# **Project Portfolio & Case Studies**

## 1. E-commerce Data Analysis

**GitHub Repository: E-commerce Data Analysis** 

I have Worked on an E-commerce project dealing with large data sets. As I can not present those work samples I have recreated an E-commerce project that can demonstrate similar Insights and visualization. I have collected this dataset from <a href="https://www.kaggle.com/">https://www.kaggle.com/</a>

## **Objective:**

Analyzing customer behavior and transaction patterns is crucial for e-commerce businesses. My goal in this project was to extract meaningful insights from customer purchase data, identify key trends, and develop predictive models to assist in decision-making. The analysis covers customer segmentation, revenue impact of discounts, and forecasting customer spending behavior.

### **Tools Used:**

- Python (Pandas, NumPy, Matplotlib, Seaborn, Scikit-learn)
- Jupyter Notebooks
- SQL for data querying

## **Methodologies Applied:**

## 1. Data Cleaning

- The dataset contained inconsistencies, missing values, and redundant information, which I addressed through preprocessing.
- Date formats and categorical variables were standardized to ensure uniformity across the dataset.
- Duplicates were identified and removed to maintain data accuracy and consistency.
- Reference: 01\_data\_cleaning.ipynb

### 2. Exploratory Data Analysis (EDA)

- I conducted an in-depth analysis of customer demographics, purchasing trends, and product preferences.
- Visualizations helped me uncover trends in sales, discount effectiveness, and customer retention.
- I examined how discount strategies influenced customer purchases to optimize future marketing campaigns.
- Reference: 02\_eda.ipynb

### 3. Feature Engineering

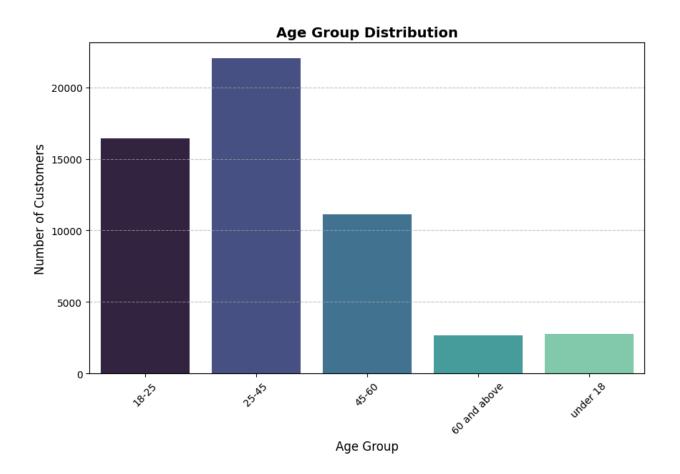
- I created new features such as <u>Customer Lifetime Value (CLV)</u>, <u>Recency</u>, <u>Frequency</u>, and <u>Monetary (RFM)</u> metrics to strengthen predictive modeling.
- Customer Lifetime Value (CLV): This metric estimates the total revenue a business can expect from a customer over their relationship duration. It helps identify high-value customers and optimize marketing efforts.
- Recency: Represents the time since a customer's last purchase, useful for predicting future purchase likelihood.
- Frequency: Measures how often a customer makes purchases, indicating their engagement level.
- Monetary (RFM) Analysis: Determines spending behavior, classifying customers into different value groups.
- Reference: 03 feature engineering.ipynb

## 4. Model Training

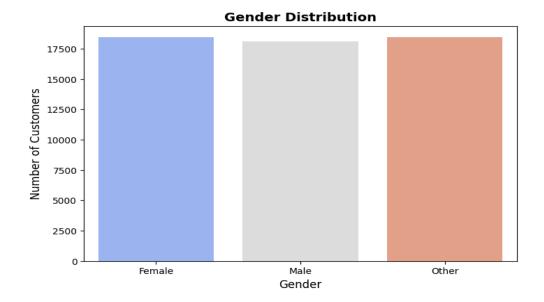
- I trained a Random Forest Classifier to predict customer churn based on CLV, Recency, Frequency, and Engagement Level.
- The model was trained using a well-balanced dataset after performing label encoding on categorical variables like Engagement Level.
- Feature Scaling: I applied StandardScaler to normalize numerical features for improved model performance.
- Performance Evaluation: The model was assessed using key metrics like Classification
  Report, Confusion Matrix, and ROC-AUC Score to ensure accuracy and reliability.
- Visualizations: Histograms were generated to compare training and testing data distributions for features like CLV, Recency, and Frequency.
- Reference: 04\_model\_training.ipynb

## **Findings from Visualizations:**

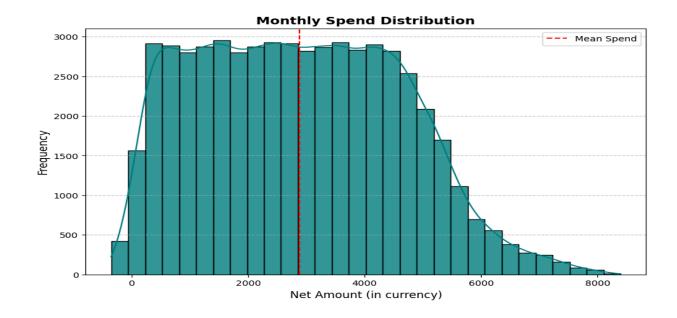
→ Customer Demographics Analysis:



→ The majority of customers fall into the 25-45 age group, with a significant portion being repeat buyers.



- → Gender distribution shows a slight skew toward female customers.
- → Discount Impact Analysis:
  - ◆ While discounts increase sales volume, they do not always lead to higher profit margins.
  - ◆ Returning customers tend to purchase at regular prices rather than relying on discounts.
- → Revenue Contribution by Customer Segments:
  - ♦ High-value customers contribute disproportionately to total revenue.



♦ Identifying these customers allows businesses to tailor personalized marketing efforts.

## **Outcomes:**

- I successfully identified high-value customers using CLV analysis, aiding in targeted marketing.
- Discount strategies were evaluated for their effectiveness in boosting sales while maintaining profitability.
- The predictive model provides insights into expected future sales based on past purchase behaviors.

## 2. Cricket Analysis - SQL-Based Data Processing

**GitHub Repository:** Cricket Analysis

## **Objective:**

This project is centered on structuring cricket match data using SQL, focusing on designing efficient queries, stored procedures, and views to streamline data retrieval and analysis.

### **Tools Used:**

- SQL (MySQL, PostgreSQL)
- Stored Procedures & Views
- Data Aggregation Functions

## **Methodologies Applied:**

## 1. SQL Query Development

- I wrote SQL queries to extract match statistics, such as player averages, team performance, and match outcomes.
- SQL views were created to simplify structured data retrieval for further analysis.
- 2. Stored Procedures Implementation
  - Automated stored procedures were developed to calculate batting and bowling statistics efficiently.
  - I ensured modularity so that the procedures could be reused for future cricket datasets.
- 3. Optimized Data Retrieval
  - Indexing and query optimization techniques were implemented to enhance database performance.
  - The structured approach allows for handling large volumes of match data efficiently.

### **Outcomes:**

- I developed efficient SQL queries and stored procedures for structured cricket match data analysis.
- Database performance was optimized, ensuring faster query execution.
- The SQL-based solution provides reusable components for similar future projects.

# **Code Samples & Notebooks**

- E-commerce Data Analysis: Jupyter notebooks document the entire workflow, from data cleaning to model building.
- Cricket Analysis: SQL scripts demonstrate structured query design and stored procedure implementation.

## **Data Visualizations & Dashboards**

### **E-commerce Project:**

- Customer segmentation and purchase trends visualized using bar charts and scatter plots.
- The effectiveness of discounts displayed through comparative revenue charts
- Sales predictions plotted against historical data for validation.

# **Documentation & Reports**

- E-commerce Analysis: Markdown explanations are included in Jupyter notebooks to provide clear context on each step.
- Cricket Analysis: SQL documentation outlines query logic, stored procedures, and optimization strategies.

# **Data Samples & Management**

Datasets Used:

dataset.csv: Original e-commerce transaction data.

cleaned\_dataset.csv: Processed dataset after data cleaning.

engineered\_dataset.csv: Feature-enriched dataset for modeling.

- > Data Quality Management:
  - I ensured data integrity through validation checks and preprocessing steps.
  - Outliers and missing values were carefully handled to maintain data consistency.
  - Query performance was optimized for handling large datasets efficiently.

## **GitHub Repositories:**

- 1. E-commerce Data Analysis
- 2. Cricket Analysis