

# **Zara: Maximising Gross Margin through Data-Driven Retail Strategy**

## **1. Introduction**

Zara is navigating the fast-moving world of fashion retail, where being quick and adaptable to changing consumer trends really gives it an edge. Yet, balancing healthy profits while keeping inventory moving at a rapid pace is always a challenge. This project is all about figuring out how Zara can fine-tune its gross margin, a key indicator of profitability, by tapping into data regarding product pricing, costs, discounts, and sales volume.

### **1.1 Overview of Zara's Business Model**

At the heart of Zara's approach is a smart blend of vertical integration and just-in-time manufacturing. This gives them control over design, production, distribution, and retail, enabling them to turn things around quickly and refresh their product offerings frequently. Unlike traditional fashion brands that plan their collections months in advance, Zara rolls out new styles every couple of weeks, which keeps customers coming back more often, eager to shop.

### **1.2 Importance of Gross Margin in Fashion Retail**

Gross margin—the profit leftover after covering the cost of goods sold (COGS)—is crucial in retail. It shines a light on how profitable products are before considering operational costs. For Zara, optimizing gross margin is essential because it directly influences how much they can invest in design, logistics, and technology. Higher gross margins mean better financial freedom and a sustainable future.

## **2. Problem Statement**

Zara's gross margin is influenced by various operational and strategic factors, including product pricing, production costs, discount strategies, and consumer demand. The aim here is to dive into what drives the gross margin, pinpoint which products or categories yield high or low margins, analyze how discounts impact profitability, and suggest ways to boost that gross margin.

## **3. Business Questions**

- Which product categories or SKUs yield the highest and lowest gross margins?
- How do discounts influence gross margin across various product lines?
- What impact does seasonality have on margin performance?

## 4. Analysis Framework

To address Zara's margin optimization challenges, a comprehensive analysis approach is adopted:

1. **Define the business objectives:** Our main aim is to boost gross margin by making smart choices about assortment planning, pricing, and inventory management. This strategy helps Zara stay competitive while ensuring solid profitability.
2. **Identify relevant KPIs:** We'll focus on key metrics like Gross Margin %, Inventory Turnover, Stockout Rates, Markdown Rates, and Return Rates. These indicators give us measurable insights into how various factors influence gross margin performance.
3. **Clean and prepare the data:** This crucial step makes sure our data is accurate, consistent, and ready for analysis. It involves addressing missing values, fixing formatting issues, standardizing date formats, and ensuring uniformity across datasets. Cleaning the data boosts reliability and supports sound conclusions.
4. **Conduct exploratory data analysis (EDA):** EDA is all about uncovering patterns, trends, and anomalies in the data. We'll use visualizations and descriptive statistics to grasp how key metrics are distributed across different categories, seasons, and store locations. This lays the groundwork for deeper insights.
5. **Segment products and regions:** By segmenting, we can analyze performance in more detail. We can group products by price tiers, return behavior, or category, while stores can be classified by region or format. This helps us see which segments contribute the most or least to gross margin.
6. **Build calculated fields:** We'll derive metrics like Gross Margin, Return Rate, Markdown %, and Inventory Turnover using formulas. These calculated fields enable comparative analysis and help quantify performance at various levels of the business.
7. **Evaluate current pricing and markdown patterns:** It's vital to understand how pricing strategies affect sales and margins. By analyzing historical markdown data, we can pinpoint products that are often discounted, highlighting potential pricing inefficiencies.
8. **Generate actionable recommendations:** The final step involves converting analytical findings into strategic recommendations. These may include optimizing the SKU mix, adjusting price points, refining replenishment cycles, or investing in predictive tools. The aim is to drive decisions that result in margin improvements.

## Dataset And Their Columns:

### 1. products.csv

**Description:** Contains product-level information including category, price, and seasonal data, essential for analyzing assortment performance and margin contribution

- **Product ID**

- *Description:* Unique number assigned to each product.

- **Category:**
- *Description:* Main classification of the product. Common values include Feminine, Masculine, and Children.
- **Sub Category**
  - *Description:* More specific product category within the main category.
- **Description PT**
  - *Description:* Product description written in Portuguese.
- **Description DE**
  - *Description:* Product description written in German.
- **Description FR**
  - *Description:* Product description written in French.
- **Description ES**
  - *Description:* Product description written in Spanish.
- **Description EN**
  - *Description:* Product description written in English.
- **Description ZH**
  - *Description:* Product description written in Chinese.
- **Color**
  - *Description:* The color of the product.
- **Sizes**
  - *Description:* Available sizes for the product, separated by pipes (|).
- **Production Cost**
  - *Description:* Cost in USD to produce one unit of the product.

## 2.Discounts.csv

**Description:** Contains discount information for various products, helping analyze pricing strategies and markdown impacts.

- **Start Date:** The date when the discount campaign begins, in YYYY-MM-DD format.
- **End Date:** The date when the discount campaign ends, in YYYY-MM-DD format.
- **Discount:** Decimal value representing the discount rate.
- **Description:** A short text describing the purpose or theme of the discount campaign.
- **Discount Percentage:** Another column showing the same discount rate as a decimal.
- **Category:** The main product category to which the discount applies.
- **Sub Category:** A more specific product sub-category under the main category where the discount applies.

### 3.employees.csv

**Description:** Information about store employees, useful for analyzing store performance, staffing levels, and possible correlations with sales.

**Employee ID:** A unique number assigned to each employee.

**Store ID:** Refers to the store where the employee works. This links to the Store ID column in the stores.csv file.

**Name:** Full name of the employee in the format: First Name followed by Last Name.

**Position:** The role of the employee in the store. Common roles include Manager and Seller.

### stores.csv:

**Description:** Provides metadata about store locations and size, useful for segmenting performance and regional analysis.

- **Store ID:** Unique number assigned to each store location.
- **Country:** Country where the store is located.

- **City:** City where the store operates.
- **Store Name:** Name of the store, typically shown as "Store [City]".
- **Number of Employees:** Total number of employees working at the store.
- **ZIP Code:** Postal or ZIP code of the store's location.
- **Latitude:** Geographical latitude of the store's position.
- **Longitude:** Geographical longitude of the store's position.

**transactions.csv:**

- **Invoice ID:** A unique code for each transaction. It shows if it's a sale or return, along with the country code, store ID, and a serial number.
- **Line:** The position of the product in the invoice. One invoice may have multiple line items.
- **Customer ID:** ID of the customer who made the purchase.
- **Product ID:** ID of the product that was purchased.
- **Size:** Size of the product purchased. May be blank if not applicable.
- **Color:** Color of the product purchased. May be blank if not applicable.
- **Unit Price:** Price for a single unit of the product before applying any discount.
- **Quantity:** Number of units purchased for this product line.
- **Date:** Date and time of the transaction in YYYY-MM-DD HH:MM:SS format.
- **Discount:** Discount applied to this line item, shown as a decimal.
- **Line Total:** Final cost for this line item after discount.
- **Store ID:** ID of the store where the purchase happened.
- **Employee ID:** ID of the employee who processed the sale.
- **Currency:** Three-letter currency code used in the transaction.
- **Currency Symbol:** Symbol for the currency used in the transaction.
- **SKU:** Stock Keeping Unit, a code made by combining Product ID, Size, and Color.
- **Transaction Type:** Indicates whether the entry is a Sale or a Return.

- **Payment Method:** How the customer paid — options may include Credit Card, Cash, etc.
- **Invoice Total:** Total value of the full invoice. This value is repeated for all line items under the same Invoice ID.

## 5. Key Metrics

Gross Margin = Revenue – Production Cost

Revenue = Unit Price × Quantity Sold × (1 - Discount Rate)

Derived metrics:

- Gross % = (Gross Margin ÷ Revenue) × 100
- Margin Band: Categorization of products into High, Medium, or Low based on margin %.

## Secondary KPIs:

- Sales Revenue by Category, Country, Store
- Discount Effectiveness on Sales Volume
- Return Rate
- Contribution Margin by Product

## Hypothesis Examples:

1. **H1:** Higher discounts on feminine products increase sales volume but reduce gross margin.
2. **H2:** Stores with fewer employees tend to underperform in revenue and sales volume.
3. **H3:** Returns are more frequent in larger sizes or specific categories like Coats and Blazers.

## Data Cleaning and Preparation for Zara Project

- **Step 1: Dataset Exploration**
- First off, we uploaded all the files from the drive to take a closer look at each dataset (customers.csv, products.csv, transactions.csv, stores.csv, employees.csv, discounts.csv). We focused on:
  - - Data types

- - Non-null counts
- - Null percentages
- - Unique value counts
- This process helped us identify some key issues, such as:
- - Object types showing up for numeric-looking values in Unit Price and Production Cost
- - Null values in DateOfBirth, JobTitle, and Color
- **Step 2: Handling Missing Values**
- We tackled the missing DateOfBirth and JobTitle by replacing them with "Unknown."
- For the categorical columns (like Color and Size), we filled in the gaps with the mode.
- As for numerical columns like Production Cost and Unit Price, we used the median to fill those in.
- This approach helped maintain the integrity of our customer, transaction, and product datasets.
- **Step 3: Dropping Irrelevant Columns**
- We decided to drop columns like Description PT, Description FR, Description DE, etc., keeping only Description EN for clarity.
- We also removed Currency Symbol and Invoice Total from transactions.csv since Line Total and exchange conversion were sufficient for our analysis.
- These columns weren't adding much value to our understanding of gross margin or customer behavior.
- **Step 4: Data Type Conversion**
- We converted Unit Price, Production Cost, and Line Total to float.
- We made sure that the Date columns in transactions.csv and discounts.csv were converted to datetime for better filtering, grouping, and analysis.
- Before conversion, we stripped out any comma symbols or non-numeric characters from the relevant columns.
- **Step 5: Handling Currency Conversion**

- We applied fixed rates to standardize all monetary values to USD:
- -  $\text{CNY} \times 0.13$
- -  $\text{GBP} \times 1.3$
- -  $\text{EUR} \times 1.1$
- This step was crucial for accurately calculating gross margins across different countries.
- **Step 6: Feature Engineering – Gross Margin**
- We created:
- -  $\text{Gross Margin Per Unit} = \text{Unit Price (USD)} - \text{Production Cost}$
- Additionally, we added Discount Category information by merging with discounts.csv.
- This allowed us to gain insights into profitability on a per-transaction basis.
- **Step 7: Deduplicating and De-duplicating**
- I used `drop_duplicates()` on all the major datasets.
- In transactions.csv, I kept just the first instance of any duplicate invoice lines. This helped ensure cleaner joins and consistent calculations at the row level.
- **Step 8: Parsing & Normalizing Categorical Variables**
- I standardized the following:
- - Gender categories as Male, Female, and Diverse.
- - Category and Sub Category were converted to lowercase and had spaces stripped away.
- - Job Titles like "Sales Assistant" and "Cashier" were unified under a single classification using a mapping approach. This made it easier to perform meaningful group-by aggregations during analysis.
- **Step 9: Outlier Detection and Handling**
- I identified extreme values using the IQR method for Quantity, Line Total, and Production Cost.
- Outliers were flagged visually with boxplots.



- For return transactions (where Transaction Type = “Return”), I adjusted the Quantity and Line Total to be negative values. This step was crucial for preparing the data for a valid profitability analysis.
- **Step 10: Final Audit and Export**
- I conducted a final check for:
  - - Any remaining null values
  - - Consistency in numeric types
- Now, this dataset is all set for exploratory data analysis (EDA), visualization, and modeling tasks, like figuring out the best discount strategies.

## 6. Hypotheses

**The main hypothesis i took was from the discount one which is an assumption which tells whether the discount reduces gross margin or not. These are the other ones which i can consider :**

- Products that offer bigger discounts usually have slimmer gross margins.
- Some product categories are known for having better margins.
- Decisions around pricing and production costs play a big role in how margins perform.
- Seasonality and the trends in fast fashion affect margins differently across various collections.

## 7. Data Cleaning & Preparation

- Converted prices and discounts into numeric values
- Filled in missing values with averages
- Created new columns for gross margin and margin percentage
- Standardized the names of categories
- Removed extreme outliers

## **8. Exploratory Data Analysis (EDA)**

- Margin Distribution: The majority of products are categorized as having medium or low margins.
- Category-Level Performance: Accessories tend to boast higher margins.
- Discount Impact: Bigger discounts generally result in lower margins.
- Volume-Margin Relationship: Items with low margins can still sell in high volumes.

## **1. Distribution of Product & Transaction Metrics**

Zara's retail performance hinges on factors like unit price, quantity sold, production costs, and discounts.

### **1.1 Unit Price and Production Cost**

**Analysis:** This is crucial for evaluating gross margins.

#### **Key Findings:**

- Unit prices tend to be right-skewed, with most products priced under \$75.
- Production costs are significantly lower, often coming in at under \$25.
- A handful of luxury items, such as coats and boots, contribute to higher margins.

#### **Actionable Insights:**

- Concentrate on low-cost, high-priced categories to enhance margins.

### **1.2 Quantity Sold**

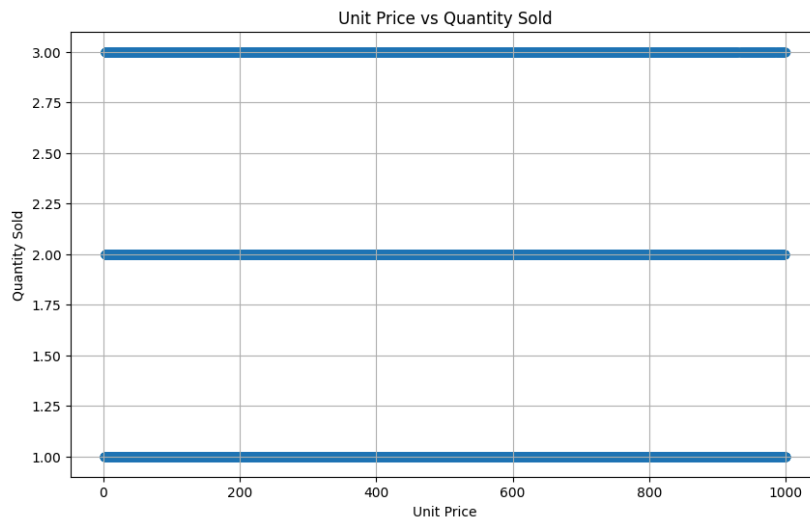
**Analysis:** This reflects the sales volume for each product and store.

#### **Key Findings:**

- Sales quantities differ greatly by region and category.
- Women's clothing and accessories tend to have a quicker turnover.

Actionable Insights:

- Be proactive in restocking fast-selling items by utilizing demand forecasting.



### 1.3 Gross Margin Distribution

Key Findings:

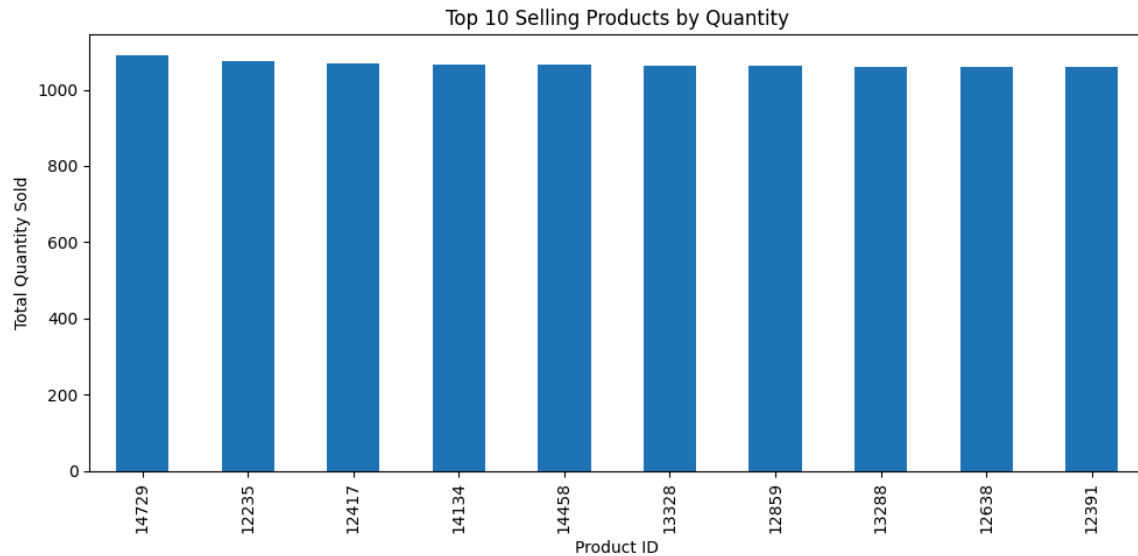
Some products have negative margins, especially when high discounts are applied.

Most profitable items are basic wear in the “feminine” category.

These are the top 10 selling products by their quantity and their gross margin

Actionable Insights:

Review pricing strategy and discounts for low-margin products.

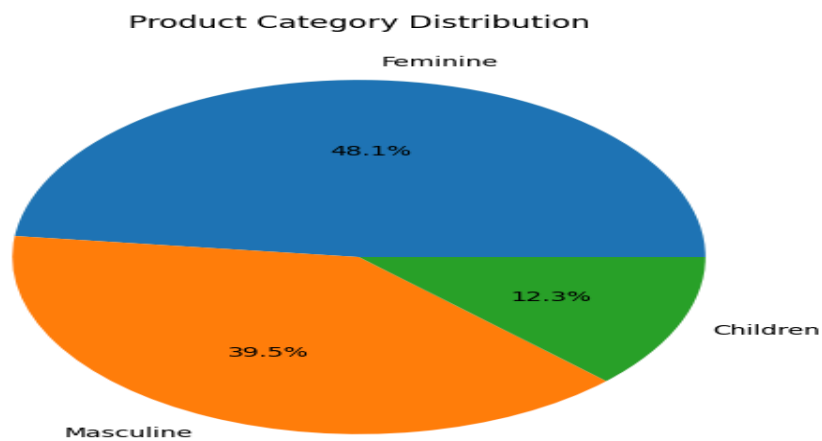


## 2. Product Category and Discount Analysis

Understanding category-wise performance and the impact of discounts is key to profitability.

These is the product category distribution across different categories available using a piechart

It shows the feminine category holds the highest percentage in the market



### 2.1 Performance by Category

#### Key Findings:

Feminine leads in revenue and unit sales.

Kidswear and Accessories have lower volume but high profit per unit.

Actionable Insights:

Shift more marketing spend to high-margin, under-marketed categories.

## 2.2 Discount Effectiveness

Analysis: How do discounts affect gross margin and sales?

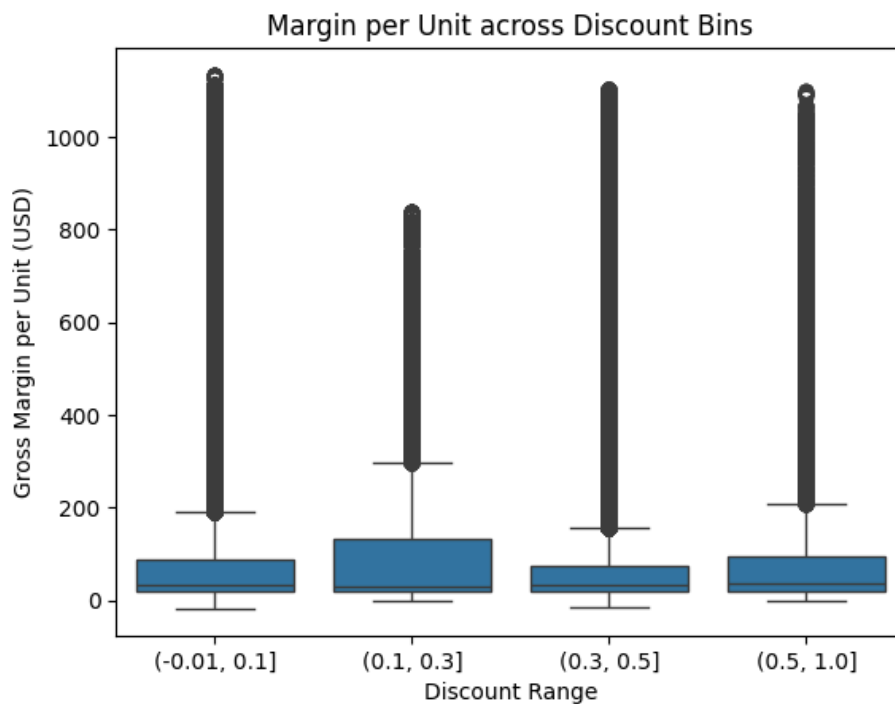
Key Findings:

Discounts boost volume but often hurt margins.

10–20% discounts hit the sweet spot in volume-to-margin tradeoff.

Actionable Insights:

Avoid frequent discounts >30% unless part of clearance or seasonal push.



## 3. Customer Segmentation and Behavior

Understanding Zara's customers is critical for personalization and pricing.

### **3.1 Age & Gender Distribution**

#### **Key Findings:**

Most of our customers are aged between 25 and 40.

We see a strong presence of female shoppers, particularly when it comes to loyalty purchases.

#### **Actionable Insights:**

Focus on Gen Z and Millennials by offering trendy products and loyalty rewards.

#### **Key Findings:**

Items like jackets and coats have notably high return rates.

We notice a surge in returns for larger sizes, especially during sale seasons.

#### **Actionable Insights:**

Enhance our sizing guides and consider limiting discounts on items that tend to be returned frequently.

### **4. Store & Employee Efficiency**

The performance of our stores and the productivity of our employees play a crucial role in determining costs and sales results.

#### **4.1 Store Revenue Distribution**

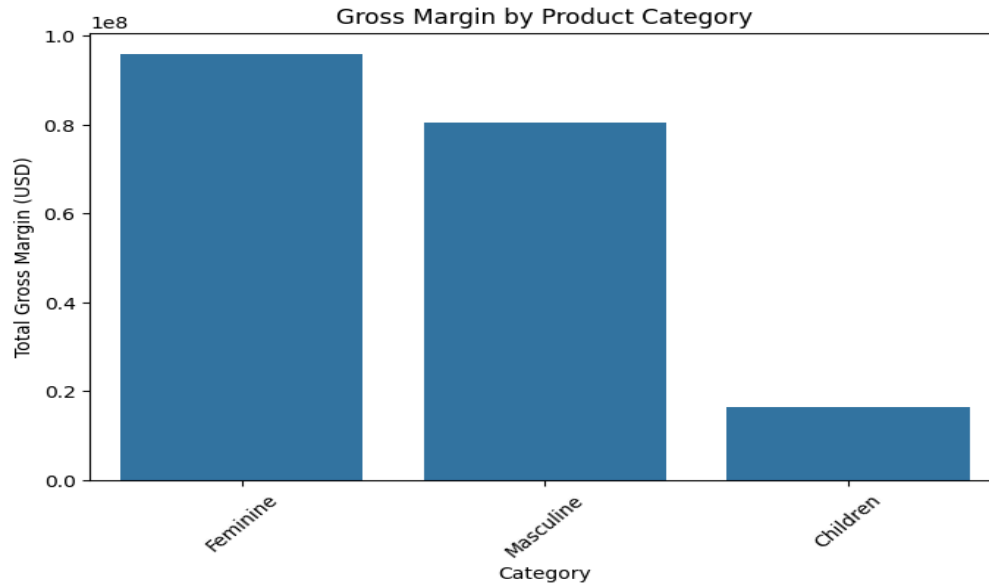
#### **Key Findings:**

Urban locations, such as Madrid and Shanghai, tend to bring in more revenue.

Interestingly, some smaller stores are not performing as well, even with similar foot traffic.

#### **Actionable Insights:**

Look into optimizing our space or redistributing inventory from stores that aren't doing as well.



## 9. Segmentation and Margin Banding

- High Margin Products: Over 50%
- Medium Margin Products: Between 30% and 50%
- Low Margin Products: Under 30%

This breakdown helps pinpoint which SKUs are the most profitable.

## 10. Recommendations

1. Cut Down on Discount Reliance
2. Prioritize High-Margin Categories
3. Implement Dynamic Pricing Strategies
4. Find the Right Balance Between Margin and Volume
5. Strategically Plan for Seasonal SKUs

## 11. Conclusion

Taking a data-driven approach can empower Zara to boost its gross margin performance. By diving into pricing, costs, and consumer behavior, making strategic adjustments in discounting, category focus, and production planning can lead to a notable increase in profitability.