# Cyclistic Data Analysis Case Study

### Introduction

Welcome to the Cyclistic bike-share analysis case study conducted by the marketing analytics team at Cyclistic, a leading bike-share program in Chicago. This case study explores the behavior of casual riders and annual members to formulate effective strategies aimed at increasing annual memberships.

Cyclistic operates a bike-share program with a fleet of over 5,800 bicycles across 600 docking stations in Chicago. Since its inception in 2016, Cyclistic has provided inclusive transportation options, including reclining bikes and hand tricycles, catering to various rider needs.

## Methodology

The analysis will follow the standard data analysis process: Ask, Prepare, Process, Analyze, Share, and Act.

### 1. Ask Phase: Define the Objectives

Objective: Understand how annual members and casual riders use Cyclistic bikes differently.

**Business Task:** Identify the key differences in usage patterns between annual members and casual riders to inform marketing strategies and operational improvements.

**Key Deliverables:** - A detailed report highlighting the differences in usage patterns between annual members and casual riders. - Visualizations that clearly depict these differences. - Actionable recommendations based on the insights derived from the data analysis.

**Stakeholders:** - Marketing Team: Interested in understanding user behavior to tailor marketing campaigns. - Operations Team: Looking to optimize bike distribution and availability. - Executive Team: Focused on strategic decisions to increase membership and improve customer satisfaction.

### Prepare Phase: Data Preparation

Used Cyclistic's historical trip data to analyze and identify trends.

Data has been made available by Motivate International Inc. under this license.

**Key Tasks:** - Downloaded data and store it appropriately. - Identify how it's organized. - Sort and filter the data. - Determine the credibility of the data.

**Deliverable:** A description of all data sources used.

### 3. Process Phase: Data Processing

#### Preprocessing

**Key Tasks:** - Check the data for errors. - Choose your tools. - Transform the data so you can work with it effectively. - Document the cleaning process.

**Deliverable:** Documentation of any cleaning or manipulation of data.

#### Standardizing both the datasets

- Data\_2019 Renamed Columns, Mutated Columns, Changed Datatyped & Dropped irrelevant columns
- Data\_2020 Renamed Columns, Done Feature Engeering, Dropped Irrelevant Columns
   & Reordered Columns
- Combined Datasets- Explored and Done Preprocessing required like Handling missing values, feature engineering, dropped unnecessary columns & checked apt. datatypes

### 4. Analyze Phase

#### **Key Tasks:**

- · Aggregate your data so it's useful and accessible.
- Organize and format your data.
- · Perform calculations.
- · Identify trends and relationships.

Deliverable: A summary of your analysis.

#### **Exploratory Data Analysis**

1. User Type Analysis

#### Count Plot for Casual and Member Users

**Insight**: Member users significantly outnumber members based on the graph.

#### 2. Station Analysis

#### **Top 10 Start Stations Insights:**

- Canal St & Adams St dominates: This station tops the list, indicating it's the most popular starting point for bike rides.
- Potential hotspots: Other stations appearing frequently (e.g., Clinton St & Washington Blvd, Clinton St & Madison St) could be hubs or areas with high demand for starting bike trips.

**Top 10 End Stations Insight:** - Canal St & Adams St reigns supreme: This station tops the list, making it the most popular ending point for bike rides. - Multiple destinations emerge: Other stations like Clinton St & Madison St and Clinton St & Washington Blvd frequently appear, suggesting they are also popular ending points.

#### 3. Behavioural Analysis Day of the Week Ride Counts by User Type Insights:

- Weekends for leisure, weekdays for members:
- · Casual riders spike on weekends, suggesting recreational use.
- Members, likely commuting, have a flatter distribution across the week.

 Weekdays see higher overall ridership: Though casual riders peak on weekends, the combined ridership is higher on weekdays, potentially due to consistent member usage.

#### 4. Temporal Analysis

#### Monthly Ride Counts by User Type Insights:

- March has the most ridership, followed by February and January.
- Membership growth drives overall ridership increase.
- Casual ridership also grew in March, but at a slower rate than memberships.

#### Time of Day Counts by User Type Insights:

#### Distinct Usage Peaks:

- Casual riders peak in the mornings and evenings, likely using the service for commuting and occasional leisure rides.
- Members show a strong afternoon peak, indicating consistent usage for commuting, possibly with additional morning activity.

#### Afternoon Dominance:

• Afternoons consistently see the highest overall ride volume, driven by both casual and member riders. This period stands out as the busiest time for the service.

#### Usage Patterns Reflecting Needs:

- Casual riders demonstrate flexibility with morning and evening usage, suggesting both commuting and leisure motivations.
- Members' concentrated afternoon usage aligns with commuter patterns, indicating a primary use case for work-related travel.

#### Ride Count by Hour and Day for Casual Users Insights:

- Daylight dominates: The heatmap is hotter during the day (6am-6pm) compared to nights, suggesting casual riders favor daytime trips.
- **Afternoon rush:** The brightest spot centers around afternoons (12-6pm), indicating peak usage for casual users.

#### Ride Counts by Hour and Day for Member Users Insights:

- **Morning & afternoon rides:** Members tend to ride more in the mornings and afternoons, possibly for commutes.
- Less night rides: Similar to casual users, member rides are lower at night.

#### 5. Usage Patterns Analysis

#### **Trip Duration Comparison Insights:**

- Two trip types emerge: A clear split exists between shorter member trips (around 13 minutes) and longer casual trips (around 83 minutes). This suggests potential commutes vs. leisure rides.
- Casual riders more varied: The wider spread in casual trip durations indicates more variation in their riding patterns compared to members.

#### **Overall Inferences**

User Segmentation: There is a clear distinction between member users, who primarily
use the service for regular commuting, and casual users, who tend to use it for leisure or
occasional purposes.

**Implication:** Tailoring marketing strategies and service offerings to cater to these distinct user behaviors can enhance user engagement and satisfaction.

 Seasonal Variability: Bike usage shows significant fluctuations across different seasons, with peak activity observed during summer months and reduced activity during winter.

**Implication:** Adjusting operational strategies, such as bike deployment and station management, based on seasonal demand patterns can optimize resource utilization and service reliability.

3. **Station Utilization:** Certain stations consistently experience high demand as starting or ending points for bike trips.

**Implication:** Allocating resources to maintain sufficient bike availability and docking spaces at these popular stations is critical to meeting user demand effectively.

4. **Trip Characteristics:** Member trips are typically shorter and more frequent, reflecting commuting patterns, while casual users tend to have longer, less frequent trips.

**Implication:** Implementing service adjustments, such as varying bike distribution and station capacities, can accommodate different trip durations and enhance overall user convenience.

5. **Behavioral Insights**: Member users exhibit more consistent usage patterns throughout the week, whereas casual users show greater variability.

**Implication**: Designing targeted promotions and incentives to encourage casual users to convert to membership can potentially increase user retention and loyalty.

### 5. Share Phase

#### **Key Findings:**

- User Segmentation: Tailor marketing strategies to increase engagement based on commuting and leisure patterns.
- Seasonal Variability: Adjust operations to optimize resources across peak and off-peak seasons. Station Utilization: Ensure adequate bikes and docking space at high-traffic stations to meet demand effectively.
- 3. **Trip Characteristics:** Optimize bike distribution to accommodate different trip lengths for enhanced user convenience.
- 4. **Behavioral Insights:** Design targeted promotions to encourage casual users to convert to membership for improved retention.

#### Strategic Recommendations:

- Membership Promotion: Develop targeted marketing campaigns to convert casual users into members, emphasizing the cost-effectiveness and convenience of membership benefits.
- 2. **Station Management:** Regularly adjust bike distribution across stations to meet demand fluctuations, focusing on high-traffic areas identified in the analysis.
- 3. **Seasonal Campaigns:** Implement promotions during peak tourist seasons to attract more casual users, leveraging observed demand patterns.
- 4. **Operational Efficiency:** Optimize bike availability and station capacity based on user behavior insights to improve service reliability and efficiency.

## 6. Act Phase: Recommendations

#### **Top Three Recommendations:**

#### 1. Implement Targeted Membership Promotion Campaigns:

**Rationale:** Member users significantly outnumber casual users, indicating a strong potential for growth by converting casual users into members.

**Action Plan:** - Launch targeted marketing campaigns highlighting the cost-effectiveness and convenience of membership benefits. - Consider offering incentives like discounted memberships for first-time users or referral bonuses to encourage sign-ups.

#### 2. Optimize Station Management Based on Popularity Insights:

**Rationale:** Certain stations serve as popular starting and ending points, suggesting high-demand areas for bike availability.

**Action Plan:** - Increase bike allocation and consider expanding docking spaces at these high-traffic stations, particularly during peak hours, to improve availability and reduce wait times. - Regularly monitor and adjust bike distribution based on real-time demand data to enhance service efficiency.

#### 3. Seasonal Promotions to Capture Peak Demand:

**Rationale:** Seasonal variations influence bike usage, with summer months showing higher overall activity, especially among casual users.

**Action Plan:** - Plan and execute seasonal promotional campaigns targeting casual users during peak tourist seasons. - Offer promotions or discounts aligned with increased demand to attract more users and maximize ridership.

These recommendations aim to capitalize on user behavior insights to enhance marketing effectiveness, operational efficiency, and overall user satisfaction for Cyclistic. By focusing on converting casual users into members, optimizing station management based on popularity trends, and leveraging seasonal demand with targeted promotions, Cyclistic can strategically grow its user base and improve service quality.

# Conclusion

By analyzing user behaviors, sharing key insights, and acting on targeted recommendations, Cyclistic can enhance operational efficiency, improve user satisfaction, and strengthen market position in the competitive bike-share industry. Continual adaptation to user trends and behaviors will be crucial for sustaining growth and meeting evolving customer expectations.