

# SQL + Python Retail Order Analysis

## Performed Extact, Transform and Load (ETL)

### 1.Extraction

Extracted the data by using kaggle API.

```
In [110]: #!pip install Kaggle
import kaggle
```

```
In [111]: !kaggle datasets download ankitbansal06/retail-orders -f orders.csv
```

```
In [112]: #extract file from zip file
import zipfile
zip_ref = zipfile.ZipFile('orders.csv.zip')
zip_ref.extractall() # extract file to dir
zip_ref.close() # close file
```

### 2. Transform

```
In [113]: #read data from the file and handle null values
import pandas as pd
df = pd.read_csv('orders.csv',na_values=['Not Available','unknown'])
df['Ship Mode'].unique()
```

```
Out[113]: array(['Second Class', 'Standard Class', nan, 'First Class', 'Same Day'],
dtype=object)
```

```
In [114]: #rename columns names and make them Lower case and replace space with underscore
df.columns = df.columns.str.lower()
df.columns = df.columns.str.replace(' ', '_')
df.columns
```

```
Out[114]: Index(['order_id', 'order_date', 'ship_mode', 'segment', 'country', 'city',
'state', 'postal_code', 'region', 'category', 'sub_category',
'product_id', 'cost_price', 'list_price', 'quantity',
'discount_percent'],
dtype='object')
```

```
In [115]: #derive new columns discount , sale price and profit
df['discount']=df['list_price']*df['discount_percent']*0.01
df['sale_price']= df['list_price']-df['discount']
df['profit']=df['sale_price']-df['cost_price']
df.head(5)
```

```
Out[115]:
```

	try	city	state	postal_code	region	category	sub_category	product_id	cost_price	list_price	quantity	discount_percent	discount	sale_price	profit
ted	tes	Henderson	Kentucky	42420	South	Furniture	Bookcases	FUR-BO-10001798	240	260	2	2	5.2	254.8	14.8
ted	tes	Henderson	Kentucky	42420	South	Furniture	Chairs	FUR-CH-10000454	600	730	3	3	21.9	708.1	108.1
ted	tes	Los Angeles	California	90036	West	Office Supplies	Labels	OFF-LA-10000240	10	10	2	5	0.5	9.5	-0.5
ted	tes	Fort Lauderdale	Florida	33311	South	Furniture	Tables	FUR-TA-10000577	780	980	5	2	19.2	940.8	160.8
ted	tes	Fort Lauderdale	Florida	33311	South	Office Supplies	Storage	OFF-ST-10000760	20	20	2	5	1.0	19.0	-1.0

```
In [116]: #convert order date from object data type to datetime
df['order_date']=pd.to_datetime(df['order_date'],format="%Y-%m-%d")
```

```
In [117]: df.drop(columns=['list_price','cost_price','discount_percent'],inplace=True)
```

```
In [118]: df.head(5)
```

```
Out[118]:
```

	order_id	order_date	ship_mode	segment	country	city	state	postal_code	region	category	sub_category	product_id	quantity	discount	sal
0	1	2023-03-01	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Bookcases	FUR-BO-10001798	2	5.2	
1	2	2023-08-15	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Chairs	FUR-CH-10000454	3	21.9	
2	3	2023-01-10	Second Class	Corporate	United States	Los Angeles	California	90036	West	Office Supplies	Labels	OFF-LA-10000240	2	0.5	
3	4	2022-06-18	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Furniture	Tables	FUR-TA-10000577	5	19.2	
4	5	2022-07-13	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Office Supplies	Storage	OFF-ST-10000760	2	1.0	

## Load data in postgresSQL

```
Out[121]: Index(['order_id', 'order_date', 'ship_mode', 'segment', 'country', 'city',
                  'state', 'postal_code', 'region', 'category', 'sub_category',
                  'product_id', 'quantity', 'discount', 'sale_price', 'profit'],
                  dtype='object')
```

Query

Query History

```

1  -- Find top 10 highest revenue-generating products:
2  SELECT product_id, SUM(sale_price) AS sales
3  FROM df_orders
4  GROUP BY product_id
5  ORDER BY sales DESC
6  LIMIT 10;

```

Data Output

Messages

Notifications

≡+

📄

▼

📋

▼

🗑️

🗄️

⬇️

📈

	product_id character varying (50) 🔒	sales numeric 🔒
1	TEC-CO-10004722	59514.00
2	OFF-BI-10003527	26525.30
3	TEC-MA-10002412	21734.40
4	FUR-CH-10002024	21096.20
5	OFF-BI-10001359	19090.20
6	OFF-BI-10000545	18249.00
7	TEC-CO-10001449	18151.20
8	TEC-MA-10001127	17906.40
9	OFF-BI-10004995	17354.80
10	OFF-SU-10000151	16325.80

Query

Query History

```
8  -- Find top 5 highest selling products in each region:
9  WITH cte AS (
10     SELECT region, product_id, SUM(sale_price) AS sales
11     FROM df_orders
12     GROUP BY region, product_id)
13  SELECT *
14  FROM (SELECT *, ROW_NUMBER() OVER(PARTITION BY region ORDER BY sales DESC) AS rn
15        FROM cte) AS A
16  WHERE rn <= 5;
17
```

Data Output

Messages

Notifications

	region character varying (20)	product_id character varying (50)	sales numeric	rn bigint
1	Central	TEC-CO-10004722	16975.00	1
2	Central	TEC-MA-10000822	13770.00	2
3	Central	OFF-BI-10001120	11056.50	3
4	Central	OFF-BI-10000545	10132.70	4
5	Central	OFF-BI-10004995	8416.10	5
6	East	TEC-CO-10004722	29099.00	1
7	East	TEC-MA-10001047	13767.00	2
8	East	FUR-BO-10004834	11274.10	3
9	East	OFF-BI-10001359	8463.60	4
10	East	TEC-CO-10001449	8316.00	5
11	South	TEC-MA-10002412	21734.40	1
12	South	TEC-MA-10001127	11116.40	2

Total rows: 20 of 20

Query complete 00:00:00.085

Query

Query History

```

34  -- For each category, find the month with the highest sales:
35  WITH cte AS (
36      SELECT category,
37              TO_CHAR(order_date, 'YYYYMM') AS order_year_month,
38              SUM(sale_price) AS sales
39      FROM df_orders
40      GROUP BY category, TO_CHAR(order_date, 'YYYYMM')
41  )
42  SELECT *
43  FROM (
44      SELECT *,
45              ROW_NUMBER() OVER(PARTITION BY category ORDER BY sales DESC) AS rn
46      FROM cte
47  ) AS a
48  WHERE rn = 1;
49

```

Data Output

Messages

Notifications

+

📄

▼

📋

▼

🗑️

🗄️

⬇️

📈

	category character varying (20) 🔒	order_year_month text 🔒	sales numeric 🔒	rn bigint 🔒
1	Furniture	202210	42888.90	1
2	Office Supplies	202302	44118.50	1
3	Technology	202310	53000.10	1



Query

Query History

```

19 -- Find month-over-month growth comparison for 2022 and 2023 sales:
20 WITH cte AS (
21     SELECT EXTRACT(YEAR FROM order_date) AS order_year,
22            EXTRACT(MONTH FROM order_date) AS order_month,
23            SUM(sale_price) AS sales
24     FROM df_orders
25     GROUP BY EXTRACT(YEAR FROM order_date), EXTRACT(MONTH FROM order_date)
26 )
27 SELECT order_month,
28        SUM(CASE WHEN order_year = 2022 THEN sales ELSE 0 END) AS sales_2022,
29        SUM(CASE WHEN order_year = 2023 THEN sales ELSE 0 END) AS sales_2023
30 FROM cte
31 GROUP BY order_month
32 ORDER BY order_month;

```

Data Output

Messages

Notifications

	order_month numeric	sales_2022 numeric	sales_2023 numeric
1	1	94712.50	88632.60
2	2	90091.00	128124.20
3	3	80106.00	82512.30
4	4	95451.60	111568.60
5	5	79448.30	86447.90
6	6	94170.50	68976.50
7	7	78652.20	90563.80
8	8	104808.00	87733.60
9	9	79142.20	76658.60

Total rows: 12 of 12

Query complete 00:00:00.075

Query

Query History

```

50 -- Identify the sub-category with the highest growth in profit from 2022 to 2023:
51 WITH cte AS (
52     SELECT sub_category,
53            EXTRACT(YEAR FROM order_date) AS order_year,
54            SUM(sale_price) AS sales
55     FROM df_orders
56     GROUP BY sub_category, EXTRACT(YEAR FROM order_date)),
57     cte2 AS (
58     SELECT sub_category,
59            SUM(CASE WHEN order_year = 2022 THEN sales ELSE 0 END) AS sales_2022,
60            SUM(CASE WHEN order_year = 2023 THEN sales ELSE 0 END) AS sales_2023
61     FROM cte
62     GROUP BY sub_category)
63 SELECT *
64 FROM cte2
65 ORDER BY (sales_2023 - sales_2022) DESC
66 LIMIT 1;
67

```

Data Output

Messages

Notifications

	sub_category character varying (20)	sales_2022 numeric	sales_2023 numeric
1	Machines	73723.20	109178.50