# Sales Data-Analysis

#### **Data Verification**

- Three distinct data sources—Order (CSV), Shipping (JSON), and Customers (Excel)—have been provided for analysis.
- Data normalization and verification were conducted using Python to ensure consistency across formats thus enabling streamlined and accurate analytical processing.
- Below Python commands were utilized to structure and prepare it for further business intelligence activities.

```
>>> import pandas as pd
cust = pd.read_excel("C:/Users/anm/Downloads/Customer.xls",engine='xlrd')
>>> cust.to_csv("C:/Users/anm/Downloads/Customer.csv", index=False)
>>> order=pd.read_csv("C:/Users/anm/Downloads/Order.csv")
>>> Shipping=pd.read_json("C:/Users/anm/Downloads/Shipping.json")
>>> Shipping.to_csv("C:/Users/anm/Downloads/Shipping.csv")
```

Next, we check for Missing Values in all 3 data sources

```
print(order.isnull().sum())
              0
Order ID
Item
               0
               0
Amount
Customer ID
dtype: int64
print(Shipping.isnull().sum())
             0
Shipping_ID
Status
               0
Customer ID
dtype: int64
print(cust.isnull().sum())
Customer ID
               0
First
Last
               0
               0
Age
Country
               0
dtype: int64
```

 Duplicate records were proactively identified and removed, where present, to eliminate data redundancy and enhance the overall quality

```
print("Orders:", order.duplicated().sum())
Orders: 0
print("Shipping", Shipping.duplicated().sum())
Shipping 0
print("Customers:", cust.duplicated().sum())
Customers: 0
```

Data types for each column are verified

Each data source should have a Primary Key column which is Non-Null and Unique, next step is to ensure
this constraint is met in all 3 data sources. We compare the number of Unique and non-null values with the
complete row count in each.

```
print("Unique Values in Orders:\n",order.nunique().sort_values(ascending=False))
Unique Values in Orders:
Customer_ID 160
Amount
Item
print("Orders Total Row Count:", len(order))
Orders Total Row Count: 250
print("Unique Values in Shipping:\n", Shipping.nunique().sort_values(ascending=False))
Unique Values in Shipping:
Customer ID
Status
dtype: int64
print("Shipping Total Row Count:", len(Shipping))
Shipping Total Row Count: 250
print("Unique Values in Customers:\n",cust.nunique().sort_values(ascending=False))
Unique Values in Customers:
 Customer_ID
              189
Last
Age
Country
print("Customers Total Row Count:", len(cust))
Customers Total Row Count: 250
```

- Hence, by observation Order\_ID from Orders, Shipping\_ID from Shipping and Customer\_ID from Customers
  data can be used as Primary Keys.
- Next, we check the referential Integrity Constraint, to ensure there is no Customer ID in Orders whose details
  are not present in Customers data and similarly in Shipping data

```
print("Invalid Customers:", Shipping[~Shipping['Customer_ID'].isin(cust['Customer_ID'])])
Invalid Customers: Empty DataFrame
Columns: [Shipping_ID, Status, Customer_ID]
Index: []
print("Invalid Customers:", order[~order['Customer_ID'].isin(cust['Customer_ID'])])
Invalid Customers: Empty DataFrame
Columns: [Order_ID, Item, Amount, Customer_ID]
Index: []
```

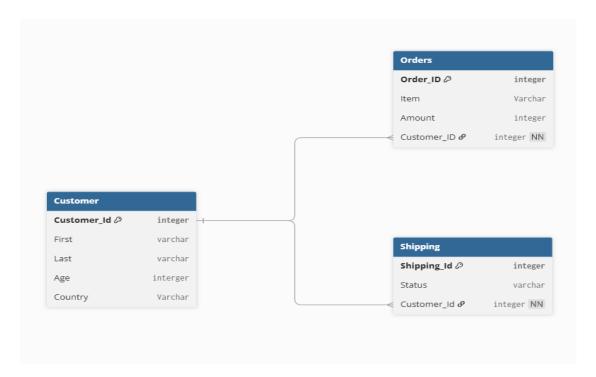
- Above output indicates that Customers table has information about all customers who have placed an order or whose order has been shipped
- Upon validation, it was observed that certain customers listed in the Orders dataset do not have corresponding Shipping details, and vice versa, indicating incomplete data linkage across sources.

```
print("Invalid Customers:",Shipping[~Shipping['Customer_ID'].isin(order['Customer_ID'])]
 Invalid Customers:
                           Shipping_ID
                                             Status Customer ID
                       Pending
                                          155
72
                       Pending
                  5 Delivered
                       Pending
 12
                13 Delivered
                                          141
               239
                      Pending
 240
243
               241
244
                       Pending
Pending
 244
249
                    Delivered
               250 Delivered
 [98 rows x 3 columns]
print("Invalid Customers:",order[~order['Customer ID'].isin(Shipping['Customer ID'])])
                                                  Amount Customer ID
                  Keyboard
                                               139
                     Mouse
                                              250
239
                  Monitor
              4 Keyboard
5 Mousepad
                                              153
 230
            231 Keyboard
                                400
 232
            233
                             12000
            238 Mousepad
                                200
            243 Monitor
249 DDR RAM
 248
                               1500
 [94 rows x 4 columns]
```

Conclusions: The 3 datasets have non-null values with no duplicates and **Order\_ID** from Orders, **Shipping\_ID** from Shipping and **Customer\_ID** from Customers data can be used as **Primary Keys**. With **Customer\_ID** being the foreign key in Orders and Shipping data. The datatype of each column is known and also we noticed that Orders and Shipping data cannot be a complete join.

# **ER Diagram**

The below represents an Entity-Relation Diagram which serves as a reference for the Data Engineer to build the tables.



### Customers:

- The Customer ID is the primary key.
- A customer can have multiple orders and shipping records, but each record in Orders and Shipping refers to exactly one customer (1:N relationship).

#### Orders:

- The Order ID is the primary key.
- Each order is associated with a Customer\_ID (foreign key from Customers).
- A customer can place multiple orders, but an order is associated with exactly one customer.

#### Shipping:

- The Shipping\_ID is the primary key.
- Each shipping record is associated with a Customer\_ID (foreign key from Customers).
- A customer can have multiple shipping records, but a shipping record is associated with exactly one customer.
- As depicted in the diagram, the Customer\_ID fields in both the Orders and Shipping dataset's function acts
  as foreign key columns referencing the Customers dataset. These fields are expected to be non-null, to maintain
  referential integrity.
- Additionally, Quality Assurance Engineers should ensure that these columns adhere to the defined data types during testing to uphold data consistency and schema compliance.

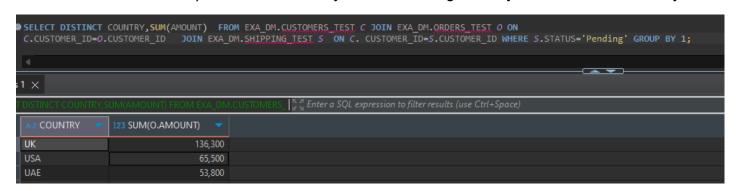
## Part-2

# **Business Reporting Queries**

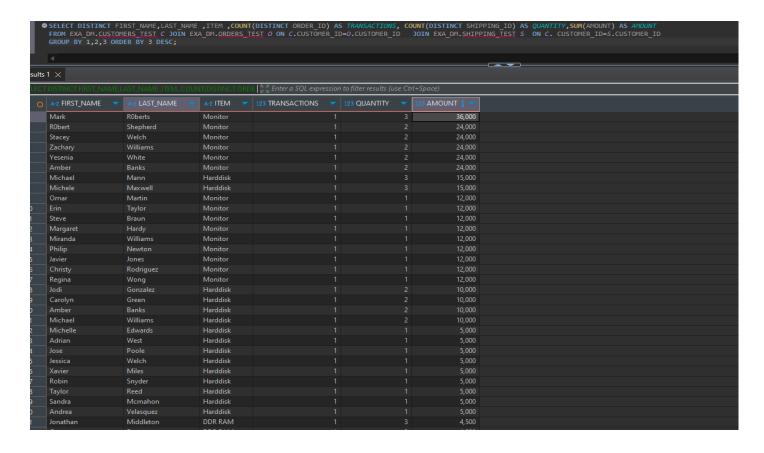
To execute the Queries 3 tables : CUSTOMERS\_TEST, ORDERS\_TEST, SHIPPING\_TEST is created as below using SQL

```
CREATE OR REPLACE TABLE EXA DM.CUSTOMERS TEST (
 CUSTOMER ID VARCHAR(30) PRIMARY KEY,
 FIRST NAME VARCHAR(30),
 LAST NAME VARCHAR(30),
 AGE INTEGER,
 COUNTRY VARCHAR(100)
IMPORT INTO EXA DM.CUSTOMERS TEST
 FROM LOCAL CSV FILE 'C:/Users/anm/Downloads/Customer.csv' skip=1;
CREATE OR REPLACE TABLE EXA DM.ORDERS TEST (
 ORDER ID INTEGER PRIMARY KEY,
 ITEM VARCHAR(30),
 AMOUNT INTEGER,
 CUSTOMER ID INTEGER);
IMPORT INTO EXA DM.ORDERS TEST
 FROM LOCAL CSV FILE 'C:/Users/anm/Downloads/Order.csv' skip=1;
CREATE OR REPLACE TABLE EXA DM.SHIPPING TEST (
 SHIPPING ID INTEGER PRIMARY KEY,
 STATUS VARCHAR(30),
 CUSTOMER_ID INTEGER);
IMPORT INTO EXA DM.SHIPPING TEST
 FROM LOCAL CSV FILE 'C:/Users/anm/Downloads/Shipping.csv'
```

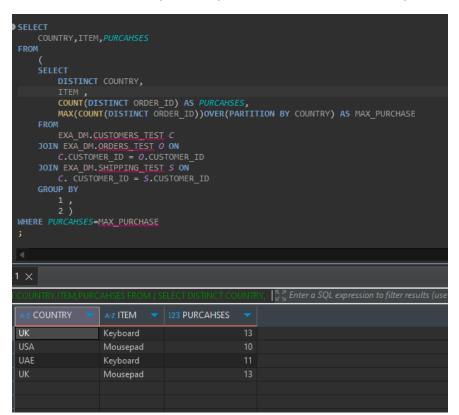
1. The total amount spent and the country for the Pending delivery status for each country.



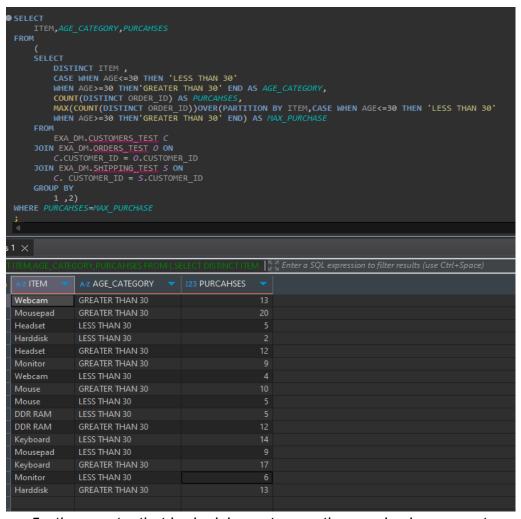
2. the total number of transactions, total quantity sold, and total amount spent for each customer, along with the product details.



3. the maximum product purchased for each country.



4. the most purchased product based on the age category less than 30 and above 30.



the country that had minimum transactions and sales amount.

