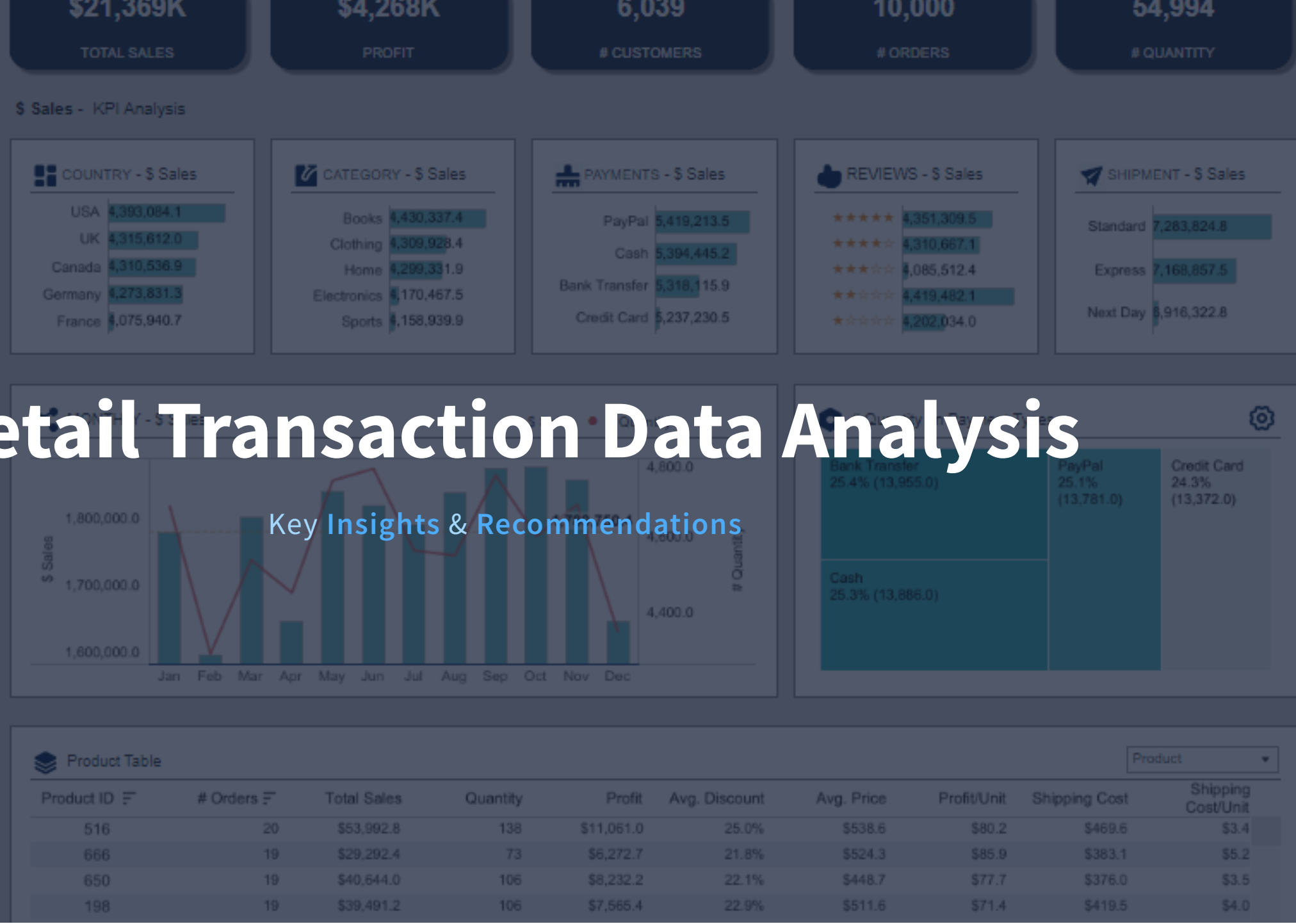




# Retail Transaction Data Analysis

## Key Insights & Recommendations






# Analysis Overview




## Data Sources

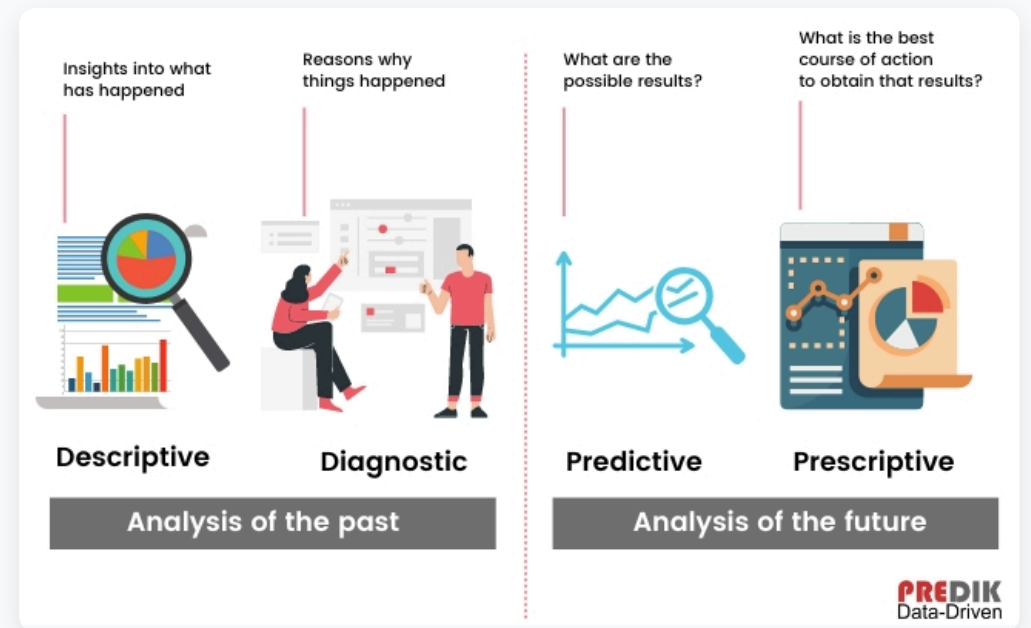
-  Retail transaction data
-  Customer response data

## Analysis Approach

-  Data preprocessing & cleaning
-  Exploratory data analysis
-  Advanced analytics (RFM, Time Series)

## Key Objectives

-  Understand **customer behavior**
-  Identify sales **trends**
-  Provide actionable insights



# Data Preprocessing & Cleaning

## 📡 Data Integration

- 🔗 Merged transaction and response datasets
- 🔑 Joined on **customer\_id** field

## 🏠 Data Cleaning

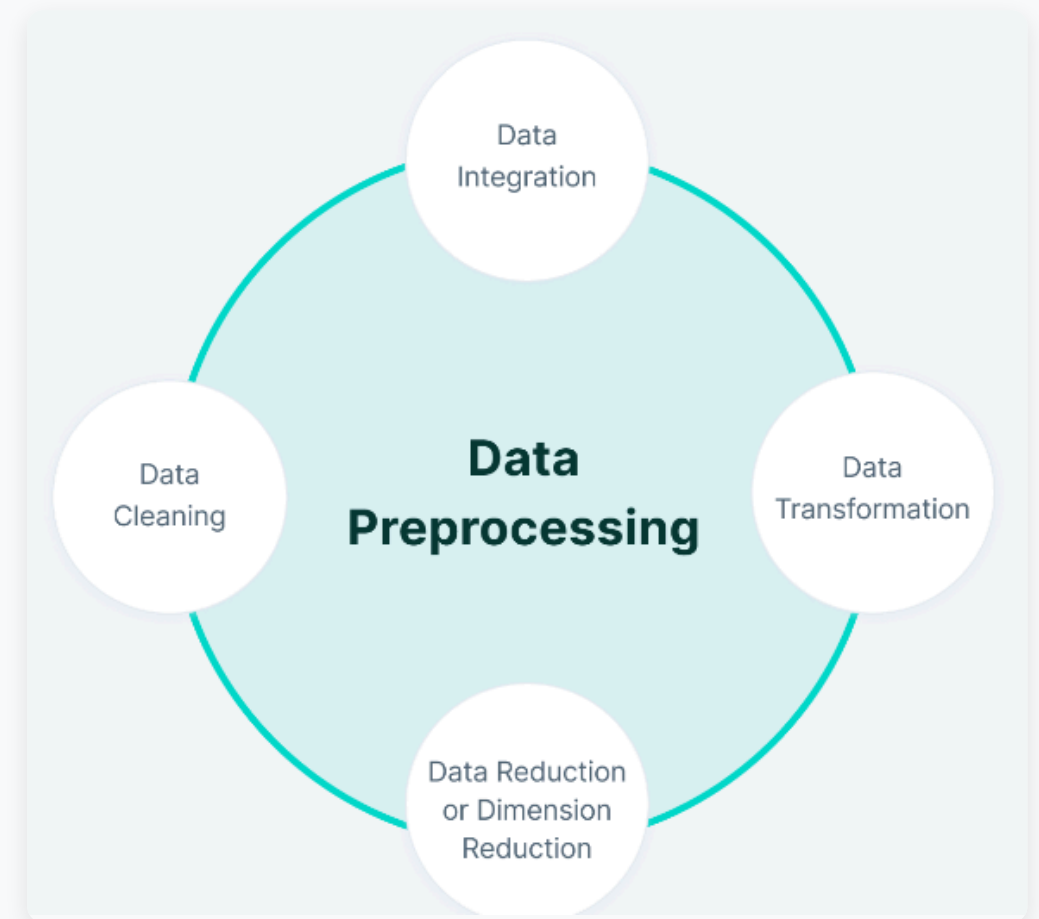
- 🔄 Handled missing values in response column
- ✅ 31 null values filled with **0**

## 🔄 Data Transformation

- 📅 Converted trans\_date to **datetime**
- 123 🔄 Converted response to **int64**

## 📊 Outlier Detection

- 📊 Applied **z-score** method
- ✅ No significant outliers detected



# Key Findings

## Top Months

↗ **August** , **October** , and **January** have the highest transaction amounts

\$ August leads with **\$726,921** in total sales

## Top Customers

★ **CS4424** has the highest order count (39)

🛒 Same customer leads with **\$2,933** in total sales

## ↗ Yearly Trends

↗ **2013** had the highest total transactions

\$ Total sales of **\$2,137,368** in 2013

## Data Distribution

✓ No significant outliers detected

○ Transaction amounts and response values are **normally distributed**



BI for Retail isn't a **Luxury**. It's a **Necessity**

# Monthly Transaction Trends


## Analysis Period

 January 2011 to March 2015

~ 50+ months of transaction data analyzed

## Seasonal Patterns

~ Clear **seasonal variations** in transaction amounts


 Consistent patterns across multiple years

## Peak Periods

★ **August** , **October** , and **January** show highest volumes

~ Monthly peaks range from **\$724,107** to **\$726,921**

## Trend Analysis

 Overall volume shows **slight variations**


→ No dramatic growth or decline observed




Time Series Graph Examples: Analyzing Time-Based Data

# RFM Analysis & Customer Segments

## RFM Model

 **Recency**: Most recent purchase date

 **Frequency**: Number of purchases

 **Monetary**: Total transaction value

## Segment Definitions

★ **P0** : 2015+ recent, 15+ orders, \$1000+ sales

👤 **P1** : 2014 recent, 10-15 orders, \$500-1000 sales

👥 **P2** : All other customers

## Segment Distribution

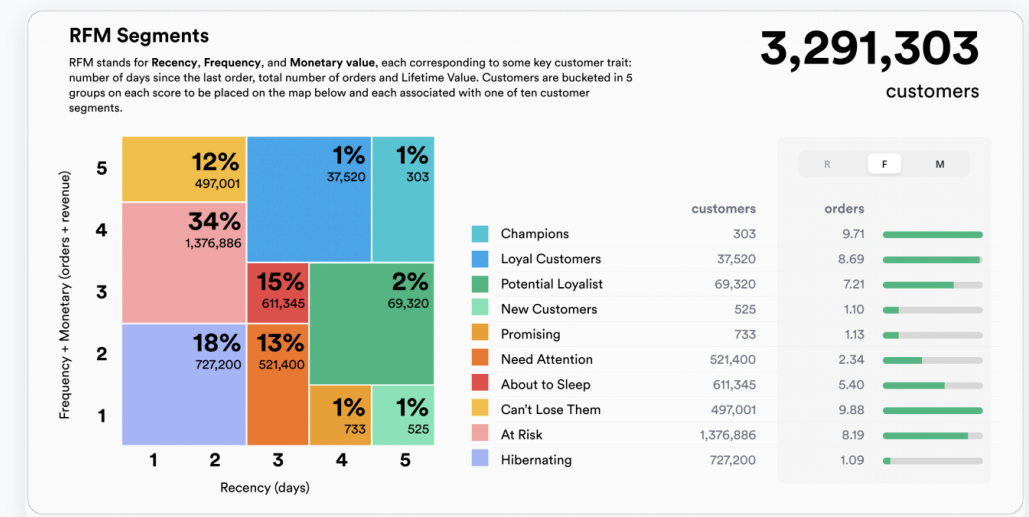
🧠 Three **distinct customer groups** identified

🔍 Clear separation in purchasing behavior

## Strategic Value

📢 Different segments require **different marketing approaches**


📈 Opportunity to convert P2 to higher segments




# Customer Churn Analysis

## Churn Rate

 **111,158** churned customers (response=0)

 **13,842** active customers (response=1)


## Churn Percentage

 Approximately **89%** of customers have churned

 Only **11%** of customers remain active


## Top Customer Analysis

 Top 5 customers show **consistent purchasing patterns**

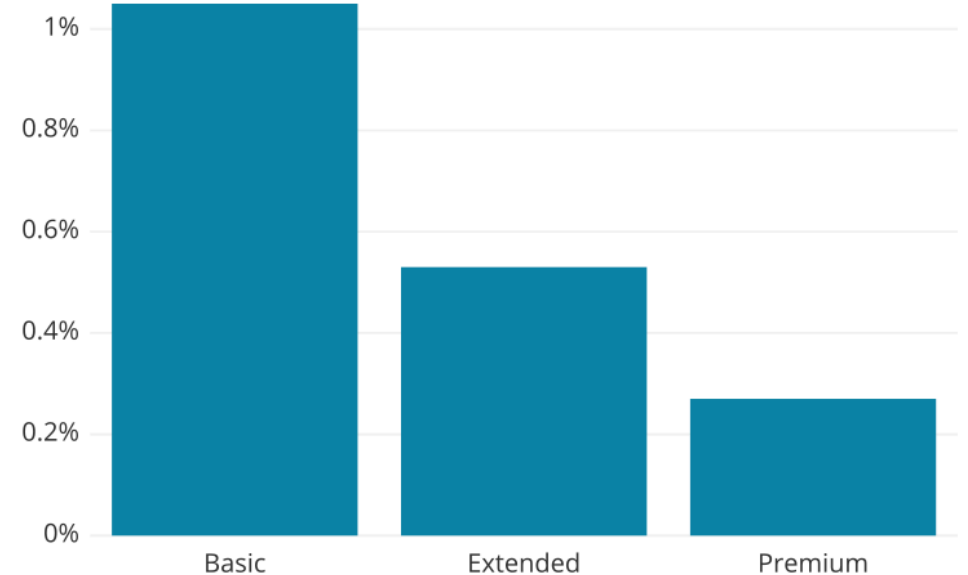
 Regular transaction intervals throughout the year

## Retention Opportunity

 Significant opportunity to **re-engage** churned customers

 Targeted campaigns could reactivate dormant segments

Churn Rate by Account Type



# Strategic Recommendations

## Seasonal Marketing

- 📢 Focus campaigns on **August** , **October** , and **January**
- 📈 Align promotions with peak transaction periods

## Customer Retention

- 🔄 Implement **re-engagement strategies** for 89% churned customers
- ✉ Targeted offers based on previous purchase history

## VIP Program

- 💬 Develop exclusive programs for **high-value customers**
- 💎 Prioritize PO segment with premium benefits

## Data-Driven Decisions

- 🕒 Continue **regular analysis** to identify trends
- 📈 Track segment performance and adjust strategies

