



# Customer Shopping Behavior Analysis

## 1. Project Overview

This project analyzes customer shopping behavior using transactional data from 3,900 purchases across various product categories. The goal is to uncover insights into spending patterns, customer segments, product preferences, and subscription behavior to guide strategic business decisions.

## 2. Dataset Summary

- Rows: 3,900
- Columns: 18
- Key Features:
  - Customer demographics (Age, Gender, Location, Subscription Status)
  - Purchase details (Item Purchased, Category, Purchase Amount, Season, Size, Color)
  - Shopping behavior (Discount Applied, Promo Code Used, Previous Purchases, Frequency of Purchases, Review Rating, Shipping Type)
- Missing Data: 37 values in Review Rating column

### 3. Exploratory Data Analysis using Python

We began with data preparation and cleaning in Python:

#### Data Loading

Imported the dataset using pandas.

#### Initial Exploration

Used df.info() to check structure and .describe() for summary statistics.

#### Missing Data Handling

Checked for null values and imputed missing values in the Review Rating column using the median rating of each product category.

#### Column Standardization

Renamed columns to **snake case** for better readability and documentation.

### 3. Exploratory Data Analysis using Python (cont.)

#### Feature Engineering

- Created **age\_group** column by binning customer ages.
- Created **purchase\_frequency\_days** column from purchase data.

#### Data Consistency Check

Verified if discount\_applied and promo\_code\_used were redundant; dropped promo\_code\_used.

#### Database Integration

Connected Python script to PostgreSQL and loaded the cleaned DataFrame into the database for SQL analysis.

## 4. Data Analysis using SQL (Business Transactions)

We performed structured analysis in MySQL to answer key business questions:

### 1 Revenue by Gender

Compared total revenue generated by male vs. female customers.

gender text	revenue numeric
Female	75191
Male	157890

### 2 High-Spending Discount Users

Identified customers who used discounts but still spent above the average purchase amount.

customer_id bigint	purchase_amount bigint
3	64
4	90
7	85
9	97
12	68
13	72
16	81
20	90
22	62

Total rows: 839

## 4. Data Analysis using SQL (Business Transactions)

### 1 Top 5 Products by Rating

Found products with the highest average review ratings.

item_purchased	Average Product Rating
Gloves	3.86
Sandals	3.84
Boots	3.82
Hat	3.80
Skirt	3.78

### 2 Shipping Type Comparison

Compared average purchase amounts between Standard and Express shipping.

shipping_type	round
Standard	58.46
Express	60.48

### 3 Subscribers vs. Non-Subscribers

Compared average spend and total revenue across subscription status.

subscription_status	total_customer_count	average_spending	total_revenue
Yes	1053	59.49	62645.00
No	2847	59.87	170436.00

## 4. Data Analysis using SQL (Business Transactions)

### 1 Discount-Dependent Products

Identified 5 products with the highest percentage of discounted purchases.

item_purchased text	discount_rate numeric
Hat	50.00
Sneakers	49.66
Coat	49.07
Sweater	48.17
Pants	47.37

### 2 Customer Segmentation

Classified customers into New, Returning, and Loyal segments based on purchase history.

customer_segment text	Number of Customers bigint
Loyal	3116
New	83
Returning	701

## 4. Data Analysis using SQL (Business Transactions)

### 1 Top 3 Products per Category

Listed the most purchased products within each category.

item_rank bigint	category text	item_purchased text	total_orders bigint
1	Accessories	Jewelry	171
2	Accessories	Sunglasses	161
3	Accessories	Belt	161
1	Clothing	Blouse	171
2	Clothing	Pants	171
3	Clothing	Shirt	169
1	Footwear	Sandals	160
2	Footwear	Shoes	150
3	Footwear	Sneakers	145
1	Outerwear	Jacket	163
2	Outerwear	Coat	

## 4. Data Analysis using SQL (Business Transactions)

### 1 Repeat Buyers & Subscriptions

Checked whether customers with >5 purchases are more likely to subscribe.

subscription_status text	repeat_buyers bigint
No	2518
Yes	958

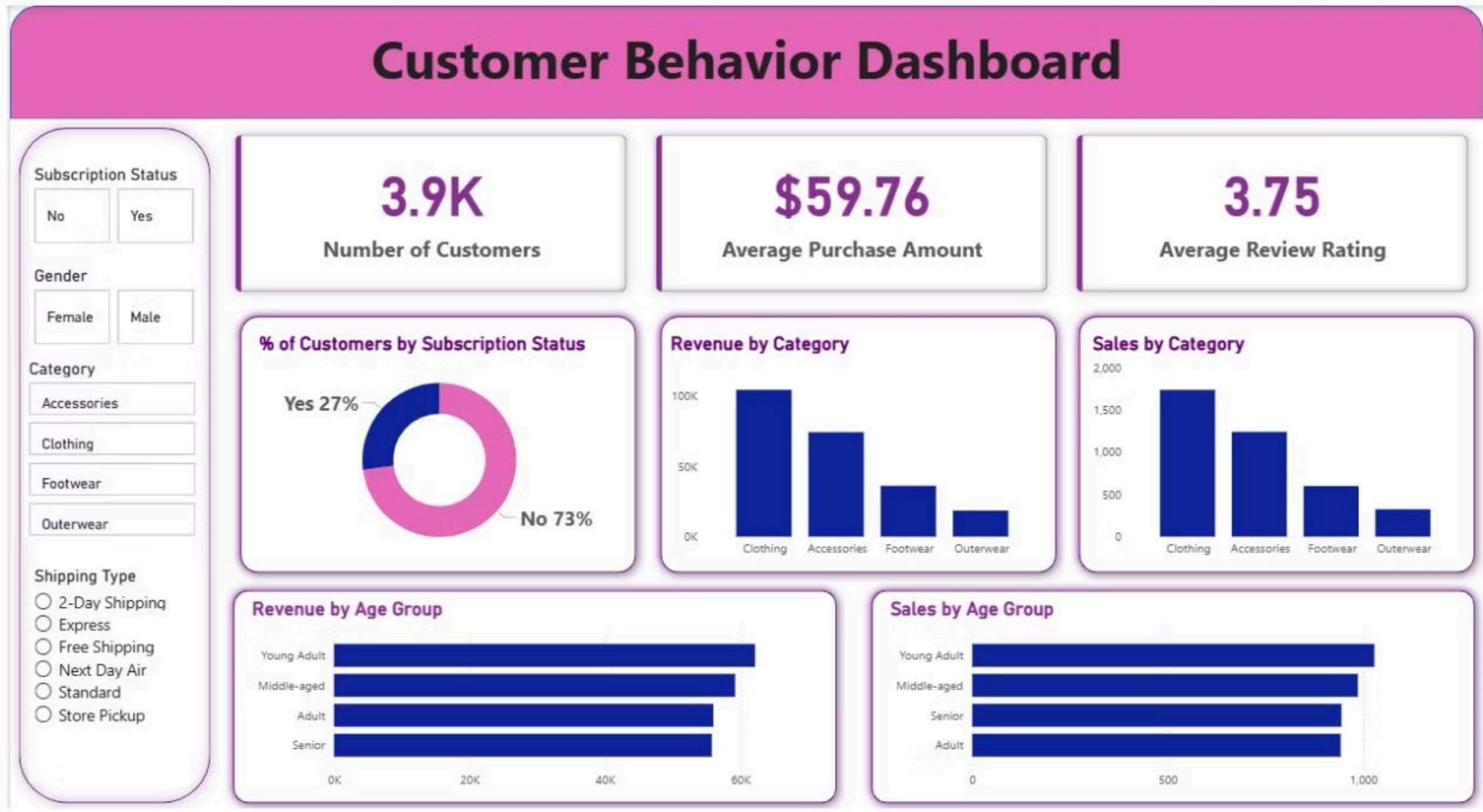
### 2 Revenue by Age Group

Calculated total revenue contribution of each age group.

age_group text	total_revenue numeric
Young Adult	62143
Middle-aged	59197
Adult	55978
Senior	55763

## 5. Dashboard in Power BI

Finally, we built an interactive dashboard in **Power BI** to present insights visually.



## 6. Business Recommendations

### → Boost Subscriptions

Promote exclusive benefits for subscribers.

### → Customer Loyalty Programs

Reward repeat buyers to move them into the "Loyal" segment.

### → Review Discount Policy

Balance sales boosts with margin control.

### → Product Positioning

Highlight top-rated and best-selling products in campaigns.

### → Targeted Marketing

Focus efforts on high-revenue age groups and express-shipping users.

THANK U!