|357849|50|prior|7|3|13|null|
|1673791|50|prior|8|6|14|null|

```

Fig.4: `schema_orders` = schema for orders data; `df_order` = dataframe for orders schema.

d. prior_order.csv

```

>>> schema_prior_order = StructType()\
    .add("order_id", IntegerType(), True)\
    .add("product_id", IntegerType(), True)\
    .add("add_to_cart_order", IntegerType(), True)\
    .add("reordered", IntegerType(), True)... ..
>>> df_po = spark.read.csv("/home/fai10104/Data_sets/prior_order.csv", header=True, schema=schema_prior_order)
>>> df_po.printSchema()
root
|-- order_id: integer (nullable = true)
|-- product_id: integer (nullable = true)
|-- add_to_cart_order: integer (nullable = true)
|-- reordered: integer (nullable = true)
>>> df_po.show()
+-----+-----+-----+-----+
|order_id|product_id|add_to_cart_order|reordered|
+-----+-----+-----+-----+
| 12| 30597| 1| 1|
| 12| 15221| 2| 1|
| 12| 43772| 3| 1|
| 12| 37886| 4| 1|
| 12| 37215| 5| 0|
| 12| 34335| 6| 1|
| 12| 3164| 7| 0|
| 12| 26910| 8| 0|
| 12| 38888| 9| 0|
| 12| 38850| 10| 1|
| 12| 25213| 11| 0|
| 12| 11175| 12| 0|
| 12| 23390| 13| 0|

```

Fig.5: `schema_prior_order` = schema for `prior_order` data; `df_po` = dataframe for `prior_order` schema.

e. product.csv

```

>>> schema_pdt = StructType()\
    .add("product_id", IntegerType(), False)\
    .add("product_name", StringType(), True)\
    .add("aisle_id", IntegerType(), True)\
    .add("department_id", IntegerType(), True)... ..
>>> df_pdt = spark.read.csv("/home/fai10104/Data_sets/products.csv", header=True, schema=schema_pdt)
>>> df_pdt.printSchema()
root
|-- product_id: integer (nullable = true)
|-- product_name: string (nullable = true)
|-- aisle_id: integer (nullable = true)
|-- department_id: integer (nullable = true)
>>> df_pdt.show()
+-----+-----+-----+-----+
|product_id|product_name|aisle_id|department_id|
+-----+-----+-----+-----+
| 1|Chocolate Sandwiches| 61| 19|
| 2| All-Seasons Salt| 104| 13|
| 3|Robust Golden Uns...| 94| 7|
| 4|Smart Ones Classi...| 38| 1|
| 5|Green Chile Anytl...| 5| 13|
| 6| Dry Nose Oil| 11| 11|
| 7|Pure Coconut Mate...| 98| 7|
| 8|Out Russet Potato...| 116| 1|
| 9|Light Strawberry ...| 120| 16|
| 10|Sparkling Orange ...| 115| 7|
| 11| Peach Mango Juice| 35| 7|
| 12|Chocolate Fudge L...| 119| 11|
| 13| Saline Nasal Mist| 11| 11|
| 14|Fresh Scent Dishw...| 74| 17|

```

Fig.6: `schema_pdt` = schema for product data; `df_pdt` = dataframe for product schema.

f. train_order.csv

```

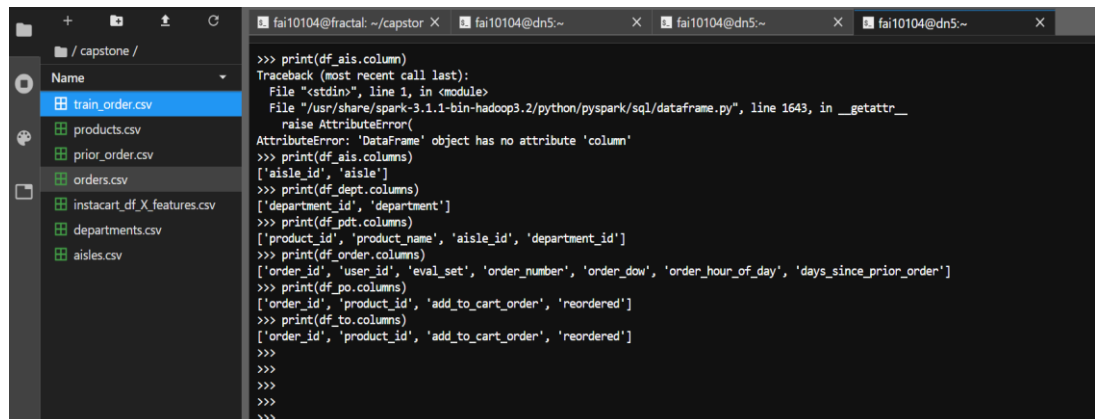
>>> schema_train_order = StructType()\
    .add("order_id", IntegerType(), True)\
    .add("product_id", IntegerType(), True)\
    .add("add_to_cart_order", IntegerType(), True)\
    .add("reordered", IntegerType(), True)... ..
>>> df_to = spark.read.csv("/home/fai10104/Data_sets/train_order.csv", header=True, schema=schema_train_order)
>>> df_to.printSchema()
root
|-- order_id: integer (nullable = true)
|-- product_id: integer (nullable = true)
|-- add_to_cart_order: integer (nullable = true)
|-- reordered: integer (nullable = true)
>>> df_to.show()
+-----+-----+-----+-----+
|order_id|product_id|add_to_cart_order|reordered|
+-----+-----+-----+-----+
| 1077| 13176| 1| 1|
| 1077| 39922| 2| 1|
| 1077| 5258| 3| 1|
| 1077| 21137| 4| 1|
| 1119| 6046| 1| 1|
| 1119| 47626| 2| 1|
| 1119| 18465| 3| 1|
| 1119| 27104| 4| 1|
| 1119| 8424| 5| 0|
| 1119| 15937| 6| 1|
| 1342| 13176| 1| 1|
| 1342| 36827| 2| 1|
| 1342| 14966| 3| 1|
| 1342| 21137| 4| 1|

```

Fig.7: `schema_train_order` = schema for train_orders data; `df_to` = dataframe for train_order schema

7. Displaying column names:

Use `print(df.columns)` to print column names as a list:

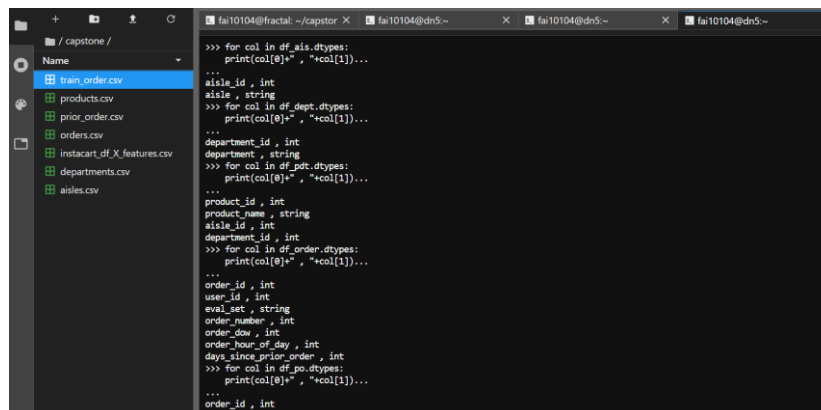


```
>>> print(df.ais.column)
Traceback (most recent call last):
  File "<stdin>", line 1, in <module>
  File "/usr/share/spark-3.1.1-bin-hadoop3.2/python/pyspark/sql/dataframe.py", line 1643, in __getattr__
    raise AttributeError(
AttributeError: 'DataFrame' object has no attribute 'column'
>>> print(df.ais.columns)
['aisle_id', 'aisle']
>>> print(df.dept.columns)
['department_id', 'department']
>>> print(df.pdt.columns)
['product_id', 'product_name', 'aisle_id', 'department_id']
>>> print(df.order.columns)
['order_id', 'user_id', 'eval_set', 'order_number', 'order_dow', 'order_hour_of_day', 'days_since_prior_order']
>>> print(df.po.columns)
['order_id', 'product_id', 'add_to_cart_order', 'reordered']
>>> print(df.to.columns)
['order_id', 'product_id', 'add_to_cart_order', 'reordered']
>>>
>>>
>>>
>>>
```

8. Display the datatypes of the columns:

Use `df.dtypes` for displaying data types of all columns in a dataframe.

For loop is used to display column and its datatypes one after the other, separated by a comma.



```
>>> for col in df.ais.dtypes:
...     print(col[0]+", "+col[1])...
...
aisle_id, int
aisle, string
>>> for col in df.dept.dtypes:
...     print(col[0]+", "+col[1])...
...
department_id, int
department, string
>>> for col in df.pdt.dtypes:
...     print(col[0]+", "+col[1])...
...
product_id, int
product_name, string
aisle_id, int
department_id, int
>>> for col in df.order.dtypes:
...     print(col[0]+", "+col[1])...
...
order_id, int
user_id, int
eval_set, string
order_number, int
order_dow, int
order_hour_of_day, int
days_since_prior_order, int
>>> for col in df.po.dtypes:
...     print(col[0]+", "+col[1])...
...
order_id, int
```

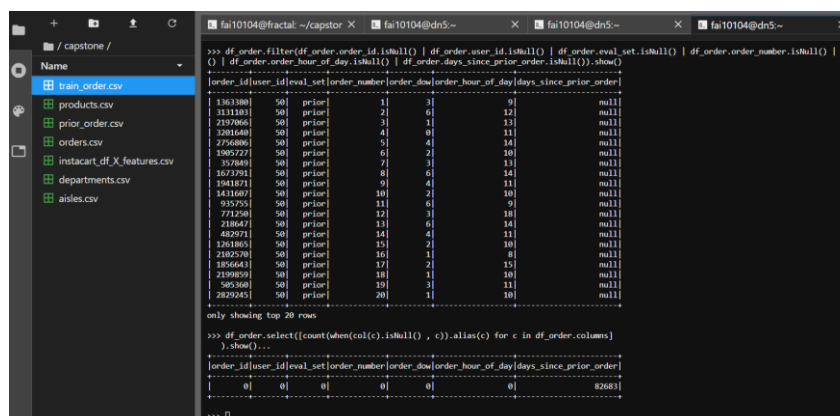
9. Check for Null values in the columns:

Use `df.column_name.isNull()` in each column of a dataframe to display all the null values present.

Use `df.filter()` to display according to the specified condition (null values, in this case)

Use `count()` to display the count of null values in each column.

a. Orders.csv (df_order)



```
>>> df_order.filter(df_order.order_id.isNull() | df_order.user_id.isNull() | df_order.eval_set.isNull() | df_order.order_number.isNull() | df_order.order_dow.isNull() | df_order.order_hour_of_day.isNull() | df_order.days_since_prior_order.isNull()).show()
[order_id|user_id|eval_set|order_number|order_dow|order_hour_of_day|days_since_prior_order]
-----
1363380    50    prior     1     3     9      null
3131180    50    prior     2     6    12     null
2197866    50    prior     3     1    13     null
1201640    50    prior     4     0    11     null
2756886    50    prior     5     4    14     null
1305727    50    prior     6     2    10     null
357849     50    prior     7     3    11     null
1673791    50    prior     8     6    14     null
1841871    50    prior     9     4    11     null
1431687    50    prior    10     2    10     null
935755     50    prior    11     6     9     null
712540     50    prior    12     3    10     null
218647     50    prior    13     6    14     null
482971     50    prior    14     4    11     null
1261865    50    prior    15     2    10     null
2182570    50    prior    16     1     8     null
1856643    50    prior    17     2    15     null
2199859    50    prior    18     1    10     null
505340     50    prior    19     3    11     null
2892453    50    prior    20     1    10     null
only showing top 20 rows

>>> df_order.select([count(when(col(c).isNull(), c)).alias(c) for c in df_order.columns]).show()
[order_id|user_id|eval_set|order_number|order_dow|order_hour_of_day|days_since_prior_order]
-----
| 0 | 0 | 0 | 0 | 0 | 0 | 82683|
```

b. Department.csv (df_dept) and aisle.csv (df_ais)

```

>>> df_dept.filter(df_dept.department_id.isnull() | df_dept.department.isnull()).show()
+-----+
|department_id|department|
+-----+
|             |          |
+-----+

>>> df_dept.filter(df_dept.department_id.isnull() | df_dept.department.isnull()).show()
+-----+
|department_id|department|
+-----+
|             |          |
+-----+

>>> df_ais.filter(df_ais.aisle_id.isnull() | df_ais.aisle.isnull()).show()
+-----+
|aisle_id|aisle|
+-----+
|         |     |
+-----+

>>> df_ais.select([count(when(col(c).isnull() , c)).alias(c) for c in df_ais.columns])
.show()
+-----+
|aisle_id|aisle|
+-----+
|         |     |
+-----+

```

c. Prior_order.csv (df_po) and train_orders.csv (df_to)

```

>>> df_po.filter(df_po.order_id.isnull() | df_po.product_id.isnull() | df_po.add_to_cart_order.isnull() | df_po.reordered.isnull()).show()>>>
+-----+
|order_id|product_id|add_to_cart_order|reordered|
+-----+
|         |          |                 |         |
+-----+

>>> df_po.select([count(when(col(c).isnull() , c)).alias(c) for c in df_po.columns])
.show()...
+-----+
|order_id|product_id|add_to_cart_order|reordered|
+-----+
|         |          |                 |         |
+-----+

>>> df_to.filter(df_to.order_id.isnull() | df_to.product_id.isnull() | df_to.add_to_cart_order.isnull() | df_to.reordered.isnull()).show()>>>
+-----+
|order_id|product_id|add_to_cart_order|reordered|
+-----+
|         |          |                 |         |
+-----+

>>> df_to.select([count(when(col(c).isnull() , c)).alias(c) for c in df_to.columns])
.show()...
+-----+
|order_id|product_id|add_to_cart_order|reordered|
+-----+
|         |          |                 |         |
+-----+

```

d. Products.csv (df_pdt)

```

>>> df_pdt.filter(df_pdt.product_id.isnull() | df_pdt.product_name.isnull() | df_pdt.aisle_id.isnull() | df_pdt.department_id.isnull()).show()
+-----+
|product_id|product_name|aisle_id|department_id|
+-----+
|6816|Scotch Kids 5" ...| null | null |
+-----+

>>> df_pdt.select([count(when(col(c).isnull() , c)).alias(c) for c in df_pdt.columns])
.show()...
+-----+
|product_id|product_name|aisle_id|department_id|
+-----+
|         |          |         |             |
+-----+

```

10. Merging of datasets:

Use `df.join()` to merge two or more datasets.

Join conditions used:

- (i) Left Join Product with aisle using aisle_id.
- (ii) Left join above joined product with department using department_id.
- (iii) Inner join above new joined product with prior_orders and train_orders with product_id.
- (iv) Left join Orders with prior_order and train_orders using order_id.

a. Product – Left Join - aisle:

```

>>> pa_lj = df_pdt.join(df_ais,["aisle_id"], "left")
>>> pa_lj.show()

```

aisle_id	product_id	product_name	department_id	aisle
61	1	Chocolate Sandwic...	19	cookies cakes
104	2	All-Seasons Salt	13	spices seasonings
94	3	Robuck Golden Uns...	7	tea
38	4	Smart Ones Classi...	1	frozen meals
5	5	Green Chile Anyti...	13	marinades meat pr...
11	6	Dry Nose Oil	11	cold flu allergy
98	7	Pure Coconut Mate...	7	juice nectars
116	8	Cut Russet Potato...	1	frozen produce
120	9	Light Strawberry ...	16	yogurt
115	10	Sparkling Orange ...	7	water seltzer spa...
31	11	Peach Mango Juice	7	refrigerated
119	12	Chocolate Fudge L...	1	frozen dessert
11	13	Saline Nasal Mist	11	cold flu allergy
74	14	Fresh Scent Dishw...	17	dish detergents
56	15	Overnight Diapers...	18	diapers wipes
103	16	Mint Chocolate Fl...	19	ice cream toppings
35	17	Rendered Duck Fat	12	poultry counter
79	18	Pizza for One Sup...	1	frozen pizza
63	19	Gluten Free Quino...	9	grains rice dried...
98	20	Pomegranate Cranb...	7	juice nectars

only showing top 20 rows

b. Product_aisle – Left Join - department:

```

>>> pad_lj = pa_lj.join(df_dept,["department_id"], "left")
>>> pad_lj.show()

```

department_id	aisle_id	product_id	product_name	aisle	department
19	61	1	Chocolate Sandwic...	cookies cakes	snacks
13	104	2	All-Seasons Salt	spices seasonings	pantry
7	94	3	Robuck Golden Uns...	tea	beverages
1	38	4	Smart Ones Classi...	frozen meals	frozen
13	5	5	Green Chile Anyti...	marinades meat pr...	pantry
11	11	6	Dry Nose Oil	cold flu allergy	personal care
7	98	7	Pure Coconut Mate...	juice nectars	beverages
1	116	8	Cut Russet Potato...	frozen produce	frozen
16	120	9	Light Strawberry ...	yogurt	dairy eggs
7	115	10	Sparkling Orange ...	water seltzer spa...	beverages
7	31	11	Peach Mango Juice	refrigerated	beverages
1	119	12	Chocolate Fudge L...	frozen dessert	frozen
11	11	13	Saline Nasal Mist	cold flu allergy	personal care
17	74	14	Fresh Scent Dishw...	dish detergents	household
18	56	15	Overnight Diapers...	diapers wipes	babies
19	103	16	Mint Chocolate Fl...	ice cream toppings	snacks
12	35	17	Rendered Duck Fat	poultry counter	meat seafood
1	79	18	Pizza for One Sup...	frozen pizza	frozen
9	63	19	Gluten Free Quino...	grains rice dried...	dry goods pasta
7	98	20	Pomegranate Cranb...	juice nectars	beverages

only showing top 20 rows

c. Prior_order – Inner Join – joined_product :

```

>>> pop_ij = df_po.join(pad_lj,["product_id"], "inner")
>>> pop_ij.show()

```

product_id	order_id	add_to_cart_order	reordered	department_id	aisle_id	product_name	aisle	department
148	17545	21	1	4	24	Nectarines fresh fruits	produce	
148	18530	5	1	4	24	Nectarines fresh fruits	produce	
148	27215	10	0	4	24	Nectarines fresh fruits	produce	
148	80551	5	1	4	24	Nectarines fresh fruits	produce	
148	112557	2	1	4	24	Nectarines fresh fruits	produce	
148	128624	8	1	4	24	Nectarines fresh fruits	produce	
148	193480	3	1	4	24	Nectarines fresh fruits	produce	
148	195519	9	0	4	24	Nectarines fresh fruits	produce	
148	197230	12	0	4	24	Nectarines fresh fruits	produce	
148	198783	4	1	4	24	Nectarines fresh fruits	produce	
148	255837	16	0	4	24	Nectarines fresh fruits	produce	
148	296420	1	0	4	24	Nectarines fresh fruits	produce	
148	308051	16	0	4	24	Nectarines fresh fruits	produce	
148	382161	3	1	4	24	Nectarines fresh fruits	produce	
148	414559	2	0	4	24	Nectarines fresh fruits	produce	
148	419410	15	0	4	24	Nectarines fresh fruits	produce	
148	431361	10	0	4	24	Nectarines fresh fruits	produce	
148	463426	13	0	4	24	Nectarines fresh fruits	produce	
148	464922	4	0	4	24	Nectarines fresh fruits	produce	
148	482428	8	1	4	24	Nectarines fresh fruits	produce	

only showing top 20 rows

d. Train_order – Inner Join – joined_product:

```

>>> top_ij = df.join(pad_lj, ["product_id"], "inner")
>>> top_ij.show()

```

product_id	order_id	add_to_cart_order	reordered	department_id	aisle_id	product_name	aisle	department
11	1290664	3	0	19	61	Chocolate Sandic...	cookies cakes	snacks
2	1455935	26	0	13	104	All-Seasons Salt	spices seasonings	pantry
3	2188722	1	1	7	94	Robust Golden Uns...	tea	beverages
10	2188721	4	0	7	115	Sparkling Orange ...	water seltzer spa...	beverages
10	1965683	13	1	7	115	Sparkling Orange ...	water seltzer spa...	beverages
12	2260187	9	1	11	119	Chocolate Fudge L...	frozen dessert	frozen
23	2152897	9	0	12	49	Organic Turkey Bu...	packaged poultry	meat seafood
25	12143783	7	1	19	3	Salted Caramel le...	energy granola bars	snacks
25	1345991	8	0	19	3	Salted Caramel le...	energy granola bars	snacks
25	378904	16	1	19	3	Salted Caramel le...	energy granola bars	snacks
25	26611	11	0	19	3	Salted Caramel le...	energy granola bars	snacks
27	1652915	3	0	11	127	Complete Spring W...	body lotions soap	personal care
28	3240767	2	1	14	121	Wheat Chex Cereal	cereal	breakfast
28	3999310	7	1	14	121	Wheat Chex Cereal	cereal	breakfast
32	3078614	3	1	19	107	Nacho Cheese Whit...	chips pretzels	snacks
32	192817	9	0	19	107	Nacho Cheese Whit...	chips pretzels	snacks
34	3395857	4	1	14	121	Peanut Butter Cereal	cereal	breakfast
34	2923280	5	0	14	121	Peanut Butter Cereal	cereal	breakfast
34	2104446	19	0	14	121	Peanut Butter Cereal	cereal	breakfast
34	1930693	33	0	14	121	Peanut Butter Cereal	cereal	breakfast

only showing top 20 rows

e. Order – Left Join – prior_orders:

```

>>> orpo_lj = df.order_join(df_po, ["order_id"], "left")
>>> orpo_lj.show()

```

order_id	user_id	eval_set	order_number	order_dow	order_hour_of_day	days_since_prior_order	product_id	add_to_cart_order	reordered
1342	156818	train	32	3	8	null	null	null	null
2142	2086	prior	5	6	16	null	26209	1	0
2142	2086	prior	5	6	16	null	35921	2	1
2142	2086	prior	5	6	16	null	28842	3	1
2142	2086	prior	5	6	16	null	28883	4	1
2142	2086	prior	5	6	16	null	14553	5	1
2142	2086	prior	5	6	16	null	25146	6	1
2142	2086	prior	5	6	16	null	7644	7	1
2142	2086	prior	5	6	16	null	42936	8	1
2142	2086	prior	5	6	16	null	22823	9	0
2142	2086	prior	5	6	16	null	41290	10	1
2142	2086	prior	5	6	16	null	2564	11	1
2142	2086	prior	5	6	16	null	17617	12	0
2142	2086	prior	5	6	16	null	16254	13	0
2659	28378	prior	3	6	7	null	31720	1	1
2659	28378	prior	3	6	7	null	33000	2	1
2659	28378	prior	3	6	7	null	27744	3	1
2659	28378	prior	3	6	7	null	24064	4	0
2659	28378	prior	3	6	7	null	21367	5	1
2659	28378	prior	3	6	7	null	11422	6	0

only showing top 20 rows

```

>>> po_merged = orpo_lj.filter(orpo_lj.eval_set == "prior")
>>> po_merged.show(5)

```

order_id	user_id	eval_set	order_number	order_dow	order_hour_of_day	days_since_prior_order	product_id	add_to_cart_order	reordered
2142	2086	prior	5	6	16	null	26209	1	0
2142	2086	prior	5	6	16	null	35921	2	1
2142	2086	prior	5	6	16	null	28842	3	1
2142	2086	prior	5	6	16	null	28883	4	1
2142	2086	prior	5	6	16	null	14553	5	1

only showing top 5 rows

```

>>> po_merged.count()
793924
>>>

```

f. Order – Left Join – Train_orders:

```

>>> orto_lj = df.order_join(df_to, ["order_id"], "left")
>>> orto_lj.show()

```

order_id	user_id	eval_set	order_number	order_dow	order_hour_of_day	days_since_prior_order	product_id	add_to_cart_order	reordered
1333380	50	prior	1	3	9	null	null	null	null
3131183	50	prior	2	6	12	null	null	null	null
2197066	50	prior	3	1	13	null	null	null	null
3201640	50	prior	4	0	11	null	null	null	null
2726086	50	prior	5	4	14	null	null	null	null
1985727	50	prior	6	2	10	null	null	null	null
357849	50	prior	7	3	13	null	null	null	null
1623791	50	prior	8	6	14	null	null	null	null
1941871	50	prior	9	4	11	null	null	null	null
1431607	50	prior	10	2	10	null	null	null	null
925755	50	prior	11	6	9	null	null	null	null
772540	50	prior	12	3	18	null	null	null	null
218647	50	prior	13	6	14	null	null	null	null
482971	50	prior	14	4	11	null	null	null	null
1261865	50	prior	15	2	10	null	null	null	null
2182570	50	prior	16	1	8	null	null	null	null
1856643	50	prior	17	2	15	null	null	null	null
2190899	50	prior	18	1	10	null	null	null	null
505360	50	prior	19	3	11	null	null	null	null
2829245	50	prior	20	1	10	null	null	null	null

only showing top 20 rows

```

>>> to_merged = orto_lj.filter(orto_lj.eval_set == "train")
>>> to_merged.show(5)

```

order_id	user_id	eval_set	order_number	order_dow	order_hour_of_day	days_since_prior_order	product_id	add_to_cart_order	reordered
1750084	50	train	68	3	9	null	31720	6	1
1750084	50	train	68	3	9	null	21903	5	1
1750084	50	train	68	3	9	null	16249	4	1
1750084	50	train	68	3	9	null	13176	3	1
1750084	50	train	68	3	9	null	47209	2	1

only showing top 5 rows

```

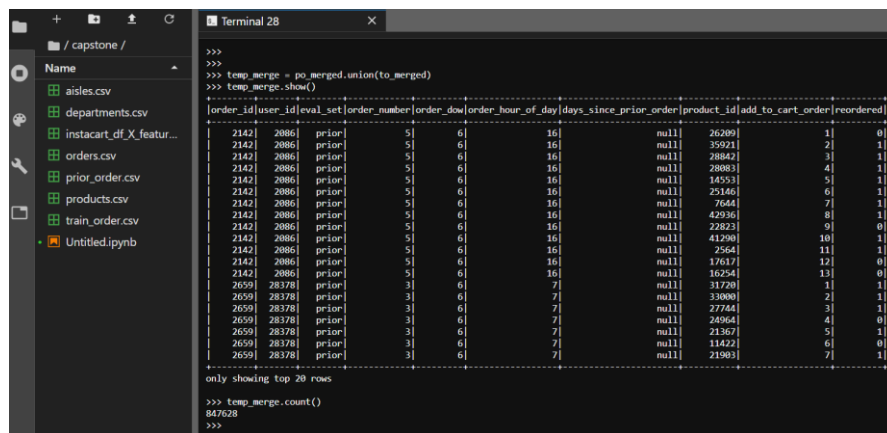
>>> to_merged.count()
53784
>>>

```

11. Aggregation of datasets:

Use `df1.union(df2)` to combine two or more datasets.

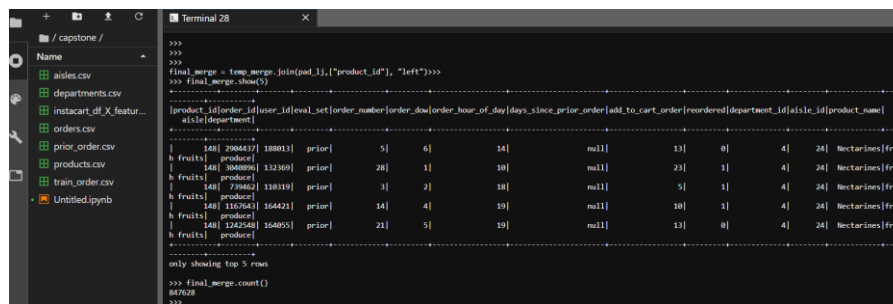
- Combining prior_orders and train_orders.



```
>>>
>>> temp_merge = po_merged.union(to_merged)
>>> temp_merge.show()
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
[order_id|user_id|eval_set|order_number|order_dow|order_hour_of_day|days_since_prior_order|product_id|add_to_cart_order|reordered]
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
[2142|2086|prior|5|6|16|null|26209|1|0]
[2142|2086|prior|5|6|16|null|35921|2|1]
[2142|2086|prior|5|6|16|null|28842|3|1]
[2142|2086|prior|5|6|16|null|28883|4|1]
[2142|2086|prior|5|6|16|null|14553|5|1]
[2142|2086|prior|5|6|16|null|25146|6|1]
[2142|2086|prior|5|6|16|null|7644|7|1]
[2142|2086|prior|5|6|16|null|42936|8|1]
[2142|2086|prior|5|6|16|null|22823|9|0]
[2142|2086|prior|5|6|16|null|41298|10|1]
[2142|2086|prior|5|6|16|null|2564|11|1]
[2142|2086|prior|5|6|16|null|17617|12|0]
[2142|2086|prior|5|6|16|null|16254|13|0]
[2659|28378|prior|3|6|7|null|31720|1|1]
[2659|28378|prior|3|6|7|null|33000|2|1]
[2659|28378|prior|3|6|7|null|27744|3|1]
[2659|28378|prior|3|6|7|null|24964|4|0]
[2659|28378|prior|3|6|7|null|21367|5|1]
[2659|28378|prior|3|6|7|null|11422|6|0]
[2659|28378|prior|3|6|7|null|21903|7|1]
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows

>>> temp_merge.count()
847628
>>>
```

- Combine all the datasets into one dataframe.



```
>>>
>>> final_merge = temp_merge.join(pod_ll, "product_id", "left")
>>> final_merge.show(5)
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
[product_id|order_id|user_id|eval_set|order_number|order_dow|order_hour_of_day|days_since_prior_order|add_to_cart_order|reordered|department_id|aisle_id|product_name]
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
[148|2904437|188013|prior|5|6|16|null|13|0|4|24|Hectarines|fres]
[h fruits|produce|112369|prior|28|1|10|null|23|1|4|24|Hectarines|fres]
[148|304806|112369|prior|28|1|10|null|23|1|4|24|Hectarines|fres]
[h fruits|produce|118119|prior|3|2|10|null|5|1|4|24|Hectarines|fres]
[148|79862|118119|prior|3|2|10|null|5|1|4|24|Hectarines|fres]
[h fruits|produce|14421|prior|14|4|19|null|10|1|4|24|Hectarines|fres]
[h fruits|produce|164055|prior|21|5|19|null|13|0|4|24|Hectarines|fres]
[h fruits|produce|1424254|prior|21|5|19|null|13|0|4|24|Hectarines|fres]
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 5 rows

>>> final_merge.count()
847628
>>>
```

12. Loading CSV file into cluster and local machine:

```
fai10104@fractal:~$ ssh fai10104@172.16.0.121
fai10104@172.16.0.121's password:
Last login: Fri Nov 19 21:39:30 2021 from 172.16.0.68
[fai10104@dn5 ~]$ scp fai10104@172.16.0.121:/home/fai10104/project/Final_output/*.csv .
fai10104@172.16.0.121's password:
part-00000-0c2c41b1-3218-44b0-8f34-e48c70927a50-c000.csv                                100% 78MB 87.3MB/s 00:00
[fai10104@dn5 ~]$
[fai10104@dn5 ~]$
[fai10104@dn5 ~]$

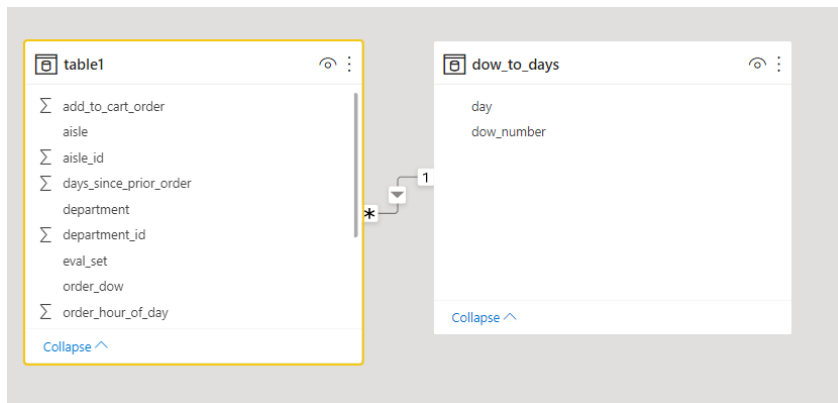
fai10104@fractal:~$ scp fai10104@172.16.0.121:/home/fai10104/project/Final_output/*.csv .
fai10104@172.16.0.121's password:
part-00000-0c2c41b1-3218-44b0-8f34-e48c70927a50-c000.csv                                100% 78MB 37.0MB/s 00:02
fai10104@fractal:~$
```

13. Visualisation:

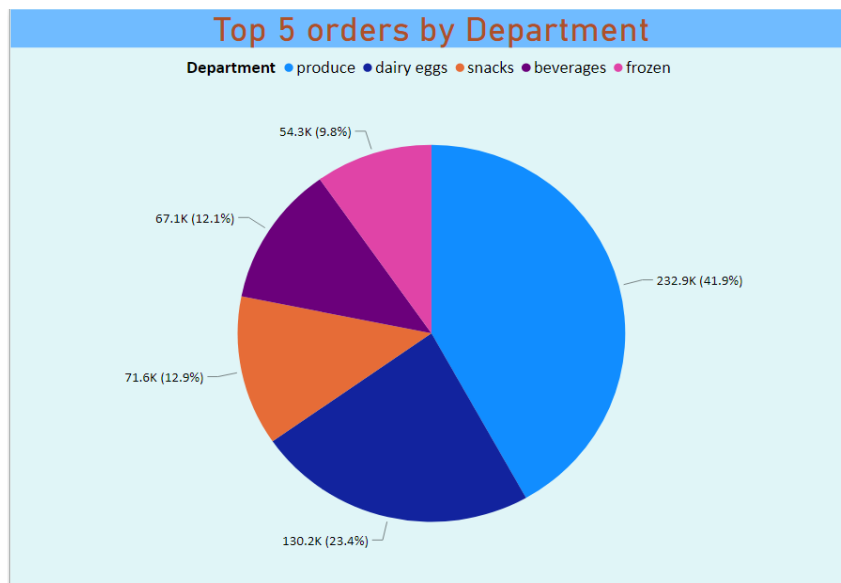
Transformation made before loading:

- Removed duplicated rows and columns
- Removed empty/null values
- Created new table and relationship with the other table.

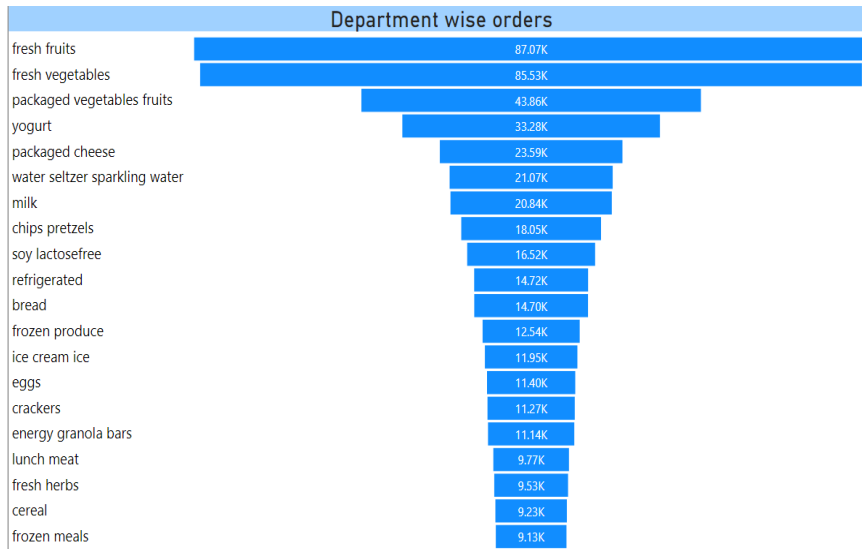
dow_number	day
0	Sunday
1	Monday
2	Tuesday
3	Wednesday
4	Thursday
5	Friday
6	Saturday



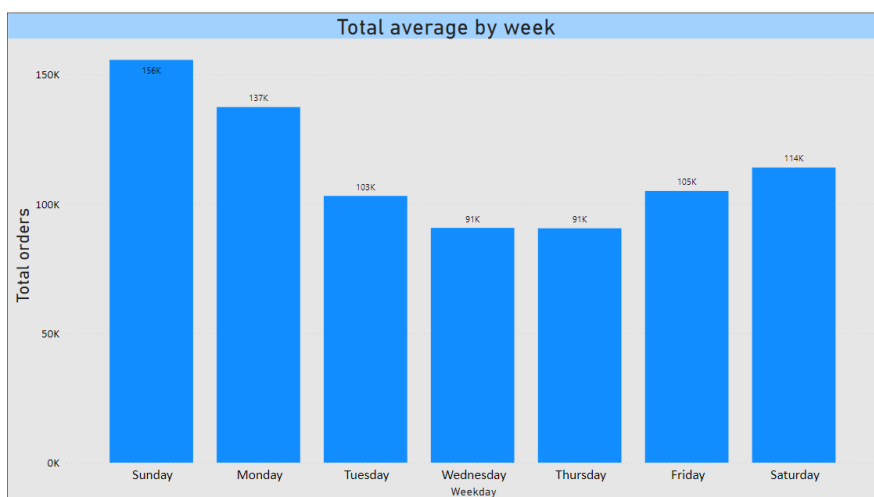
d. Created visuals:



Insight: 'Produce' department accounted 41.9% of top 5 orders.



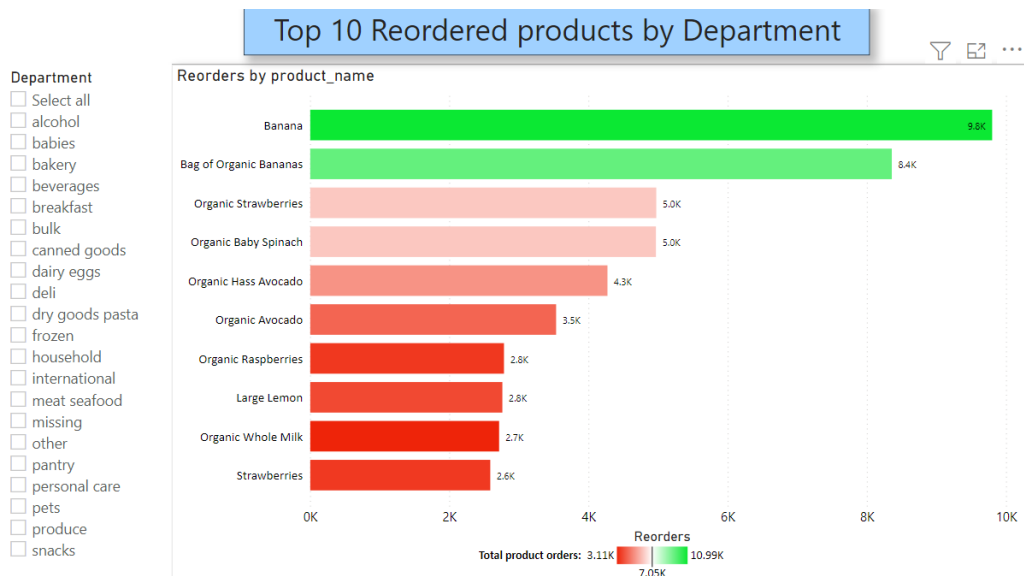
Insight: Fresh Fruits accounted around 87 thousand orders and frozen meals having lower sales. Improving cereals, frozen meals sales can drive business orders.



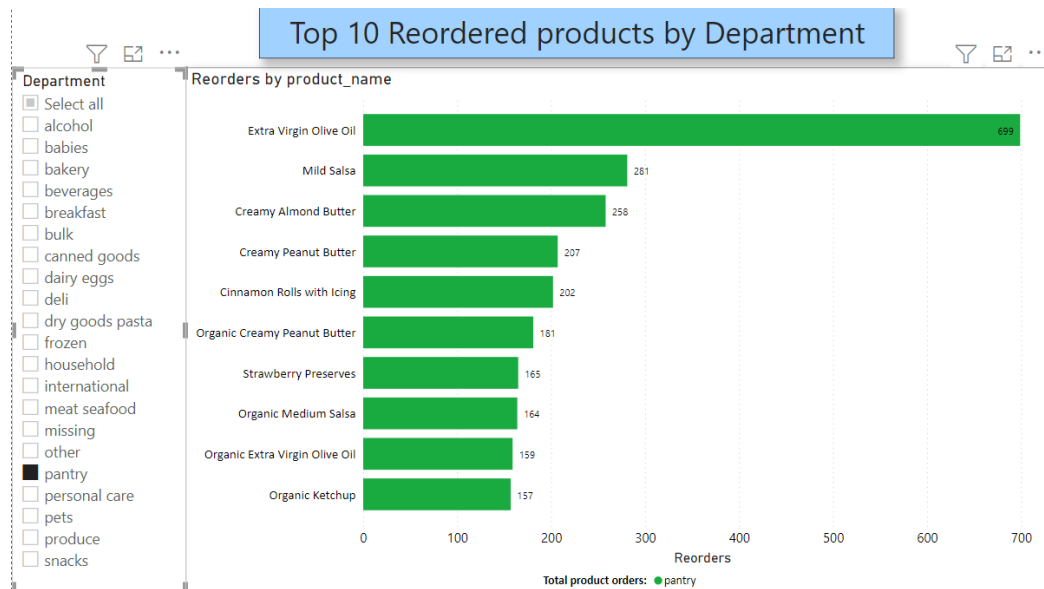
Insight: The highest orders are made around 11AM in an average week. 8AM-6PM is peak hours on average. Sunday has the highest sales while Wednesday has the lowest. Improving sales on middle of weekdays can increase business growth.



Insight: Customer_number = 164055 contributes the most accounting 3060 orders. Offering sales discounts, freebies for other customers can increase sales orders.



Insight: Banana has the highest reorders accounting 9800 orders.



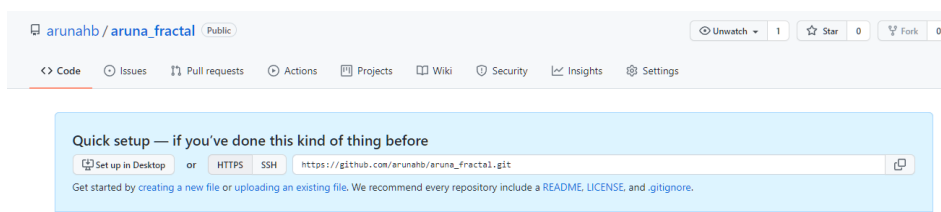
Insight: Selecting by department slicer, pantry for example provides corresponding sales orders.

Conclusion:

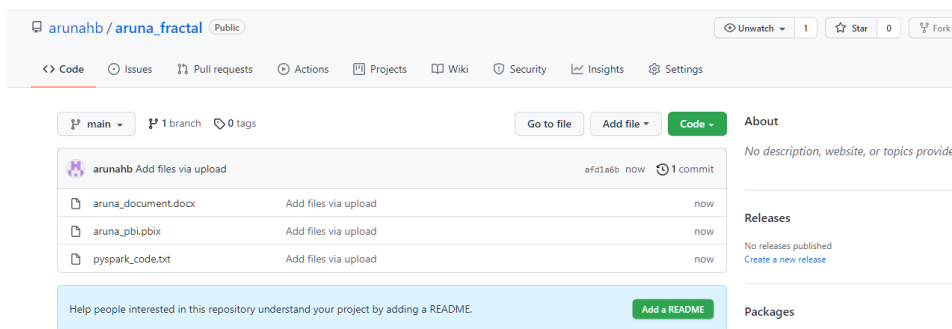
The sales orders can be increased by focusing on beverages and frozen department. Irregular customers can avail more discounts and offers than regular customers, on less peak hours on least sales day of the week (Wednesday). This can maintain business growth throughout.

Github repository:

a. Repository creation:



b. Upload supporting files:



Github_link: https://github.com/arunahb/aruna_fractal

Files submitted:

aruna_document, aruna_pbi, pyspark_code, aruna_ouput.csv