

# *Data warehousing Project*

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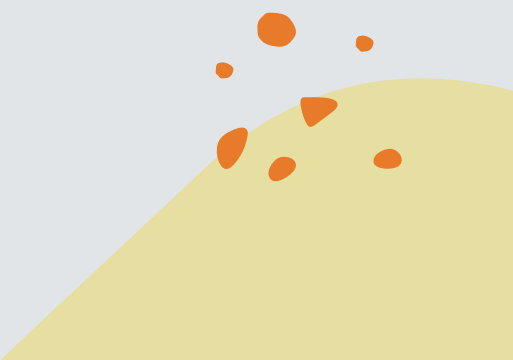
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# 1. Introduction: *The "Business Case"*

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**Problem Statement:** Traditionally, sales data in Excel is "flat" and "static." It lacks historical tracking and makes it difficult to see relationships between different entities (like how a specific category performs across different months).

**The Goal:** To transition from **Operational Data** (Excel) to **Analytical Data** (Data Warehouse).

**Define exactly what you are measuring:**

**Gross Profit:** Total Sales - Cost of Goods Sold (COGS).

**Sales Volume:** Total units (pieces) sold.

**Product Penetration:** Which products are moving the fastest.

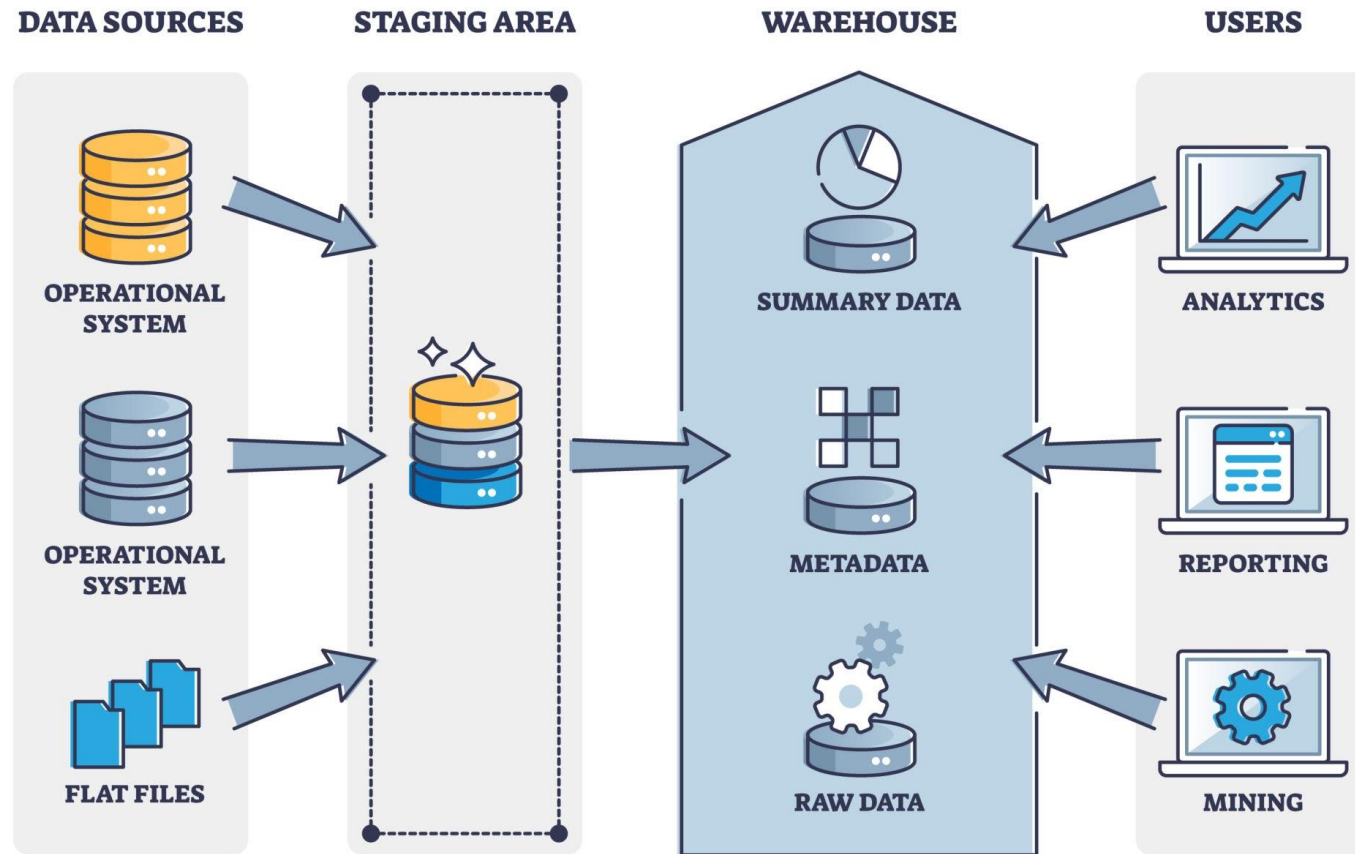
## 2. *The Data Architecture* (*The "How"*)

**In this part we will describe the ETL process (Extract, Transform, Load)**

**Data Source Layer:** our source was Excel, which represents "unstructured" or "semi-structured" raw business data.

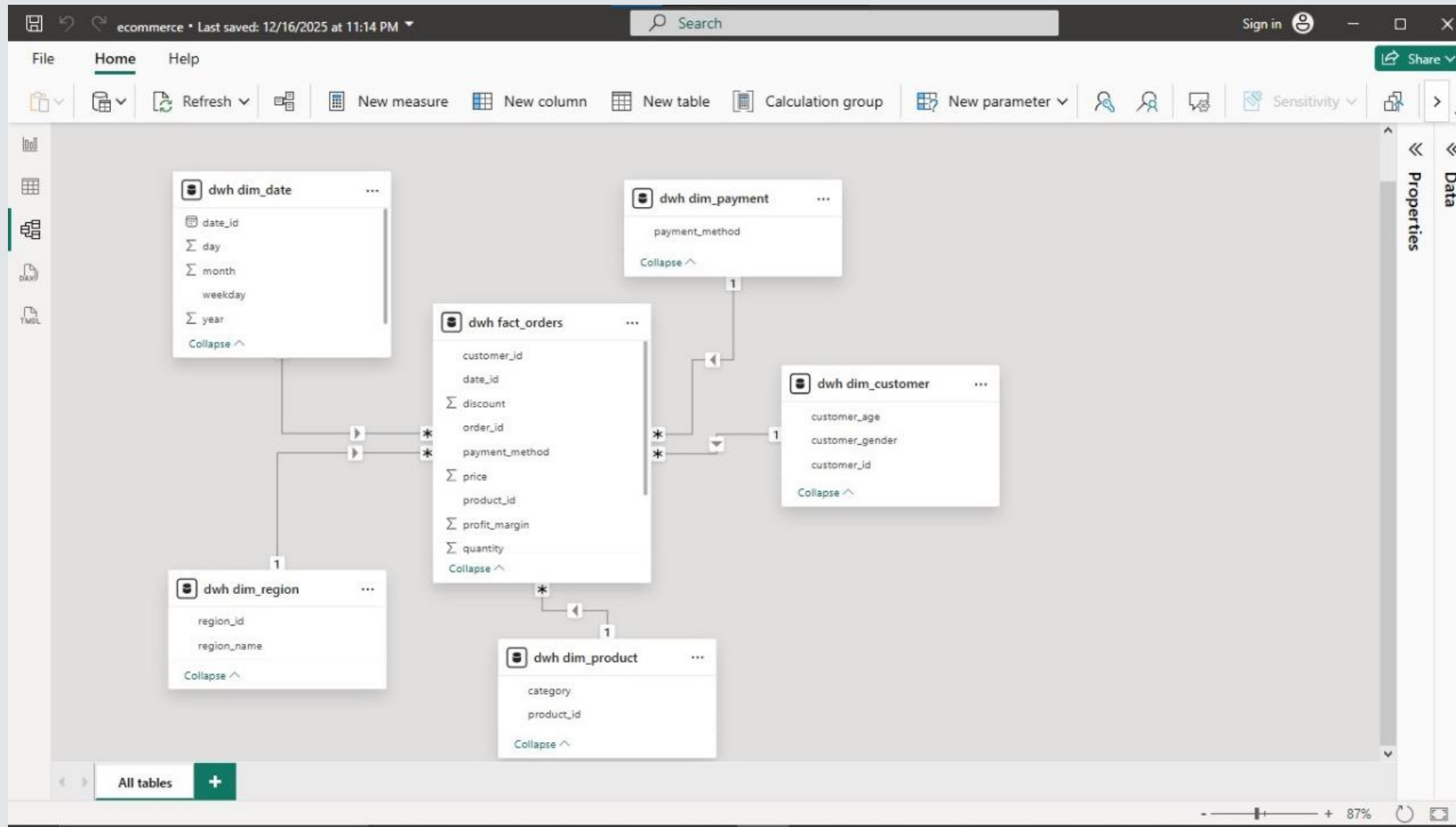
- **Staging Area (The "Buffer"):** This is crucial. Tell them you used a staging area to:
- **Cleanse:** Remove null values or fixing "messy" product names.
- **Standardize:** Ensure all currency and date formats are uniform before they hit the SQL database.
- **Data Warehouse Layer (SQL):** This is the permanent, structured storage where data is optimized for reading and reporting, not just storage.

# DATA WAREHOUSE



### 3. The Data Model (The "Logic")

Our data model follows a **Star Schema** architecture, designed to optimize query performance and simplify the analytical process within Power BI.



## *A. The Central Fact Table: dwh\_fact\_orders*

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This is the core of our warehouse. It captures every business event (an order) and stores the numerical data we need for calculation.

**Keys for Integration:** It contains Foreign Keys such as customer\_id, date\_id, product\_id, and region\_id to connect with our descriptive dimensions.

**Core Metrics:** We store the fundamental values here: price, quantity, and discount.

**Derived Logic:** We included a profit\_margin field directly in the fact table to allow for immediate profitability analysis without complex runtime calculations.

**Granularity:** The "grain" of this table is a **single order line item**, ensuring we can aggregate data up to any level (e.g., total sales by region or average profit by customer gender).

## ***B. The Dimension Tables (The Descriptive Entities)***

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We created five distinct dimension tables to provide context to our order data:

**dwh\_dim\_date:** Enables time-series analysis through attributes like day, month, weekday, and year. This allows us to track "Pieces Sold" over specific calendar periods.

**dwh\_dim\_product:** Stores product metadata, specifically category. This is essential for identifying which product lines are the most profitable.

**dwh\_dim\_customer:** Contains demographic data like customer\_age and customer\_gender, allowing us to see which audience segments are buying the most pieces.

**dwh\_dim\_region:** Organized by region\_id and region\_name to visualize geographic sales performance.

**dwh\_dim\_payment:** Tracks the payment\_method used for each order, helping us understand customer purchasing behavior.



## *C. Relationship Logic & Schema Integrity*

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**Star Schema Efficiency:** As seen in our model diagram, the `dwh_fact_orders` table is at the center, surrounded by dimensions. This reduces the number of "joins" needed, making the Power BI dashboard faster.

**1:N (One-to-Many) Relationships:** We established 1:N relationships between the dimensions and the fact table. For example, one entry in `dwh_dim_product` can relate to many rows in `dwh_fact_orders`.

**Directional Filtering:** The filters flow from the Dimension tables to the Fact table (represented by the arrows in the diagram). This ensures that when we select a specific "Category" in our dashboard, it correctly filters the "Total Profit" in our Fact table.

## *4. Analysis & Power BI (The "Insight")*

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In this phase, **we transformed the structured data from our SQL Warehouse into a visual "Semantic Layer."** This allows stakeholders to move from viewing rows of data to identifying business trends instantly.

## *A. Core Metrics.*

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We utilized high-level "Card" visualizations to provide an immediate snapshot of the business's health:

**Total Revenue:** indicating a high-volume operation.

**Profit Performance:** which is our primary indicator of business success beyond just sales.

**Inventory Throughput:** helping us understand the scale of physical product movement.

## *B. Visual Analysis & Behavioral Trends*

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- Demographic Profitability (Line Chart):** We analyzed the relationship between customer\_age and sales. This line chart reveals how different age groups contribute to revenue, allowing us to target marketing toward the peaks seen in the data.
- Regional Performance (Bar Chart):** By plotting region\_name, we discovered that the **South** and **North** regions are leading in sales, while the **Central** region is currently the lowest performer.
- Gender Distribution (Donut Chart):** Our customer base is almost evenly split, with **Females** slightly outperforming **Males**. This insight is crucial for product procurement and inventory planning.

## *C. Operational & Payment Analysis*

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Integrating the dwh\_dim\_payment and shipping data allowed us to see how customers interact with the business:

**Payment Preferences: Credit Cards** are the dominant payment method.

**Shipping Costs by Region:** We tracked the Max of shipping\_cost per region.

**Discount Impact:** Our analysis shows a sharp decline in sales volume as discount levels increase.

## *D. Operational Efficiency & Behavior*

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**Payment Method Impact: Credit Cards.**

**Logistics Overhead:** identifying a clear area for cost reduction.

**Discount Sensitivity:** Our analysis shows that volume peaks at **18.9K** when discounts are at **0%**, suggesting that our customers are currently more driven by product value than price-slashing.



*Screens from Our Project.*



pgAdmin 4

FileObjectToolsEditViewWindowHelp

Object Expl

Servers (2)

Local PostgreSQL

Databases (2)

ecommerce\_dw

Casts

Catalogs

Event Triggers

Extensions

Foreign Data Wrappers

Languages

Publications

Schemas (4)

dw

dwh

public

staging

Subscriptions

postgres

Login/Group Roles

Tablespaces

PostgreSQL 18

ecommerce\_dw/postgres@Local PostgreSQL\*

ecommerce\_dw/postgres@Local PostgreSQL

No limit

Query

Query History

```
1
2 CREATE TABLE staging.orders (
3     order_id TEXT,
4     customer_id TEXT,
5     product_id TEXT,
6     category TEXT,
7     price NUMERIC,
8     discount NUMERIC,
9     quantity INT,
10    payment_method TEXT,
11    order_date DATE,
12    delivery_time_days INT,
13    region TEXT,
14    returned TEXT,
15    total_amount NUMERIC,
16    shipping_cost NUMERIC,
17    profit_margin NUMERIC,
18    customer_age INT,
19    customer_gender TEXT
20 );
21
22
```

Total rows:

CRLF

Ln 16, Col 27



pgAdmin 4

File Object Tools Edit View Window Help

ecommerce\_dw/postgres@PostgreSQL 18\* x

ecommerce\_dw/postgres@PostgreSQL 18

No limit

Query Query History

```
1 CREATE TABLE dwh.fact_orders (  
2     order_id TEXT PRIMARY KEY,  
3     customer_id TEXT REFERENCES dwh.dim_customer(customer_id),  
4     product_id TEXT REFERENCES dwh.dim_product(product_id),  
5     date_id DATE REFERENCES dwh.dim_date(date_id),  
6     region_id TEXT REFERENCES dwh.dim_region(region_id),  
7     payment_method TEXT REFERENCES dwh.dim_payment(payment_method),  
8     price NUMERIC,  
9     discount NUMERIC,  
10    quantity INT,  
11    total_amount NUMERIC,  
12    shipping_cost NUMERIC,  
13    profit_margin NUMERIC  
14 );  
15
```

Data Output Messages Notifications

CREATE TABLE

Query returned successfully in 146 msec.

Total rows: Query complete 00:00:00.146 CRLF Ln 15, Col 1

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Copy code

Fact Table 2

Copy code

Explain

Microphone icon

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File Object Tools Edit View Window Help

ecommerce\_dw/postgres@PostgreSQL 18\*

ecommerce\_dw/postgres@PostgreSQL 18

No limit

Query Query History

```
1 -- Dimension Customer
2 CREATE TABLE dwh.dim_customer (
3     customer_id TEXT PRIMARY KEY,
4     customer_age INT,
5     customer_gender TEXT
6 );
7 -- Dimension Product
8 CREATE TABLE dwh.dim_product (
9     product_id TEXT PRIMARY KEY,
10    category TEXT
11 );
12 -- Dimension Date
13 CREATE TABLE dwh.dim_date (
14     date_id DATE PRIMARY KEY,
15     day INT,
16     month INT,
17     year INT,
18     weekday TEXT
19 );
20 -- Dimension Region
21 CREATE TABLE dwh.dim_region (
22     region_id TEXT PRIMARY KEY,
23     region_name TEXT
24 );
25 -- Dimension Payment
26 CREATE TABLE dwh.dim_payment (
27     payment_method TEXT PRIMARY KEY
```

Total rows: CRLF Ln 24, Col 3

ecommerce\_dw

- Casts
- Catalogs
- Event Triggers
- Extensions
- Foreign Data Wrappers
- Languages
- Publications
- Schemas (3)
  - dw
  - public
  - staging
    - Aggregates
    - Collations
    - Domains
    - FTS Configurations
    - FTS Dictionaries
    - FTS Parsers
    - FTS Templates
    - Foreign Tables
    - Functions
    - Materialized Views
    - Operators

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Copy code

Microphone icon

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File Object Tools Edit View Window Help

ecommerce\_dw/postgres@PostgreSQL 18\* X

ecommerce\_dw/postgres@PostgreSQL 18

No limit

Query Query History

```
1 -- dim_customer
2 INSERT INTO dwh.dim_customer(customer_id, customer_age, customer_gender)
3 SELECT DISTINCT customer_id, customer_age, customer_gender
4 FROM staging.orders;
5 -- dim_product
6 INSERT INTO dwh.dim_product(product_id, category)
7 SELECT DISTINCT product_id, category
8 FROM staging.orders;
9 -- dim_date
10 INSERT INTO dwh.dim_date(date_id, day, month, year, weekday)
11 SELECT DISTINCT order_date,
12     EXTRACT(DAY FROM order_date)::INT,
13     EXTRACT(MONTH FROM order_date)::INT,
14     EXTRACT(YEAR FROM order_date)::INT,
15     TO_CHAR(order_date, 'Day')
16 FROM staging.orders;
17 -- dim_region
18 INSERT INTO dwh.dim_region(region_id, region_name)
19 SELECT DISTINCT region, region
20 FROM staging.orders;
```

Data Output Messages Notifications

CREATE TABLE

Query returned successfully in 192 msec.

Total rows: Query complete 00:00:00.192 CRLF Ln 4, Col 21

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ta into Dimensions 3

Copy code

Explain

pgAdmin 4

File Object Tools Edit View Window Help

ecommerce\_dw/postgres@PostgreSQL 18\*

ecommerce\_dw/postgres@PostgreSQL 18

Query Query History

```
1 INSERT INTO dwh.fact_orders(  
2     order_id, customer_id, product_id, date_id, region_id,  
3     payment_method, price, discount, quantity, total_amount,  
4     shipping_cost, profit_margin  
5 )  
6 SELECT  
7     order_id, customer_id, product_id, order_date, region,  
8     payment_method, price, discount, quantity, total_amount,  
9     shipping_cost, profit_margin  
10 FROM staging.orders  
11 ON CONFLICT (order_id) DO NOTHING;  
12
```

Data Output Messages Notifications

INSERT 0 6

Query returned successfully in 2 secs 307 msec.

Total rows: Query complete 00:00:02.307 CRLF Ln 12, Col 1

Share Add people

Copy code

data into Fact Table 5

Copy code

Explain





5.87M

Sum of total\_amount

970K

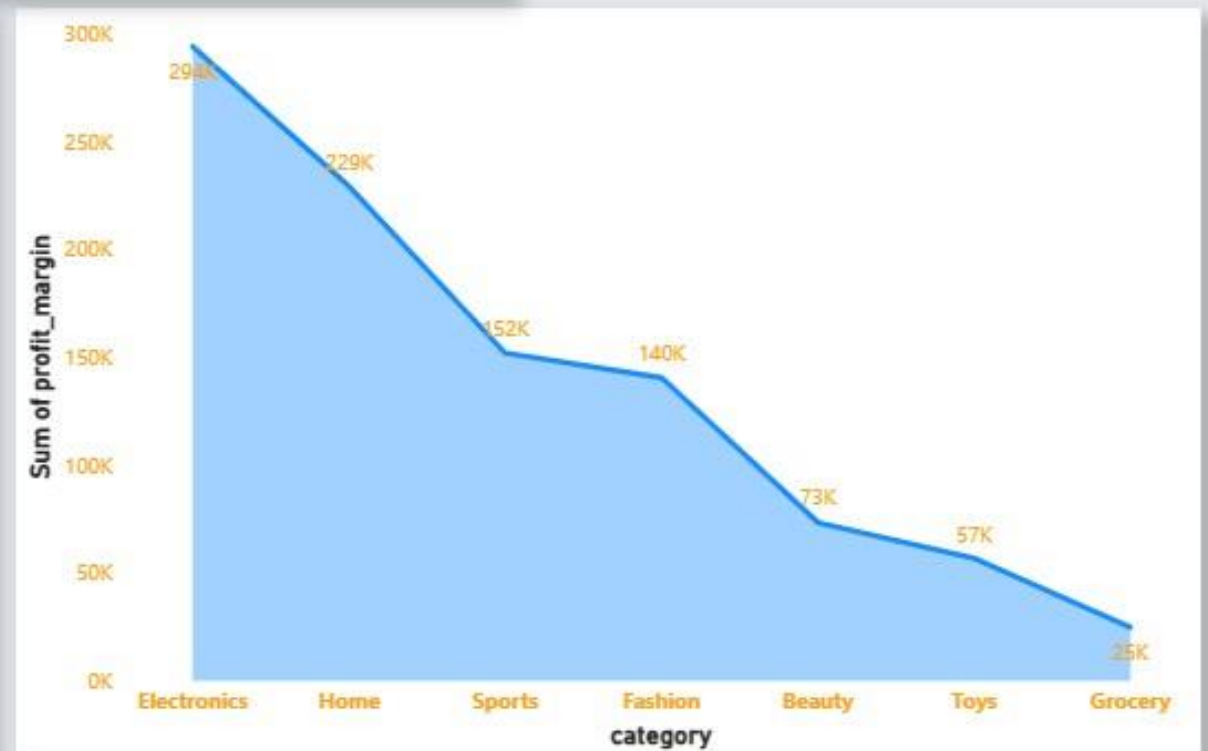
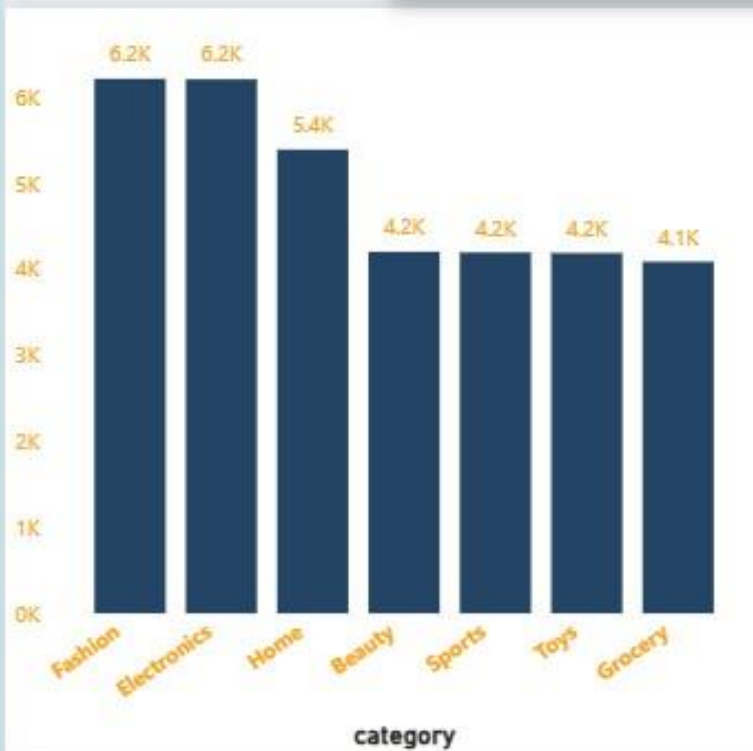
Sum of profit\_margin

51K

Sum of quantity

category

All





5.87M

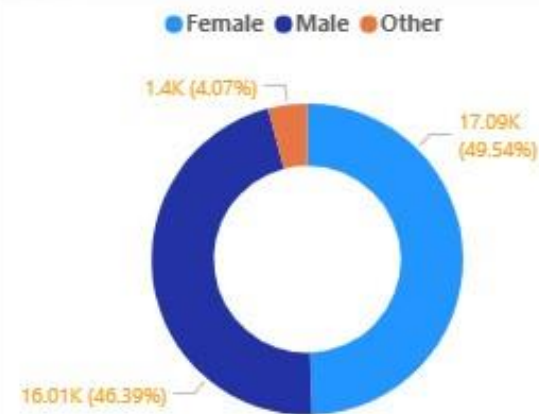
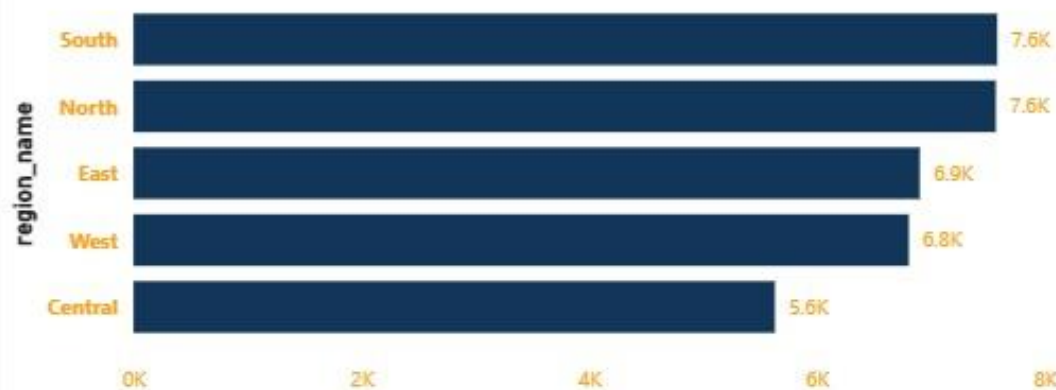
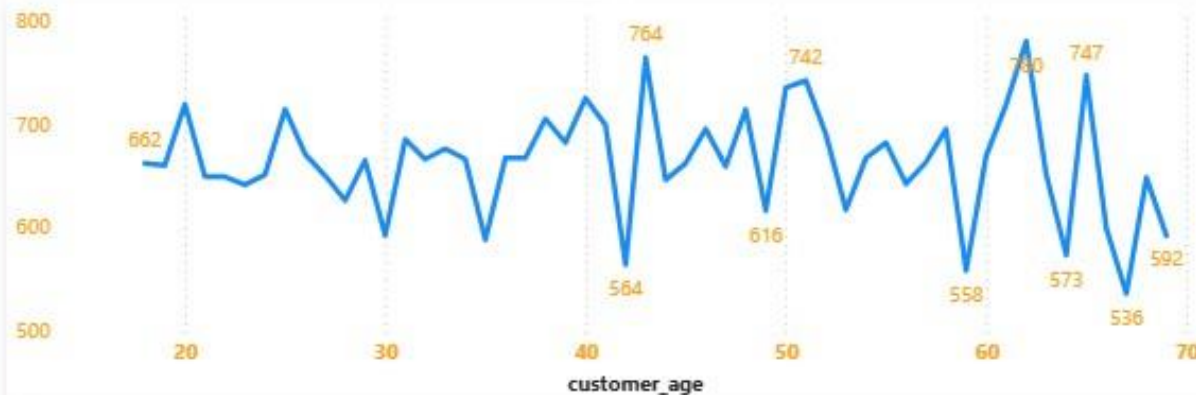
Sum of total\_amount

970K

Sum of profit\_margin

51K

Sum of quantity





# 5.87M

Sum of total\_amount

# 970K

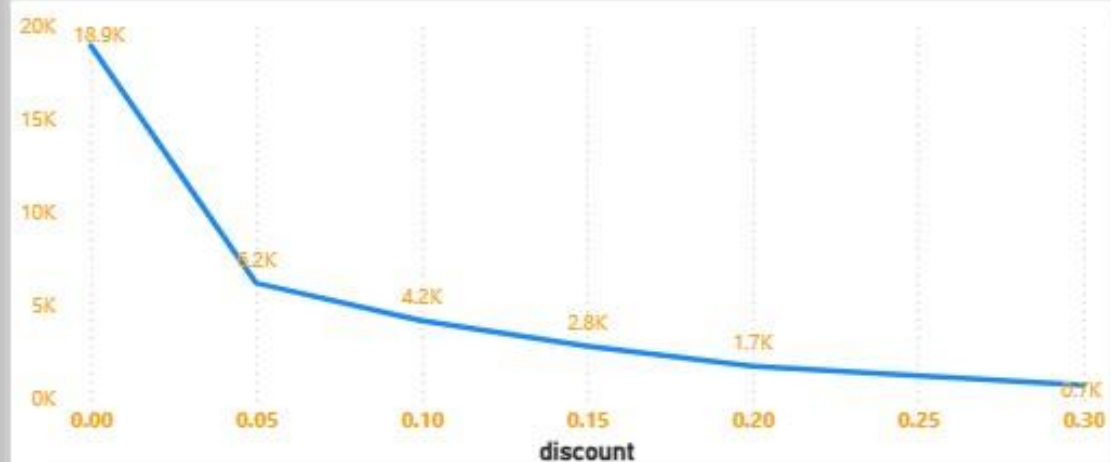
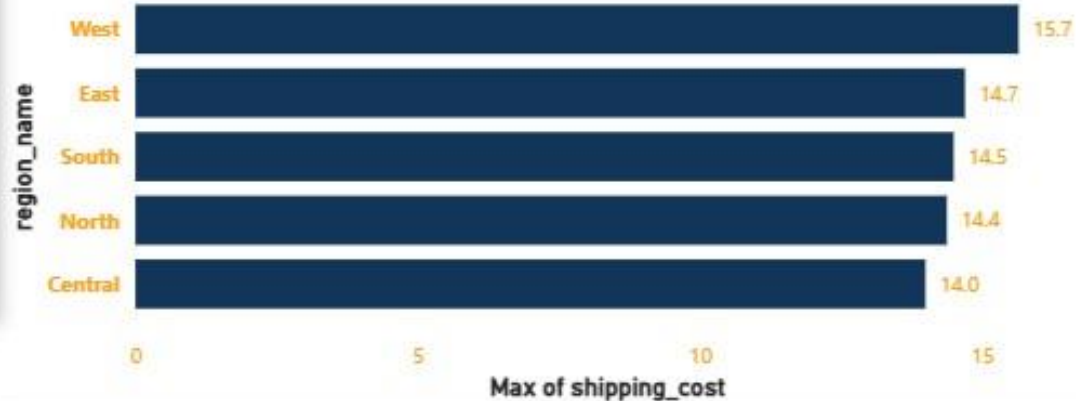
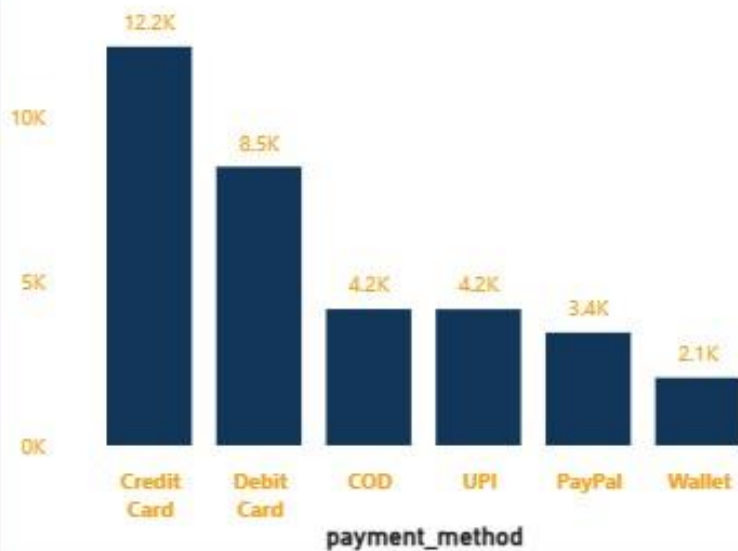
Sum of profit\_margin

# 51K

Sum of quantity

payment\_method

Select all	Credit Card	PayPal	Wallet
COD	Debit Card	UPI	



## 6. Conclusion & Value.

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This project successfully bridged the gap between raw data collection and strategic decision-making.

**From Data to Strategy:** We can now see that while we sell many "pieces" in Fashion, our true "profit" engine is the Electronics category.

**Scalability:** Because we built this in a **SQL Star Schema** rather than a flat Excel file, this warehouse can handle future growth without needing to redesign the logic.

**Final Impact:** We have created a transparent, automated system that gives the business a **Single Version of the Truth** for every piece sold and every dollar earned.



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# Bye-Bye

