

Data warehousing Project

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1. Introduction: The "Business Case"

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.



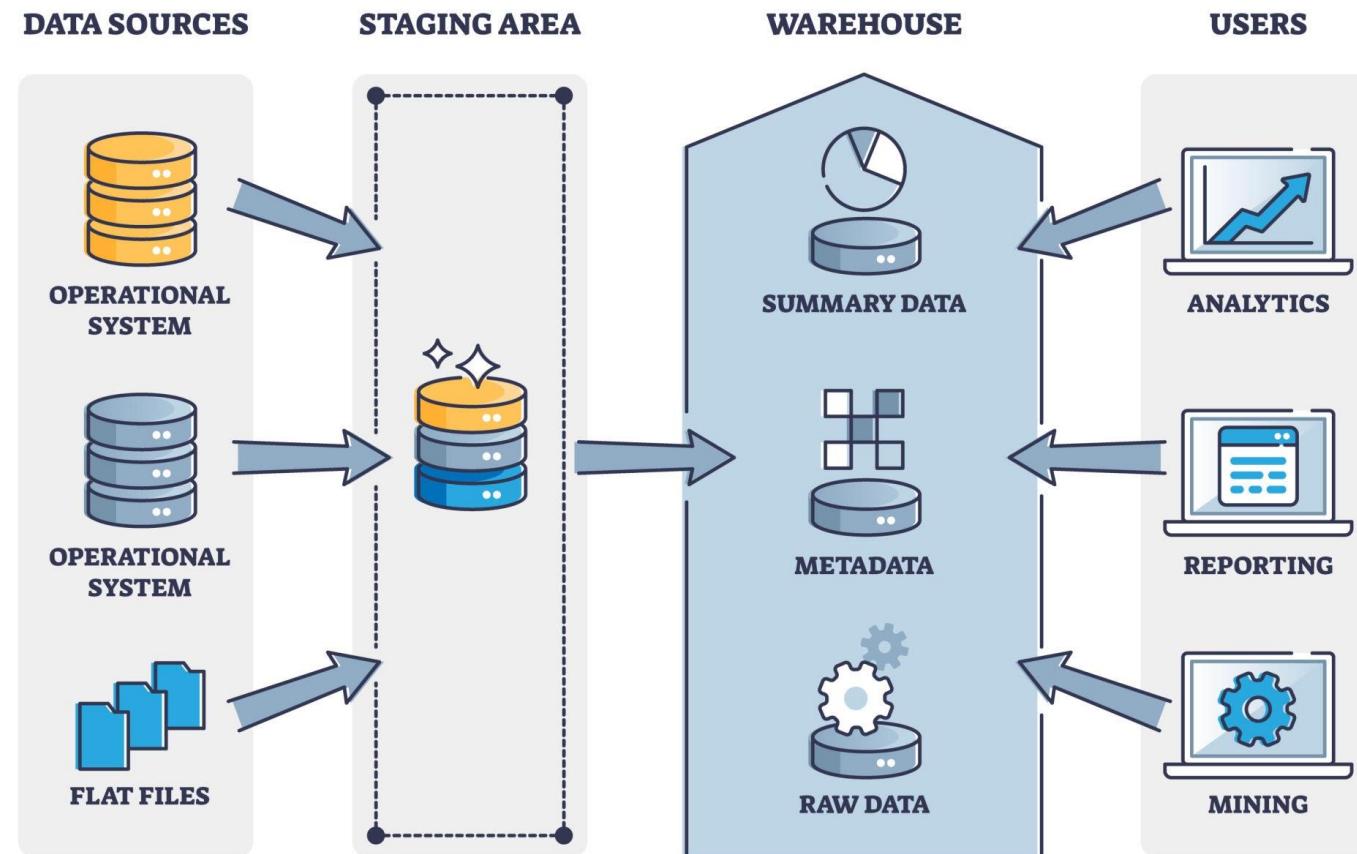
*2. The Data Architecture (*The "How"*)*

In this part we will describe the ETL process ([Extract, Transform, Load](#))

Data Source Layer: our8 source was Excel, which represents "unstructured" or "semi-structured" raw business data.

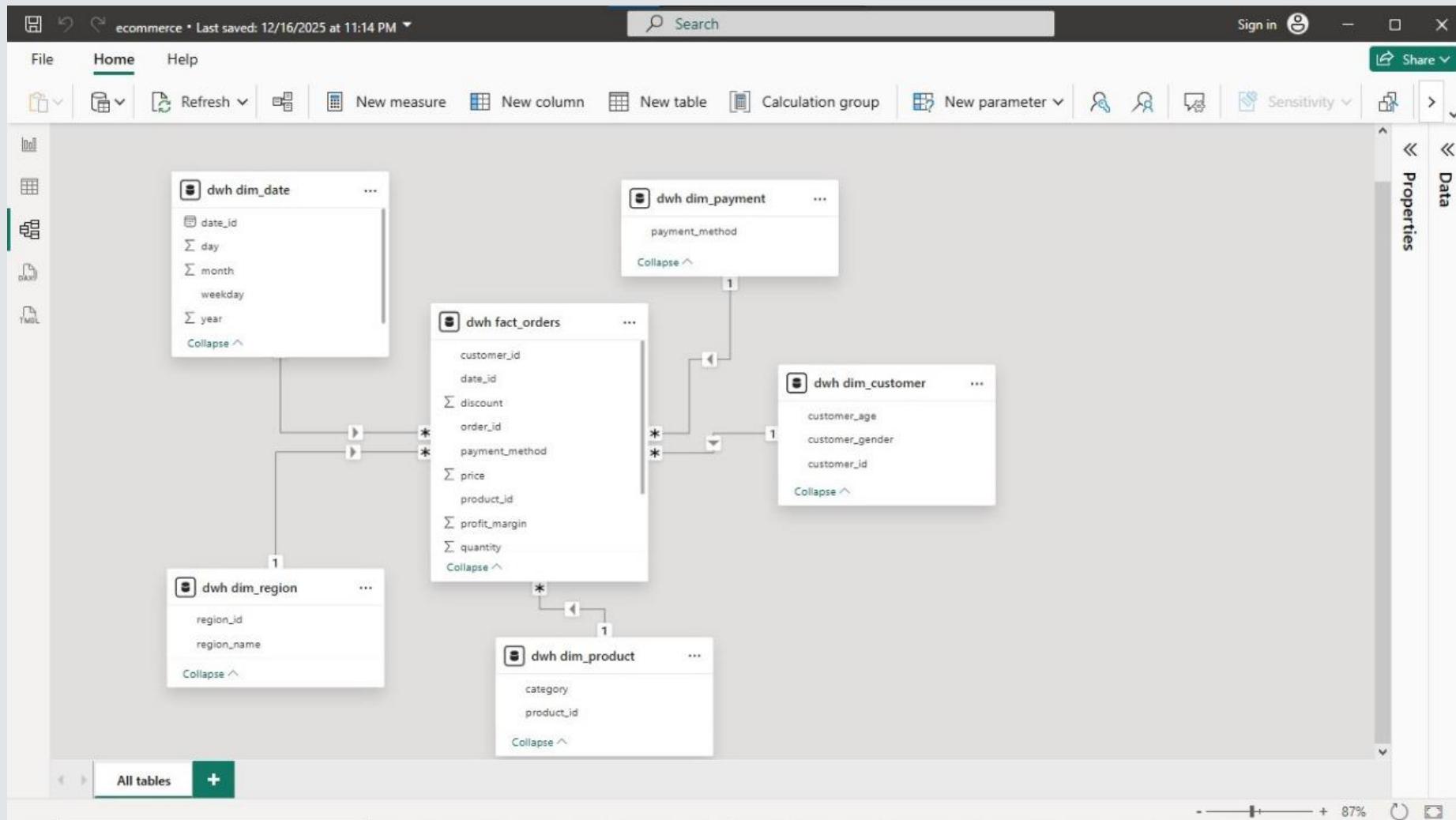
- **Staging Area (The "Buffer"):** This is crucial. Tell them you used a staging area to:
 - **Cleanse:** Remove null values or fixing "messy" product names.
 - **Standardize:** Ensure all currency and date formats are uniform before they hit the SQL database.
- **Data Warehouse Layer (SQL):** This is 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

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)

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

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")

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.

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

- **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

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

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.

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File Object Tools Edit View Window Help

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*

No limit

Query History

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

Total rows:

CRLF Ln 16, Col 27

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File Object Tools Edit View Window Help

Servers (2)
Local PostgreSQL
PostgreSQL 18
Databases (2)
ecommerce_dw
Casts
Catalogs
Event Triggers
Extensions
Foreign Data Wrap
Languages
Publications
Schemas (4)
dw
dwh
public
staging
Subscriptions
postgres
Login/Group Roles (17)
Tablespaces (2)

ecommerce_dw/postgres@PostgreSQL 18* X

ecommerce_dw/postgres@PostgreSQL 18

No limit

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

Share Add people ...

Copy code

Fact Table 2

Copy code

Explain

pgAdmin 4

File Object Tools Edit View Window Help

ecommerce_dw/postgres@PostgreSQL 18* X

Databases (2)

 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

Query History

```
-- Dimension Customer
CREATE TABLE dwh.dim_customer (
    customer_id TEXT PRIMARY KEY,
    customer_age INT,
    customer_gender TEXT
);
-- Dimension Product
CREATE TABLE dwh.dim_product (
    product_id TEXT PRIMARY KEY,
    category TEXT
);
-- Dimension Date
CREATE TABLE dwh.dim_date (
    date_id DATE PRIMARY KEY,
    day INT,
    month INT,
    year INT,
    weekday TEXT
);
-- Dimension Region
CREATE TABLE dwh.dim_region (
    region_id TEXT PRIMARY KEY,
    region_name TEXT
);
-- Dimension Payment
CREATE TABLE dwh.dim_payment (
    payment_method TEXT PRIMARY KEY
)
```

Total rows: 0 CRLF Ln 24, Col 3

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File Object Tools Edit View Window Help

Servers (2)
Local PostgreSQL
PostgreSQL 18
Databases (2)
ecommerce_dw (selected)
Casts
Catalogs
Event Triggers
Extensions
Foreign Data Wrap
Languages
Publications
Schemas (4)
dw
dwh
public
staging (selected)
Subscriptions
postgres
Login/Group Roles (17)
Tablespaces (2)

ecommerce_dw/postgres@PostgreSQL 18* X

No limit

Query History

```
-- dim_customer
INSERT INTO dwh.dim_customer(customer_id, customer_age, customer_gender)
SELECT DISTINCT customer_id, customer_age, customer_gender
FROM staging.orders;
-- dim_product
INSERT INTO dwh.dim_product(product_id, category)
SELECT DISTINCT product_id, category
FROM staging.orders;
-- dim_date
INSERT INTO dwh.dim_date(date_id, day, month, year, weekday)
SELECT DISTINCT order_date,
    EXTRACT(DAY FROM order_date)::INT,
    EXTRACT(MONTH FROM order_date)::INT,
    EXTRACT(YEAR FROM order_date)::INT,
    TO_CHAR(order_date, 'Day')
FROM staging.orders;
-- dim_region
INSERT INTO dwh.dim_region(region_id, region_name)
SELECT DISTINCT region, region
FROM staging.orders;
```

Copy code Explain

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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File Object Tools Edit View Window Help

Servers (2)
Local PostgreSQL
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Databases (2)
ecommerce_dw
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Catalogs
Event Triggers
Extensions
Foreign Data Wrap
Languages
Publications
Schemas (4)
dw
dwh
public
staging
Subscriptions
postgres
Login/Group Roles (17)
Tablespaces (2)

ecommerce_dw/postgres@PostgreSQL 18* X

No limit

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;
```

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

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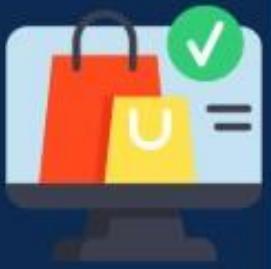
Copy code

Data into Fact Table 5

Copy code

Explain

0



5.87M

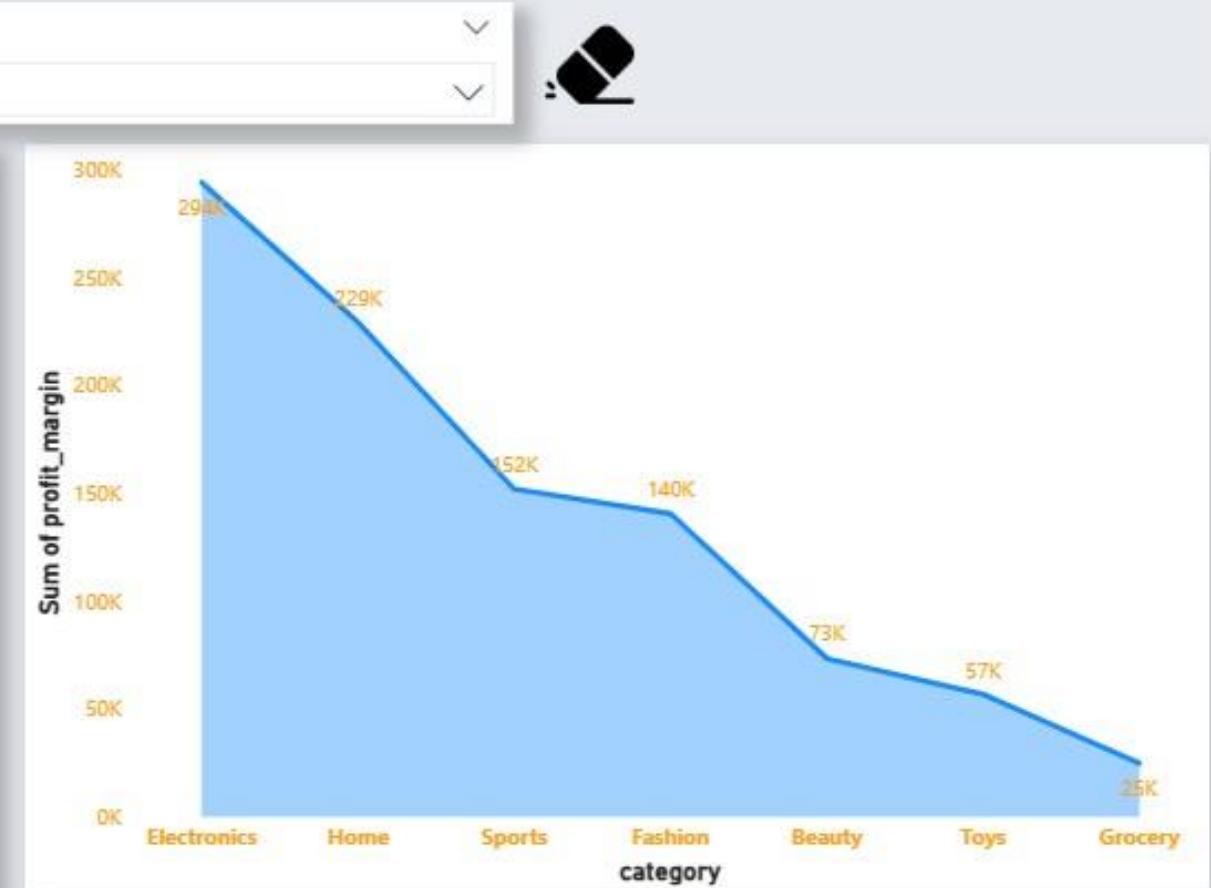
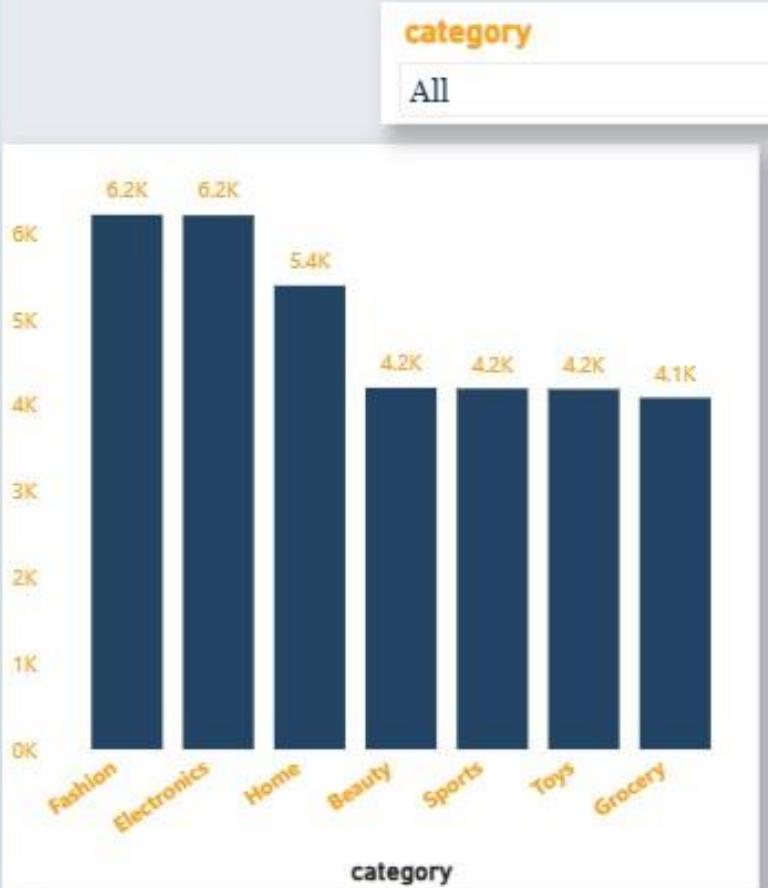
Sum of total_amount

970K

Sum of profit_margin

51K

Sum of quantity





5.87M

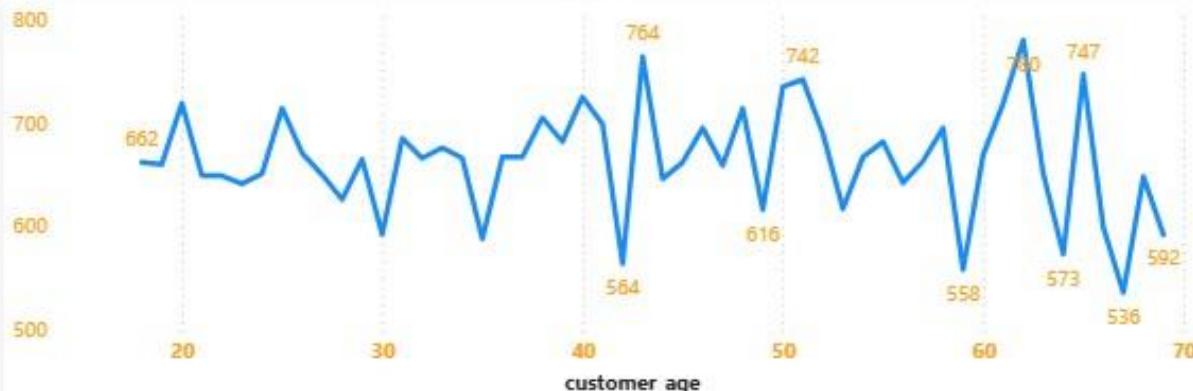
Sum of total_amount

970K

Sum of profit_margin

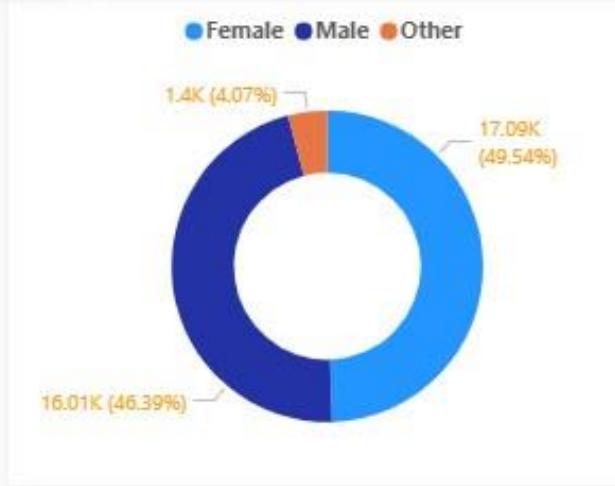
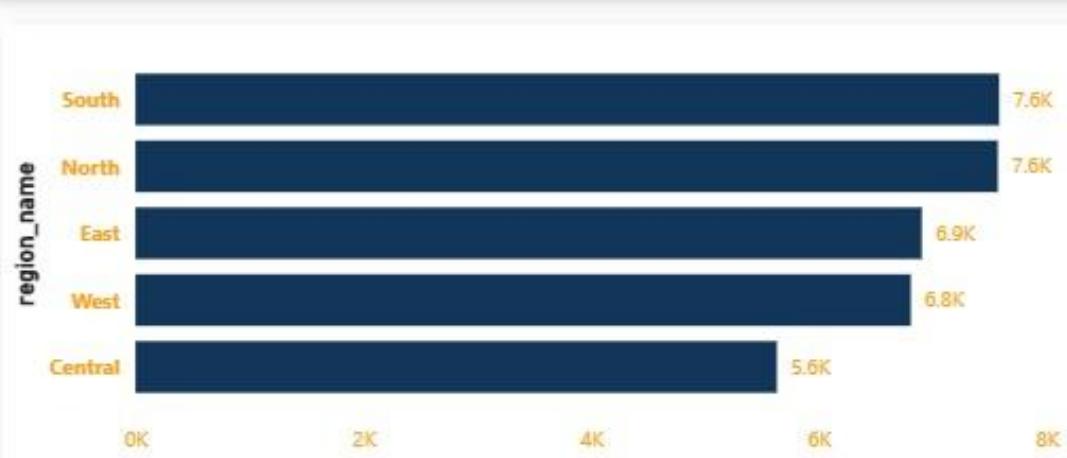
51K

Sum of quantity



region_name

| | |
|------------|-------|
| Select all | North |
| Central | South |
| East | West |





5.87M

Sum of total_amount

970K

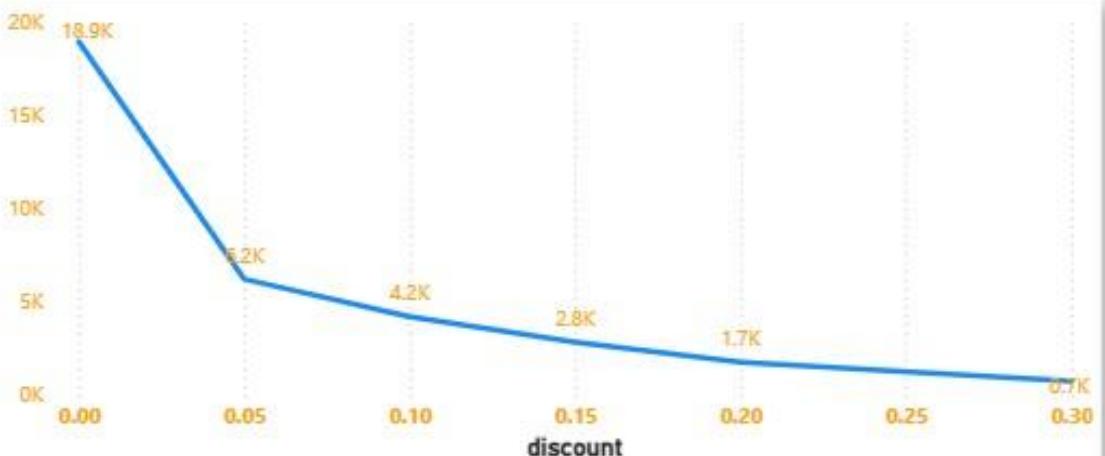
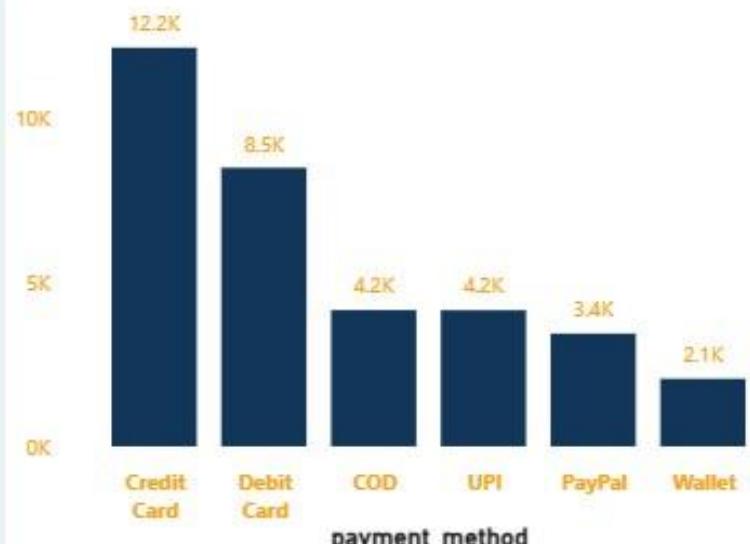
Sum of profit_margin

51K

Sum of quantity

payment_method

| | | | |
|------------|-------------|--------|--------|
| Select all | Credit Card | PayPal | Wallet |
| COD | Debit Card | UPI | |



6. Conclusion & Value.

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.



Bye-Bye

