

# Cyclistic Data Analysis Report

*Insights into Member and Casual Rider Behavior*

**Prepared for:**

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## 1. Executive Summary

This report examines how Cyclistic's annual members and casual riders use their services. Key findings indicate that although members make 1.7× more rides, casual riders enjoy significantly longer ride durations. Usage patterns reveal that both groups peak around 5 PM and show distinct behaviors regarding station usage and day-of-week preferences. Based on these insights, three targeted recommendations are provided to convert casual users into members, optimize rewards during rush hours, and launch seasonal promotions.

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## 2. Introduction

### 2.1 Business Problem & Objectives

- **Problem Statement:**
  - How do annual members and casual riders use Cyclistic bikes differently?
  - Why would casual riders consider an annual membership?
  - How can digital media be used to influence casual riders to convert to members?
- **Business Task:**

Study and compare the behaviors of casual and annual riders to develop a marketing plan that effectively converts casual users into long-term annual members.

### 2.2 Stakeholders

- **Lily Moreno:** Director of Marketing and Manager

- **Cyclistic Executive Team:** Decision makers for the recommended marketing program

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## 3. Data Sources & Methodology

### 3.1 Data Sources

- Primary data was sourced from public trip data available at: <https://divvy-tripdata.s3.amazonaws.com/index.html>
- Data provided under the license from Motivate International Inc.

### 3.2 Data Cleaning & Preparation

- Blank rows and cells were removed.
- Duplicates, particularly for rides under one minute, were identified and excluded.
- Inconsistent time entries (e.g., negative ride lengths) were corrected or removed.
- Data was backed up and processed in Excel (\_V02 and \_V03 versions) to ensure integrity.
- New columns were created (ride\_length, day\_of\_week) and functions such as TRIM and WEEKDAY were applied for accurate formatting.

### 3.3 Tools & Techniques

- **Excel:** Used for initial data cleaning, pivot tables, and preliminary analysis.
- **SQL Queries:** Deployed for detailed data segmentation (e.g., ride counts, average ride durations, station popularity, temporal trends).

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## 4. Analysis & Findings

### 4.1 General Usage Overview

- **Ride Counts:**
  - Members completed 2,688,161 rides versus 1,514,400 by casual riders.
  - Members take roughly 1.7× more rides than casual users.

- **Ride Duration:**
  - Average ride for members: 00:12:34
  - Average ride for casual riders: 00:24:11
  - Casuals ride almost twice as long on average.

## 4.2 Station Usage Patterns

- **Popular Stations:**
  - Casual riders tend to use stations for round trips (e.g., Streeter Dr & Grand Ave, DuSable Lake Shore Dr & Monroe St).
  - Members display non-round trip patterns with distinct start and end stations.

## 4.3 Temporal Analysis

- **Busiest Days:**
  - Members: Peak on Wednesdays
  - Casuals: Peak on Saturdays
- **Peak Hours:**
  - Both groups show peak usage at 5 PM, with 4 PM as a close second.
- **Seasonal Trends:**
  - Usage peaks in mid-Summer through early Fall and drops during winter.

## 4.4 Bike Type & Ride Characteristics

- **Bike Preference:**
  - Classic bikes are most popular among both members and casual riders.
- **Long-Distance Riders:**
  - Over 869,000 casual riders have rides longer than 12 minutes, suggesting they are prime candidates for conversion.

## 4.5 Day Type Analysis

- **Weekday vs. Weekend:**
  - Both members and casual riders primarily use the bikes on weekdays.

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# 5. Recommendations & Action Plan

## 5.1 Immediate Plan for Membership Conversion

- **Seamless Membership Funnel at High-Traffic Stations:**
  - "Ride 3x, Get a Discount" incentive: Casual riders earning three trips in a week unlock a one-week trial membership, with clear cost-savings shown at checkout.
  - Gamification elements (progress bars, notifications) to motivate continuous usage.
  - QR Code sign-ups at popular stations for instant activation.
  - Discounted first-time membership offers following the trial period.

## 5.2 Ongoing Perks & Rewards Strategy

- **Rush Hour Incentives:**
  - Introduce "Member-Only Lanes" or priority reservations during peak hours.
  - "Commute & Save" program: Casual riders earn ride credits during rush hours, with additional benefits for members.
  - Bonus challenges on Wednesday (members) and Saturday (casuals) to accumulate extra rewards.
  - Group memberships and referral discounts to leverage social influence.

## 5.3 Seasonal Promotion Initiatives

- **Targeted Summer Promotions:**
  - Personalized in-app messages for casual riders taking long rides, offering membership discounts.
  - Partnerships with local events to distribute discounted or free trial memberships.
  - "Summer Ride Pass": A low-risk seasonal membership to encourage trial usage, which can lead to long-term membership conversion.

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# 6. Conclusion & Next Steps

## Final Conclusions:

- Members ride more frequently but for shorter durations, while casual riders have longer rides and exhibit round-trip patterns.
- Peak usage times and days are consistent across both groups, with significant seasonal variations.

## Application & Next Steps:

- Implement the immediate conversion funnel at popular stations.
- Launch targeted digital and in-app campaigns based on peak usage insights.
- Explore additional data (e.g., rider feedback, geo-demographic information) to further refine marketing strategies and tailor membership offerings.

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## 7. Appendices

### SQL Queries

#### 1. General Usage Overview by User Type

```
SELECT member_casual, COUNT(*) AS total_rides
FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`
GROUP BY member_casual;
```

#### 2024 Results:

Member Total Rides= 2,688,161

Casual Total Rides= 1,514,400

Total riders= 4,202,561

#### Observation:

- Members take approximately 1.7x more rides than casual riders.

#### 2. Average Ride Duration by User Type

```
SELECT member_casual,
        FORMAT_TIMESTAMP('%H:%M:%S',
        TIMESTAMP_SECONDS(CAST(AVG(TIMESTAMP_DIFF(ended_at, started_at,
        SECOND))) AS INT64))) AS avg Ride Duration
FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`
```

GROUP BY member\_casual;

### 2024 Results:

Member Avg Ride= 00:12:34

Casual Avg Ride= 00:24:11

### Observation:

- Casuals take longer rides than members by almost 2x the time.

### 3. Most Popular Start & End Stations for Each User Type

```
SELECT member_casual, start_station_name, end_station_name, COUNT(*) AS  
ride_count
```

```
FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`
```

```
GROUP BY member_casual, start_station_name, end_station_name
```

```
ORDER BY ride_count DESC
```

```
LIMIT 10;
```

### 2024 Results:

member_casual	start_station_name	end_station_name	ride_count
casual	Streeter Dr & Grand Ave	Streeter Dr & Grand Ave	8,396
casual	DuSable Lake Shore Dr & Monroe St	DuSable Lake Shore Dr & Monroe St	6,868
member	State St & 33rd St	Calumet Ave & 33rd St	5,476
member	Calumet Ave & 33rd St	State St & 33rd St	5,441
casual	DuSable Lake Shore Dr & Monroe St	Streeter Dr & Grand Ave	5,234
casual	Michigan Ave & Oak St	Michigan Ave & Oak St	4,445

member	Ellis Ave & 60th St	University Ave & 57th St	3,928
member	Ellis Ave & 60th St	Ellis Ave & 55th St	3,739
member	University Ave & 57th St	Ellis Ave & 60th St	3,614
member	Ellis Ave & 55th St	Ellis Ave & 60th St	3,565

#### Observations:

- Casual riders popular stations appear to be round trips.
  - Streeter Dr & Grand Ave
  - DuSable Lake Shore Dr & Monroe St
  - Michigan Ave & Oak St
- Members popular stations appear “not” to be round trips but frequented start and end stations.

#### 4. Busiest Days of the Week for Each User Type

```
SELECT member_casual, EXTRACT(DAYOFWEEK FROM started_at) AS day_of_week,
COUNT(*) AS ride_count

FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`

GROUP BY member_casual, day_of_week

ORDER BY ride_count DESC;
```

#### 2024 Results:

member_casual	day_of_week	ride_count
member	4	448,886
member	5	418,892
member	3	416,372
member	2	391,446
member	6	382,756
member	7	335,467
casual	7	316,674
member	1	294,342
casual	1	263,992
casual	6	223,889
casual	4	187,385



casual	5	183,521
casual	2	178,039
casual	3	160,900

#### Observations:

- Members busiest day is Wednesday (4)
- Casuals busiest day is Saturday (7)

#### 5. Peak Ride Hours by User Type (ideal for time-based membership discounts)

```
SELECT member_casual, EXTRACT(HOUR FROM started_at) AS hour_of_day,
COUNT(*) AS ride_count

FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`

GROUP BY member_casual, hour_of_day

ORDER BY ride_count DESC

LIMIT 20;
```

#### 2024 Results:

member_casual	hour_of_day	ride_count
member	17	293,408
member	16	253,967
member	18	218,808
member	8	193,042
member	15	182,932
member	7	156,619
member	19	151,431
member	14	148,703
member	12	148,550
member	13	147,846
casual	17	145,864
casual	16	138,207
member	11	130,228
member	9	127,141
casual	15	122,154
casual	18	120,368

casual	14	111,871
member	10	111,713
casual	13	108,130
member	20	104,680

#### Observation:

- 5pm (17) is the peak ride time for both members and casuals, followed by 4pm (16) for both groups.

#### 6. Seasonal Trends – Rides Per Month (ideal for targeted membership promotions)

```
SELECT member_casual, EXTRACT(MONTH FROM started_at) AS month, COUNT(*)
AS ride_count
```

```
FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`
```

```
GROUP BY member_casual, month
```

```
ORDER BY member_casual, month;
```

#### 2024 Results:

member_casual	month	ride_count
casual	1	17,373
member	1	93,550
casual	2	37,613
member	2	144,465
casual	3	61,799
member	3	164,708
casual	4	92,234
member	4	200,336
casual	5	164,420
member	5	270,099
casual	6	206,705
member	6	284,522
casual	7	230,169
member	7	307,498
casual	8	226,811
member	8	311,127

casual	9	214,287
member	9	319,102
casual	10	158,441
member	10	288,467
casual	11	68,393
member	11	176,414
casual	12	36,155
member	12	127,873

#### Observation:

- Members & casuals record peak rides from mid-Summer to the start of Fall.
- Members & casuals record lowest rides for Winter season.
- Members are the highest demographic in either case.

## 7. Most Used Bike Type by User Type

```
SELECT member_casual, rideable_type, COUNT(*) AS ride_count
FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`
GROUP BY member_casual, rideable_type
ORDER BY member_casual, ride_count DESC;
```

#### 2024 Results:

member_casual	rideable_type	ride_count
casual	classic_bike	972,078
casual	electric_bike	517,044
casual	electric_scooter	25,278
member	classic_bike	1,775,014
member	electric_bike	891,471
member	electric_scooter	21,676

#### Observation:

- Classic bikes are most popular amongst members and casual riders.

## 8. Identifying Long-Distance Casual Riders (Potential Members)

```
SELECT ride_id, member_casual,  
       FORMAT_TIMESTAMP('%H:%M:%S',  
       TIMESTAMP_SECONDS(TIMESTAMP_DIFF(ended_at, started_at, SECOND)))  
       AS ride_duration  
FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`  
WHERE member_casual = 'casual'  
      AND TIMESTAMP_DIFF(ended_at, started_at, SECOND) >= 720 -- 12 minutes  
      in seconds  
ORDER BY ride_duration DESC
```

### 2024 Results:

The query returned 869,951K results for casual users who ride longer than 12 mins. With the longest ride being 23 hours and 59 minutes.

### Observation:

- Being that the average casual ride length is 2x longer than that of members (12 mins). There are 869K+ casual riders who ride longer than 12 mins and are prime candidates to convert to membership status.

## 9. Analyzing Weekend vs. Weekday Usage

```
SELECT member_casual,  
  
       CASE  
  
       WHEN EXTRACT(DAYOFWEEK FROM started_at) IN (1, 7) THEN 'Weekend'  
  
       ELSE 'Weekday'  
  
       END AS day_type,  
  
       COUNT(*) AS ride_count  
FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`  
  
GROUP BY member_casual, day_type  
  
ORDER BY member_casual, ride_count DESC;
```

### 2024 Results:

member_casual	day_type	ride_count
casual	Weekday	933,734
casual	Weekend	580,666
member	Weekday	2,058,352
member	Weekend	629,809

### Observations:

- Casual riders use bikes mostly during the week just like members do.

## 10. Identify Popular Start Stations Among Casual Riders (Ideal Membership promotions)

```
SELECT start_station_name, COUNT(*) AS total_rides
FROM `hardy-magpie-433820-n2.cyclistic_2024_data.2024`
WHERE member_casual = 'casual'
GROUP BY start_station_name
ORDER BY total_rides DESC
LIMIT 10;
```

### 2024 Results:

start_station_name	total_rides
Streeter Dr & Grand Ave	47,828
DuSable Lake Shore Dr & Monroe St	31,776
Wells St & Elm St	23,375
Michigan Ave & Oak St	23,145
DuSable Lake Shore Dr & North Blvd	21,178
Millennium Park	20,626
Shedd Aquarium	19,867
Dusable Harbor	17,127
Theater on the Lake	15,302
Michigan Ave & 8th St	12,445

### Observations:

- The top 6 stations have over 20K - 40K+ casual riders each. This totals approximately 165k+ potential new members.

### Excel & Pivot Table Example

	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O
1	ride_id	rideable_type	started_at	ended_at	ride_length	day_of_week	start_station_name	start_station_id	end_station_name	end_station_id	start_lat	start_lng	end_lat	end_lng	member_casual
2	88BC690D0B048221	classic_bike	5/30/2024 14:16	5/30/2024 14:17	0:01:00		5 California Ave & M	15622	California Ave & M	15622	41.961068	-87.695439	41.961068	-87.695439	member
3	132131456CD4A3D1	classic_bike	5/22/2024 10:40	5/22/2024 10:41	0:01:00		4 Mies van der Rohe	13338	Michigan Ave & Pe	13034	41.89694463	-87.62175769	41.89766	-87.62351	member
4	DE915E536084C851	electric_bike	5/1/2024 17:25	5/1/2024 17:26	0:01:00		4 Halsted St & Fulton	23003	Halsted St & Fulton	23003	41.88665962	-87.64810336	41.886871	-87.648089	member
5	1E9E61718FBF1B01	classic_bike	5/30/2024 15:00	5/30/2024 15:01	0:01:00		5 Streeter Dr & Grant	13022	Streeter Dr & Grant	13022	41.892278	-87.612043	41.892278	-87.612043	casual
6	77C1D5AF4B0D07B	electric_bike	5/9/2024 10:22	5/9/2024 10:23	0:01:00		5 Kedzie Ave & Chica	KA1504000114	Kedzie Ave & Chica	KA1504000114	41.89545739	-87.70609689	41.8954653	-87.70612788	member
7	55BDBEC34F240401	classic_bike	5/6/2024 6:48	5/6/2024 6:49	0:01:00		2 Mies van der Rohe	13338	Mies van der Rohe	15529	41.89694463	-87.62175769	41.89858665	-87.62191523	member
8	5D897F3B80DADF	classic_bike	5/26/2024 8:37	5/26/2024 8:38	0:01:00		1 Dearborn Pkwy & C	TA1307000128	Dearborn Pkwy & C	TA1307000128	41.898969	-87.629912	41.898969	-87.629912	casual
9	818FC5980CC48D21	classic_bike	5/27/2024 18:16	5/27/2024 18:17	0:01:00		2 Benson Ave & Chur	596	Benson Ave & Chur	596	42.048214	-87.683485	42.048214	-87.683485	casual
10	06D7AFC8A7C8B58	electric_bike	5/4/2024 16:05	5/4/2024 16:06	0:01:00		7 Clark St & North Av	13128	Clark St & North Av	13128	41.91200912	-87.63202429	41.911974	-87.631942	casual
11	55C26D924BE3DFD	classic_bike	5/27/2024 19:01	5/27/2024 19:02	0:01:00		2 Streeter Dr & Grant	13022	Streeter Dr & Grant	13022	41.892278	-87.612043	41.892278	-87.612043	casual
12	3DB1F5997AE24686	classic_bike	5/11/2024 14:46	5/11/2024 14:47	0:01:00		7 Streeter Dr & Grant	13022	Streeter Dr & Grant	13022	41.892278	-87.612043	41.892278	-87.612043	member

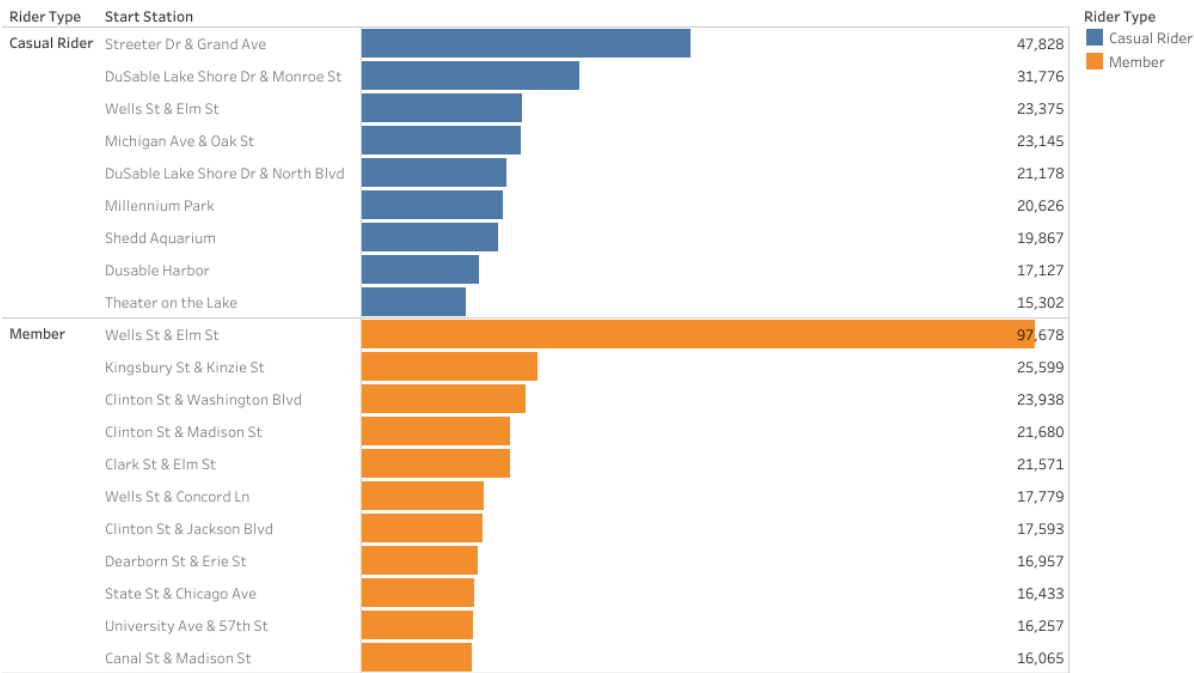
Spreadsheet snapshot displaying the newly created columns, “ride\_length” and “day\_of\_week,” calculated from “started\_at” and “ended\_at” using the formulas =D3-C3 and =WEEKDAY(C3, 1), respectively.

	A	B	C	D	E	F	G	H	I	J	K	L	M
1	Avg ride length			Max ride length									
2	Row Labels	Average of ride_length		Row Labels	Max of ride_length								
3	casual	0:27:22		casual	24:56:25								
4	member	0:13:31		member	24:35:37								
5	Grand Total	0:18:45		Grand Total	24:56:25								
6													
7	Avg ride length by day												
8	Average of ride_length												Legend
9	Row Labels	1	2	3	4	5	6	7	Grand Total				1= Sunday
10	casual	0:31:07	0:27:48	0:21:28	0:24:44	0:23:36	0:26:00	0:31:28	0:27:22				2= Monday
11	member	0:15:10	0:12:51	0:12:34	0:13:14	0:12:37	0:13:01	0:15:59	0:13:31				3= Tuesday
12	Grand Total	0:22:54	0:18:12	0:15:06	0:16:46	0:16:00	0:17:57	0:23:56	0:18:45				4= Wednesday
13													5= Thursday
14	Number of rides by day of week												6= Friday
15	Count of member												7= Saturday
16	Row Labels	1	2	3	4	5	6	7	Grand Total	Most Popular Day			
17	casual	25515	17688	13751	23271	19654	27693	36790	164362	4			
18	member	27050	31765	34583	52595	44047	45213	34798	270051				
19	Grand Total	52565	49453	48334	75866	63701	72906	71588	434413				
20													
21	Bike type rental												
22	Count of member												
23	Row Labels	classic_bike	electric_bike	Grand Total									
24	casual	114072	50290	164362									
25	member	186960	83091	270051									
26	Grand Total	301032	133381	434413									
27													

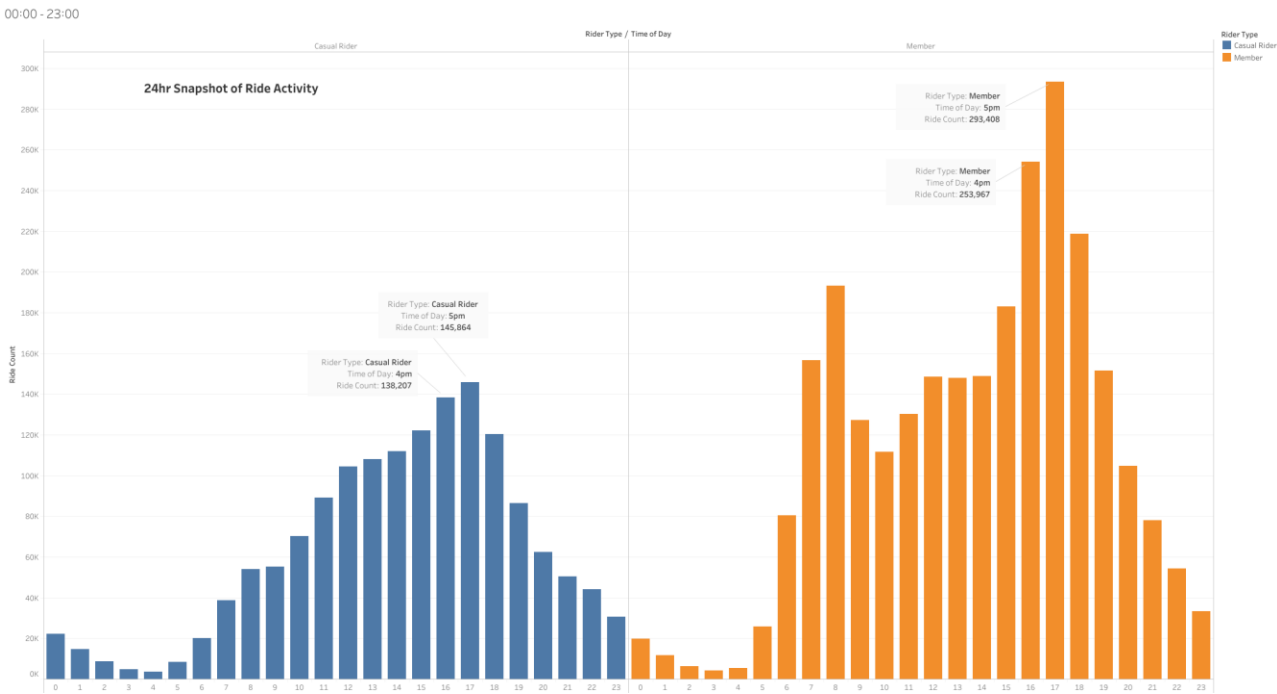
Pivot table snapshot displaying ride calculations comparing members and casual riders.

# Visualizations

## Top 20 Starting Stations

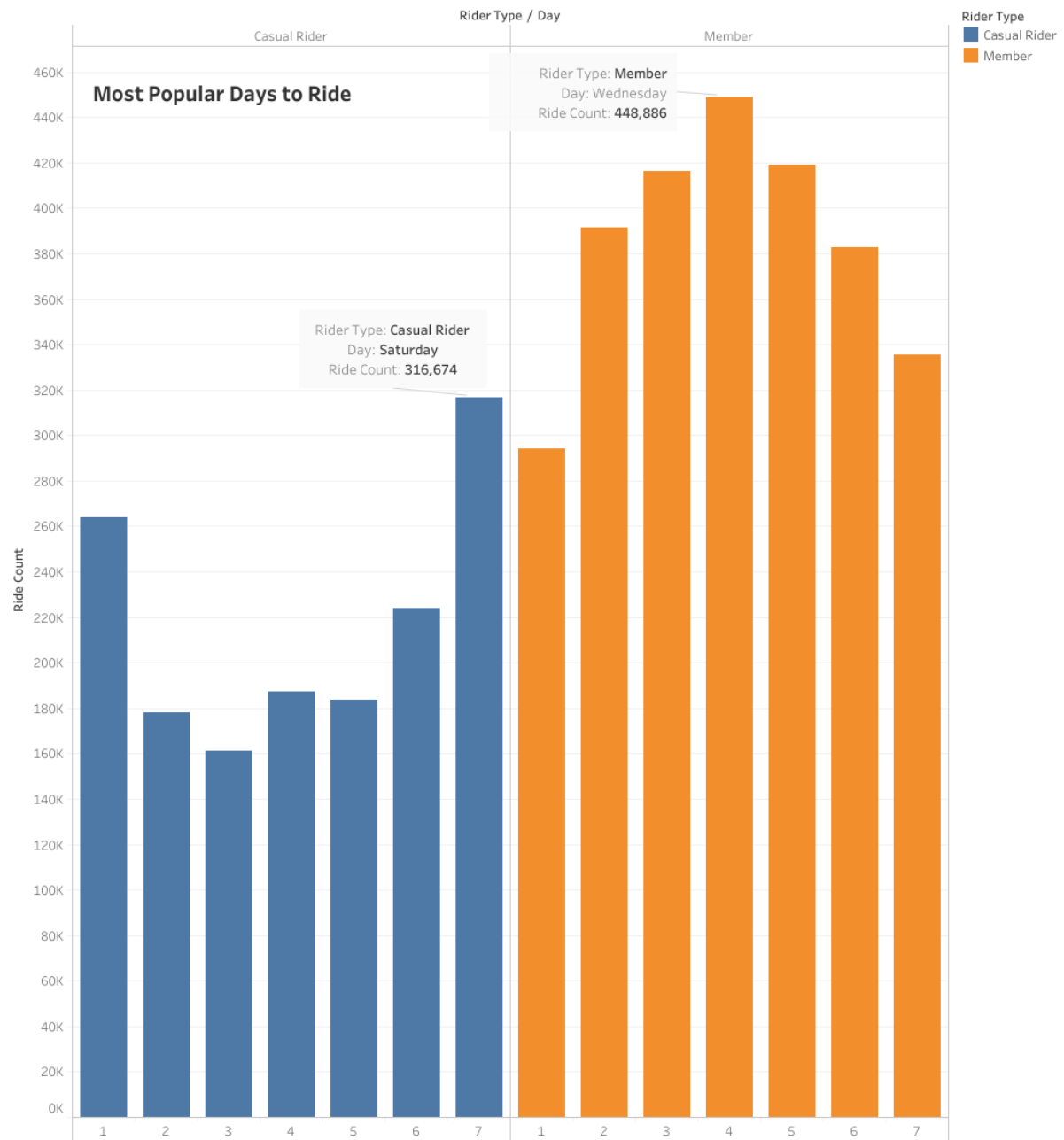


## Popular Stations



## Peak Ride Hours

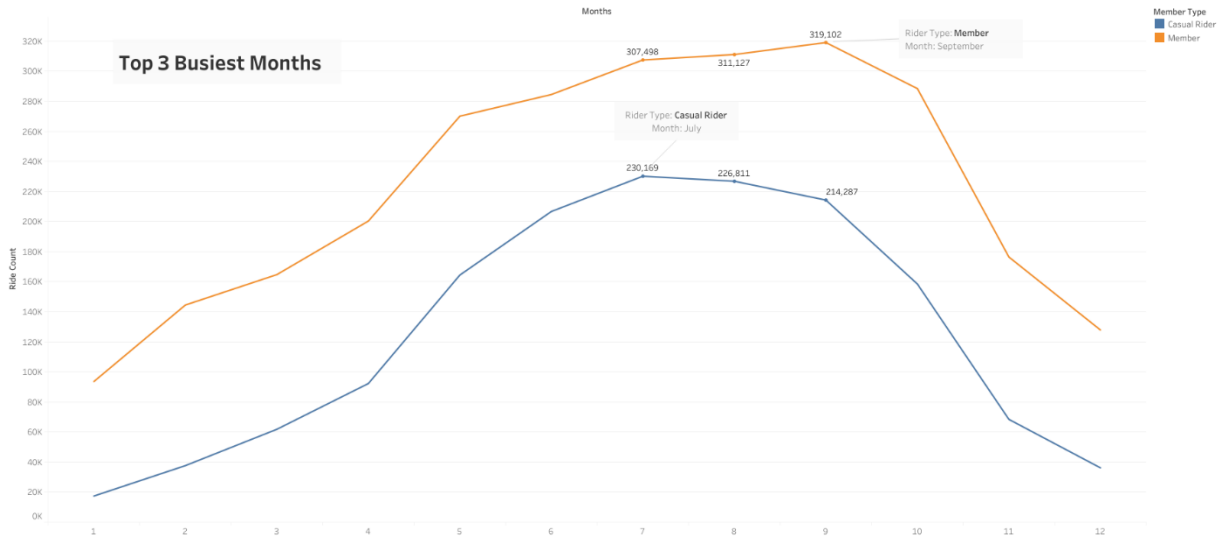
Sunday - Saturday



Peak Ride Days



Seasonal Ride Data



Peak Seasonal Rides