

# **Sales Trend & Segment Analysis for Product Strategy**

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*(A Business Analytics Case Study for Lognormal Analytics)*

**Presented by:**

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# Objective Overview

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Main  
Objective

- Segment Influence Analysis
- Trend Detection
- Competitor Impact Assessment
- Additional Strategic Insights

# Data Understanding

## Dataset Summary

- Dataset covers **10 products** across **3 categories**: Home Goods | Clothing | Electronics
- Total **1040 rows** (weekly observations)
- Time span: **May 2023 – May 2025** ( 104 weeks)

```
RangeIndex: 1040 entries, 0 to 1039
Data columns (total 14 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Week                                  1040 non-null   datetime64[ns]
1   Product ID                           1040 non-null   int64
2   Category                             1040 non-null   object
3   Product Name                         1040 non-null   object
4   Price                                1040 non-null   float64
5   Competitor Price                     1040 non-null   float64
6   Pricing Currency                     1040 non-null   object
7   Sales Volume                         1023 non-null   float64
8   Advertising Spend (EUR)              1020 non-null   float64
9   Customer Reviews                     1034 non-null   float64
10  Weather Index                        1040 non-null   int64
11  Social Media Mentions                 1040 non-null   int64
12  Warehouse Stock Level                 1040 non-null   int64
13  Market Share Rank                     1040 non-null   int64
dtypes: datetime64[ns](1), float64(5), int64(5), object(3)
```

### Note\*:

- Sales Volume had some missing values -> Imputed using median
- Currency is standardized to EUR

# Segment Influence Analysis

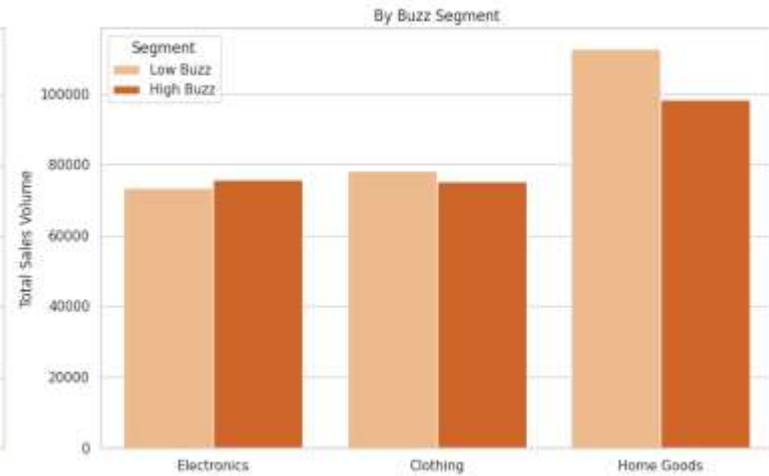
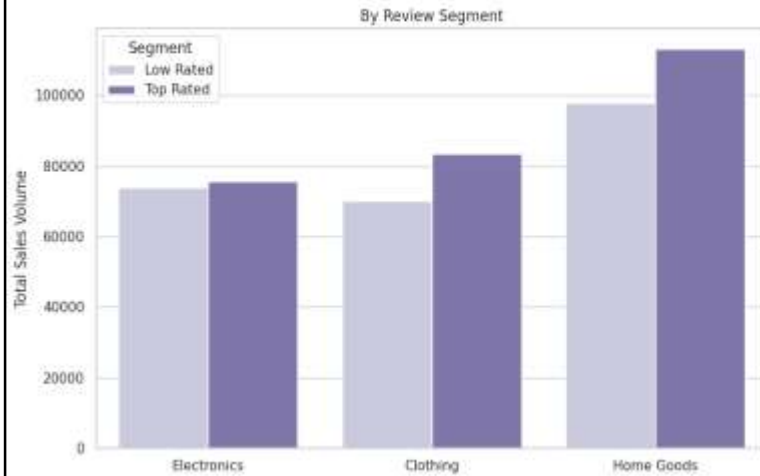
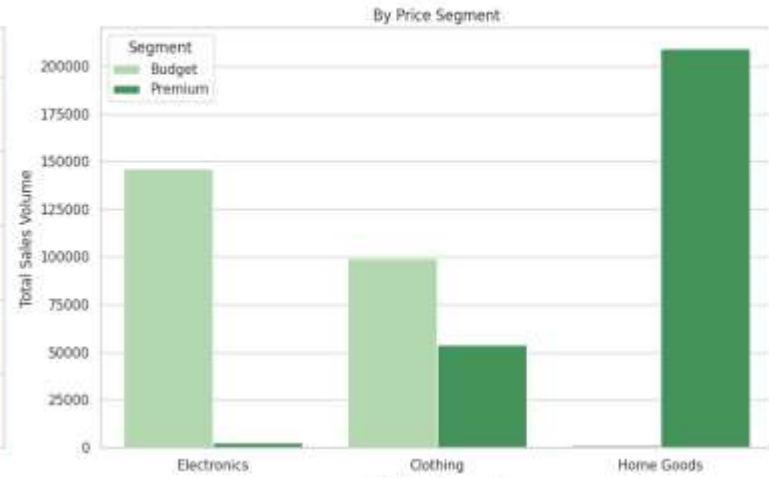
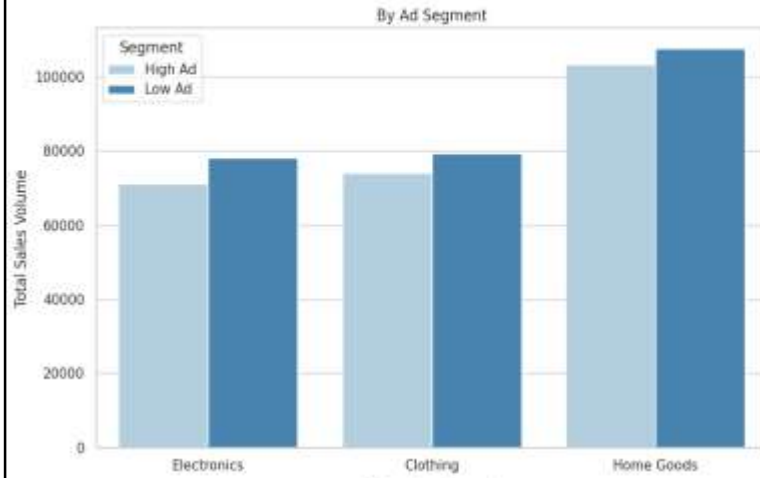
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**Goal:** Identify how different customer “types” (segments) are driving sales volume across categories.

**Segment Creation (Proxy Segments Used):** Since direct customer segments weren’t available, proxy segments were derived using:

- i. Price Tier** - Budget vs Premium
- ii. Buzz Level** - Based on Social Media Mentions
- iii. Customer Sentiment** - Based on Review Ratings
- iv. Ad Spend Level** - High vs Low Advertising Spend

## Sales Volume by Segment Within Each Category



# Key Insights

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🏠 **Home Goods** sold the **highest quantity**, especially in **premium tier** - Customers perceive value in quality.

👕 **Clothing** saw better performance in **budget segments**, indicating price-sensitive behavior.

📢 **High-buzz & high-review products** consistently sold more across all categories - Trust & awareness matter.

📈 **High Ad Spend did not guarantee higher sales** - In many cases, **low-ad products outperformed**, suggesting **ad inefficiency** or **brand-driven pull**.

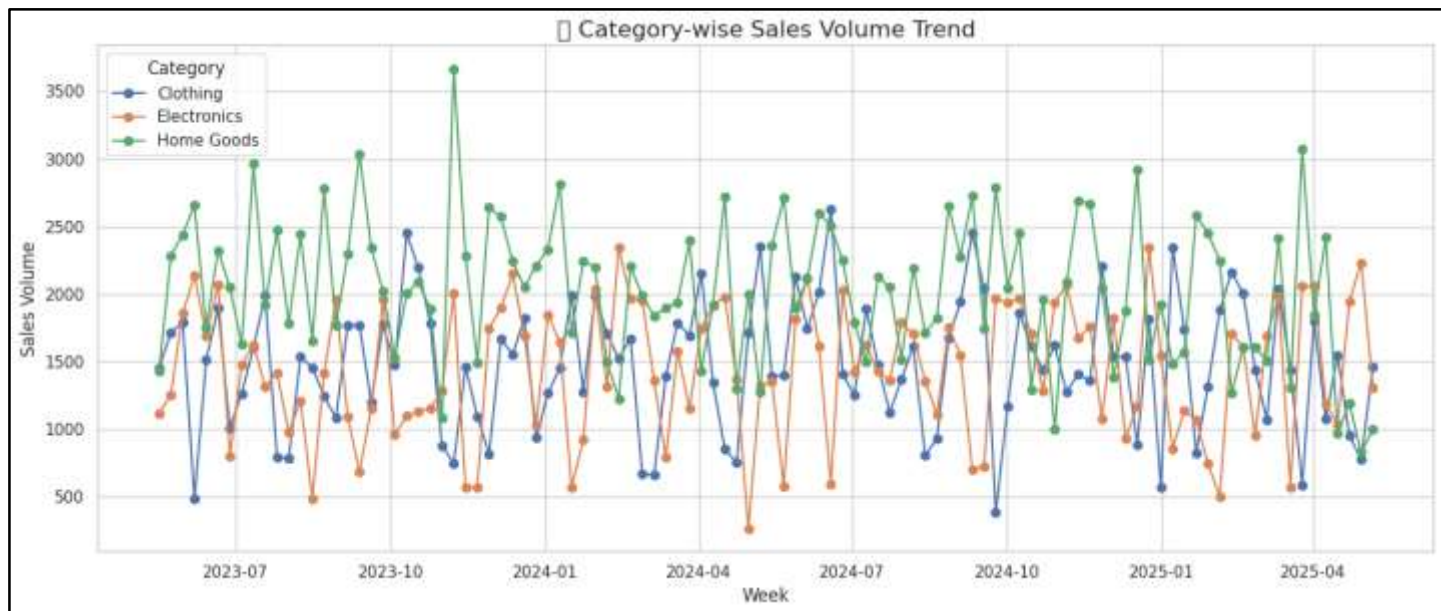
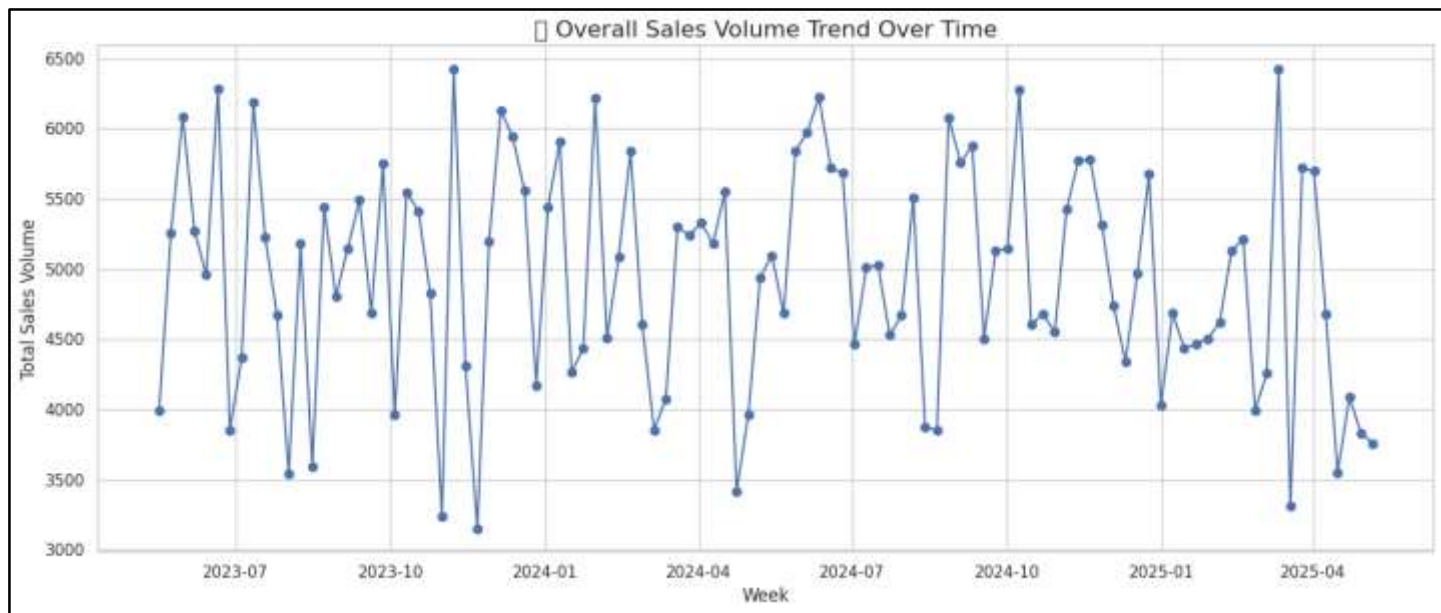
# Trend Detection

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**Goal:** Analyze how sales volume evolves over time to identify growth patterns, consistency, or seasonal trends.

## **Time Series Used**

- i. Weekly data over **104 weeks** (May 2023 – May 2025)
- ii. Used Week column to visualize time-based sales trends





# Key Trends Observed

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## Overall Sales Trend:

- Sales remained **stable** with no strong upward or downward pattern
- No visible **seasonality or weekly spikes** detected

## Category-wise Trends:

- **Home Goods:** Consistent performance over 2 years
- **Electronics:** Minor fluctuations but mostly stable
- **Clothing:** Volatile, lacking clear momentum

# Competitor Price Impact

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**Goal:** Measure how competitor pricing affects our weekly sales volume and whether undercutting or premium pricing changes buyer behavior.

**Approach:**

- Created a new metric:
  - $\text{Price Diff} = \text{Competitor Price} - \text{Our Price}$
- Check Correlation Between Price Diff v/s Sales Volume

# Key Findings

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- **Correlation coefficient  $\approx -0.005$** 
  - Virtually **no relationship** between price difference and sales volume
- Products with lower prices **did not necessarily** sell more
- In several categories, **premium-priced products performed equally well**

# Operational & Environmental Factor Check

(Objective 4: Additional Strategic Insights)

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**Goal:** Identify hidden operational or external influencers affecting sales volume.

## Variables Analysed

- A. Warehouse Stock Level vs Sales Volume**
  - Correlation: **-0.03** → Inventory availability **not a limiting factor**
- B. Weather Index vs Sales Volume**
  - Correlation: **+0.02** → No seasonal/weather-based impact detected
- C. Lag Effect of Advertising**
  - Correlation of ad spend (1-week lag) with sales: **insignificant**
  - Implies **ad strategy isn't boosting immediate sales**

# Recommendations

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## SEGMENT INFLUENCE (OBJECTIVE 1)

- i. Focus on Home Goods  
Premium Tier- high  
performance despite price
- ii. In Clothing, push budget-  
friendly SKUs - price-sensitive  
customers
- iii. Boost review generation &  
social media buzz - stronger  
sales drivers than price or ads

## TREND DETECTION (OBJECTIVE 2)

- i. No major seasonal spikes -  
Plan steady supply &  
marketing
- ii. Clothing shows volatility -  
Consider redesign or re-  
positioning
- iii. Use stable trends in  
Electronics & Home Goods for  
baseline forecasting

# Recommendations

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## **COMPETITOR PRICING (OBJECTIVE 3)**

- i. Price difference has minimal impact on volume
- ii. Avoid aggressive discounting - focus on value, trust & brand
- iii. Optimize pricing based on customer segment, not competition

## **OPERATIONAL FACTORS (OBJECTIVE 4)**

- i. Inventory levels & weather have no meaningful influence
- ii. Ads show no short-term lift - Reassess strategy (creative, timing, audience)
- iii. Invest in long-term perception-building over reactive tactics

# AI Tool Disclosure & Links

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## Used ChatGPT for:

- Structuring Python code (EDA, plots, correlations)
- Insight framing
- Report and PPT organization
- All final decisions, interpretations, and strategy directions reflect my original thought process and judgment.

## Links:

- [Google Drive](#)

(This Link Consist Google Collab Python Code & Report of the Case Study)

- [Tableau Dashboard](#)