# Retail Orders Data Analysis

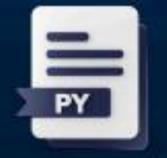












**Data Cleaning** 

and Processing



Loading Data

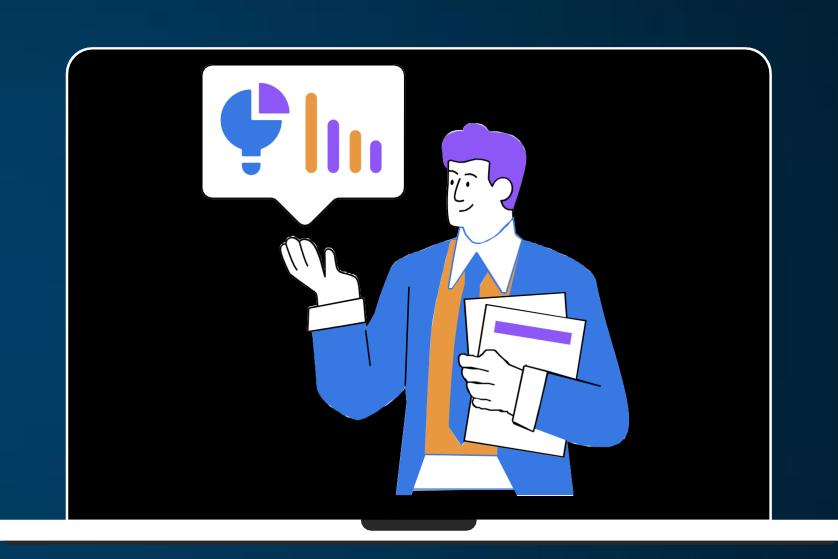


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### Introduction

This project focuses on a comprehensive analysis of sales data to uncover key performance insights. Utilizing Python for data cleaning ensured the accuracy and reliability of the results. The analysis covers revenue generation, sales performance, and profit growth, providing valuable information to guide strategic decisions.



### Objectives

- Identifying top revenue-generating products.
- Assessing regional sales performance.
- Comparing year-over-year sales growth.
- Highlighting profit growth trends.

### Scope

- Data Cleaning: Used Python for thorough data cleaning to ensure accuracy.
- Performance Analysis: Assessed revenue generation, regional sales, and profit growth trends.





## Workflow







Data Collection and Loading

Data Cleaning and Transformation

Data Storage and Management

Data Analysis and Insights

# Data Collection and Loading

#### **Data Sources**

· Orders dataset containing information about order id, order date, product id, cost price, listed price, discount etc.

### **Data Loading Process**

 Data was initially loaded into Python for preprocessing and then imported into the raw data layer of SQL.

# DATA CLEANING



# Null Handling

```
import numpy as np
df["Ship Mode"]=df["Ship
Mode"].replace({'unknown':np.nan,"Not
Available":np.nan})
```

To ensure data consistency and accuracy, we implemented null value handling by replacing occurrences of "unknown" and "Not Available" with NaN (Not a Number) in the "Ship Mode" column. This approach streamlines data preprocessing, enabling clearer insights during our analysis phase.

# Standardizing Column Names

```
df.columns=df.columns.str.lower().str.replace(" ","_")
```

To ensure uniformity and ease of access, we standardized column names in our dataset. The transformation involved converting column names to lowercase and replacing spaces with underscores for consistency.

Product Id \_\_\_\_\_ product\_id

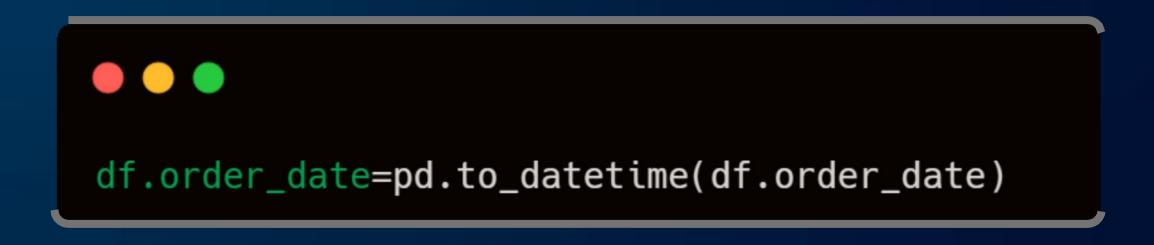
### Addition of New Columns

```
df['discount_per_unit']=df.list_price*df.discount_percent*0.01
df["sale_price"]=df.list_price-df.discount_per_unit
df["profit"]=df.sale_price - df.cost_price
```

To further enhance our dataset for comprehensive analysis, we introduced the following new columns:

- Discount per Unit: Calculated as a percentage of the list price.
- Sale Price: Derived by subtracting the discount per unit from the list price.
- Profit: Determined by subtracting the cost price from the sale price.

# Conversion of Order Date to Datetime



To enhance analysis and ensure consistency in date handling, we converted the "order\_date" column to datetime format. This transformation enables accurate date-based calculations and insights within our dataset.

# DATAANALYSIS



1.Find top 10 highest revenue generating products.

**1.Top 10** highest revenue generatin products.

```
Product ID
                    Revenue
TEC-CO-10004722
                    245,056
OFF-BI-10000545
                    163,777.7
TEC-MA-10002412
                    130,406.4
FUR-CH-10002024
                    120,090.7
TEC-PH-10001459
                    113,041.9
TEC-CO-10001449
                    107,388
OFF-BI-10003527
                    97,082.9
TEC-MA-10000822
                    89,622.3
FUR-BO-10002213
                    84,014.8
TEC-MA-10001047
                    81,549
```

2.Find top 5
highest
selling
products in
each region.

```
with cte AS
    (SELECT region ,
         product_id ,
         sum(quantity) AS quantity,
         dense_rank() over(partition by region
    ORDER BY sum(quantity) desc) AS r
    FROM orders
    GROUP BY region, product_id)
SELECT region ,
        product_id,
        quantity,
FROM cte
WHERE r<=5;
```

2.Top 5
highest
selling
products in
each region.

Ī	Region	 I	Product ID		Quantity	 	Ranking	`
ı	Central	1	OFF-BI-1000030		34	ا	1	
1	Central	1	OFF-BI-1000075	6	33	- 1	2	-
1	Central	1	OFF-BI-1000054	6	29	- 1	3	1
1	Central	1	OFF-BI-10001249	9	29	- 1	3	1
1	Central	1	FUR-CH-10002304	4	27	- 1	4	1
1	Central	1	OFF-AP-1000194	7	27	- 1	4	1
1	Central	1	FUR-CH-1000288	a	26	- 1	5	- 1
1	East	1	OFF-PA-10001970	a	33	- 1	1	1
1	East	1	OFF-BI-10003656	5	32	- 1	2	- 1
1	East	1	OFF-FA-1000062	1	31	1	3	1
1	East	1	FUR-FU-10004848	8	31	-1	3	1
1	East	1	OFF-FA-1000278	a	29	- 1	4	1
1	East	1	OFF-ST-1000261	5	29	-1	4	1
1	East	1	OFF-BI-10001524	4	28	- 1	5	- 1
1	South	1	OFF-ST-1000371	6	26	- 1	1	1
1	South	1	FUR-CH-1000051	3	24	- 1	2	- 1
1	South	1	OFF-BI-10004728	8	24	- 1	2	- 1
1	South	1	OFF-BI-10000014	4	23	- 1	3	- 1
1	South	1	FUR-FU-1000173	1	21	- 1	4	- 1
1	South	1	OFF-BI-10000069	9	21	- 1	4	1
1	South	1	OFF-BI-1000097	7	20	- 1	5	1
1	South	1	OFF-BI-1000119	1	20	- 1	5	1
1	South	1	TEC-AC-1000002	3	20	- 1	5	1
1	South	1	TEC-PH-10001459	9	20	- 1	5	- 1
1	South	1	FUR-TA-10000198	B	20	Ì	5	1
1	West	1	TEC-AC-1000383	2	45	1	1	1
1	West	1	OFF-BI-10000174	4	32	1	2	1
1	West	1	OFF-BI-1000103	5	31	1	3	1
1	West	1	OFF-BI-10001670	a	29		4	1
1	West	1	OFF-ST-1000248	6	29	1	4	1
1	West	1	FUR-FU-10001979	9	28	1	5	1
1	West	1	TEC-AC-10002000	5	28	1	5	1

3.Find month over month growth comparison for 2022 and 2023 sales.

```
SELECT month(order_date) AS month,
         round(sum(case
   WHEN year(order_date)=2022 THEN
    sale_price*quantity
    ELSE 0 end),2) AS '2022', round(sum(case
   WHEN year(order_date)=2023 THEN
    sale_price
    ELSE 0 end),2) AS '2023'
FROM orders
GROUP BY month(order_date)
ORDER BY month(order_date);
```

3.Find month over month growth comparison for 2022 and 20 23 sales.

П	Month	ı	2022 2023
Ш	1		437,431.3   88,632.6
П	2		444,011.1   128,124.2
П	3	1	394,105.2   82,512.3
П	4	1	476,400.9   111,568.6
П	5	1	413,625.5   86,447.9
П	6	1	465,300.3   68,976.5
П	7	1	375,278.4   90,563.8
П	8	1	534,562.4   87,733.6
ı	9	1	433,887   76,658.6
ı	10	1	601,707.8   121,061.5
	11	1	451,809.6   75,432.8
	12	1	447,421.8   102,556.1

4.For each category which month had highest sales.

```
with cte AS
    (SELECT category,
         CONCAT(CAST(EXTRACT(MONTH
    FROM order_date) AS CHAR), '-', CAST(EXTRACT(YEAR)
    FROM order_date) AS CHAR)) AS month,
round(sum(sale_price*quantity),2) AS sales, dense_rank()
over(partition by category
    ORDER BY sum(sale_price*quantity) desc) AS r
    FROM orders
    GROUP BY category, CONCAT(CAST(EXTRACT(MONTH)
    FROM order_date) AS CHAR), '-', CAST(EXTRACT(YEAR
    FROM order_date) AS CHAR)) )
SELECT category,
         month,
         sales
FROM cte
WHERE r=1;
```

4.For each category which month had highest sales.

5. Which sub category had highest growth by profit in 2023 compare to 2022.

```
SELECT sub_category,
         sum(case
   WHEN year(order_date)=2022 THEN
   profit
   ELSE 0 end) AS "2022", sum(case
    WHEN year(order_date)=2023 THEN
    profit*quantity
    ELSE 0 end) AS "2023", (sum(case
    WHEN year(order_date)=2023 THEN
    profit*quantity
   ELSE 0 end) - sum(case
    WHEN year(order_date)=2022 THEN
    profit*quantity
    ELSE 0 end))*100.0/sum(case
    WHEN year(order_date)=2022 THEN
    profit*quantity
   ELSE 0 end) AS growth_pct
FROM orders
GROUP BY sub_category
ORDER BY growth_pct DESC;
```

5.Sub category had highest growth by profit in 2023 compare to 2022.

ı	Sub-Catego	ory	2022	١	2023	I	Growth	(%)
ı	Supplies	ı	1,500.7		9,241.5		79.06	
	Machines	1	7,243.2	-1	56,939.8	-1	64.54	1
	Binders	1	8,685.5	-1	57,990.3	-1	42.09	1
	Envelopes	s	607.2	-1	3,502.4	-1	34.33	1
	Storage	- 1	8,907.4	-1	53,177.5	-1	23.88	1
	Phones	- 1	13,024.7	-1	76,438.2	-1	21.21	- 1
	Labels	- 1	349.6	-1	2,181.6	-1	8.69	1
	Accessories	s	7,387.2	-1	40,613.7	-1	4.68	- 1
	Bookcases	- 1	5,459.5	-1	25,776.1	-1	1.10	- 1
	Paper	- 1	3,058.9	-1	16,211.9		0.38	- 1
	Chairs	- 1	14,725.3	- 1	75,233.2		-4.37	- 1
	Art	- 1	924.1	-1	5,079.2	-1	-4.76	- 1
	Furnishings	s	4,236.2	-1	16,428.4	-1	-27.49	- 1
	Tables	- 1	10,315.9	- 1	39,899.5		-31.94	- 1
	Copiers	- 1	8,780.3	-1	26,561.8	-1	-37.80	- 1
	Appliances	s	6,374.4	-1	17,493.9		-52.83	I
	Fasteners	1	40.1		87.9		-76.29	1

# Insights and Results



# Key Findings

### Technology Leads Growth:

• Categories like Machines and Phones saw significant growth in 2023, with increases of 64.54% and 21.21% respectively.

### Declines in Traditional Categories:

• Furniture categories such as Chairs (-4.37%) and Tables (-31.94%) faced challenges in sales growth.

#### Rising Demand in Office Essentials:

• Office Supplies, including Supplies and Binders, showed strong growth, indicating increased demand driven by evolving work environments.

### **Business Impact**

### Focus on High-Growth Technology:

 Prioritize products like Machines and Phones to capitalize on substantial growth opportunities.

### **Strategic Pivot in Furniture Categories:**

 Address declines in traditional segments like Chairs and Tables through targeted adjustments and resource reallocation.

### **Expand in Office Supplies:**

Seize opportunities in growing categories such as Office Supplies (Supplies, Binders)
 to enhance market presence and meet evolving consumer demands.

### **Adapt to Remote Work Trends:**

 Tailor product offerings to support remote work environments, ensuring alignment with changing consumer behaviors and needs.

# Thank You!

Your engagement and attention are greatly appreciated.

Have a great day!