Retail Orders Data Analysis













Downloading Data

Data Cleaning

and Processing



Loading

Data



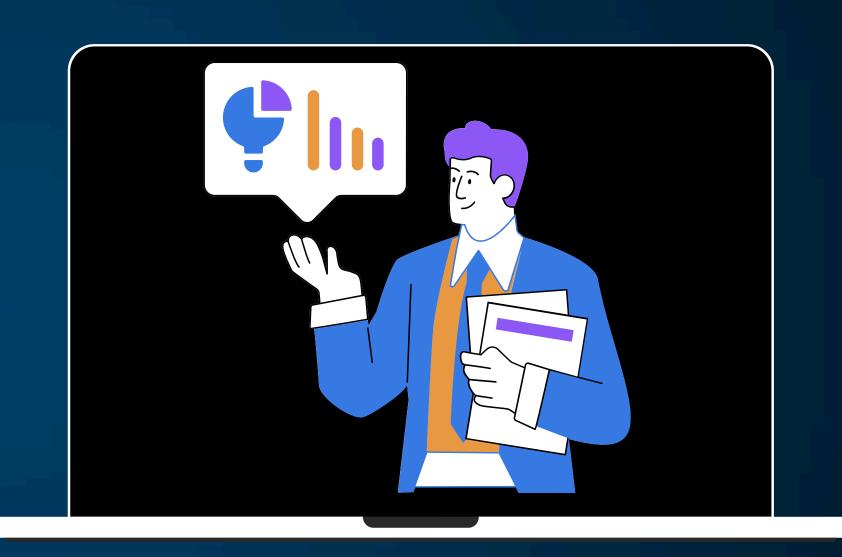
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Introduction

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.



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 generating products.

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

Region	Product ID Quantity	Ranking	1
Central	OFF-BI-10000301 34	1	 I
Central	OFF-BI-10000756 33	2	i
Central	OFF-BI-10000546 29	3	
Central	OFF-BI-10001249 29	3	
Central	FUR-CH-10002304 27	4	
Central	OFF-AP-10001947 27	4	
Central	FUR-CH-10002880 26	5	
East	OFF-PA-10001970 33	1	
East	OFF-BI-10003656 32	2	
East	OFF-FA-10000621 31	3	I
East	FUR-FU-10004848 31	3	
East	OFF-FA-10002780 29	4	- 1
East	OFF-ST-10002615 29	4	- 1
East	OFF-BI-10001524 28	5	
South	OFF-ST-10003716 26	1	
South	FUR-CH-10000513 24	2	- 1
South	OFF-BI-10004728 24	2	- 1
South	OFF-BI-10000014 23	3	- 1
South	FUR-FU-10001731 21	4	- 1
South	OFF-BI-10000069 21	4	- 1
South	OFF-BI-10000977 20	5	- 1
South	OFF-BI-10001191 20	5	- 1
South	TEC-AC-10000023 20	5	- 1
South	TEC-PH-10001459 20	5	- 1
South	FUR-TA-10000198 20	5	I
West	TEC-AC-10003832 45	1	- 1
West	OFF-BI-10000174 32	2	- 1
West	OFF-BI-10001036 31	3	I
West	OFF-BI-10001670 29	4	- 1
West	OFF-ST-10002486 29	4	I
West	FUR-FU-10001979 28	5	I
West	TEC-AC-10002006 28	5	I

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)
         month(order_date);
ORDER BY
```

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

```
2022
Month
                        2023
          437,431.3
                      88,632.6
          444,011.1 | 128,124.2
          394,105.2
                      82,512.3
          476,400.9 | 111,568.6
          413,625.5
                      86,447.9
          465,300.3
                      68,976.5
                      90,563.8
          375,278.4
          534,562.4
                      87,733.6
          433,887
                      76,658.6
          601,707.8
                     121,061.5
          451,809.6
                      75,432.8
          447,421.8 | 102,556.1
12
```

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.

```
| Category | Month | Sales |
| Furniture | 8-2023 | 230,523.5|
| Office Supplies | 2-2023 | 287,244.6|
| Technology | 10-2023 | 295,586.5|
```

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
         growth_pct DESC ;
ORDER BY
```

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

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

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!