**Google Data Analytics Capstone Project: How Does a Bike-Share Navigate Speedy Success?**

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This case study is one of a part of the Google Data Analytics Professional Certificate program on Coursera. The analysis will follow the 6 phases of the Data Analysis process: Ask, Prepare, Process, Analyze, Share, and Act.

***Ask/Identify the task***

In this case, I am a data analyst working for Cyclistic, which is a fictional company in Chicago. Since its inception in 2016, Cyclistic has grown to a fleet of 5,824 bicycles that are geotracked and locked into a network of 692 stations across Chicago. The bikes can be unlocked from one station and returned to any other station in the system anytime. There are two types of customers: casual riders (Casual) and annual members (Member).

Lily Moreno, the director of marketing and your manager has set a clear goal: **Design marketing strategies aimed at converting casual riders into annual members.** I have been tasked to analyze how annual members and casual riders use Cyclistic bikes differently based on a bike-share company’s data of its customer’s trip details from Jan-Oct 2021 to help the company maximize the number of annual memberships. To tackle this assignment, I will use Excel, SQL, and Tableau to analyze the Cyclistic historical bike trip data to identify trends by answering 6 main questions below:

1. What is the total number of trips for users, and what percentage of total trips does each user represent?

2. What is the average riding time and are there particular days of the week?

3. What is the average number of trips taken every month?

4. What is the busiest hour of the day?

5. What are the top 5 common starting and ending stations for each?

6. Is there a preference for the rideable type for members and casual?

***Prepare***

The [dataset](https://divvy-tripdata.s3.amazonaws.com/index.html) used in this case study is actual public data made by Motivate International Inc. under this [license](https://www.divvybikes.com/data-license-agreement). The data is located on the Cyclistic server and is stored by separated files for each month. Because this is a case study using public data, we are going to assume the data is credible. However, I found that start\_station\_name/end\_station\_name have multiple NULL values. I downloaded the data from the divvy trip data and saved it on the desktop. The data contains 13 columns.

ride\_id: ID rider

ideable\_type: the type of bicycle used

started\_at: the date and time that the bicycle was checked out

ended\_at: the date and time that the bicycle was checked in

start\_station\_name: where the trip started

start\_station\_id: a unique identifier for the start station

end\_station\_name: where the trip ended

end\_station\_id : a unique identifier for the end station

Start\_lat: the latitude of the start

stationstart\_lng: the longitude of the start station

end\_lat: the latitude of the end station

end\_lng: the longitude of the end station

member\_casual: a field indicating whether it was taken by casual riders or annual member

***Process and Analyze***

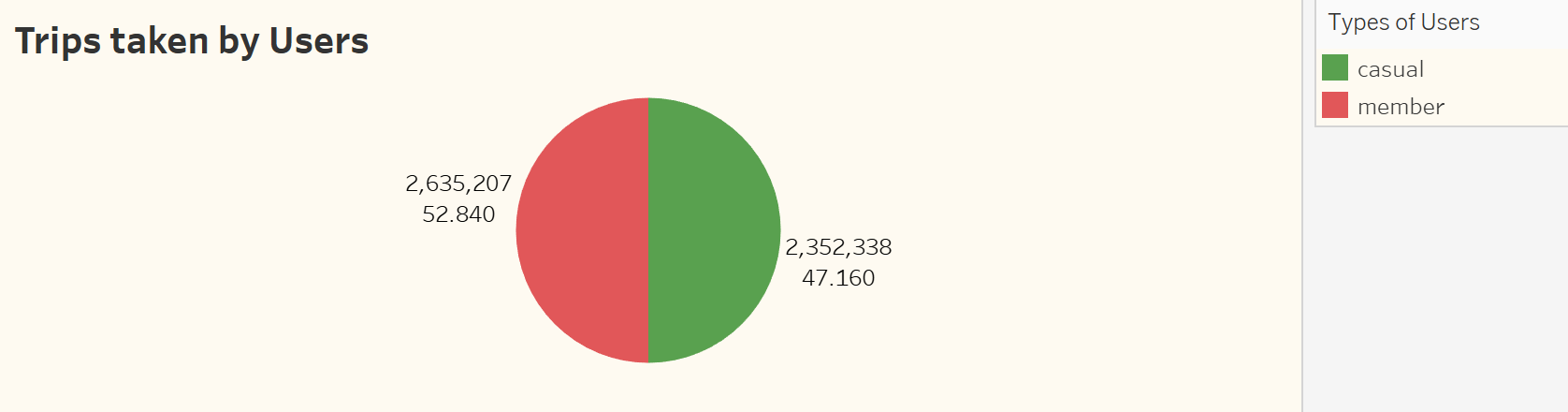
In this step, I am going to process the data and prepare it for the next step where I will find the answer to the questions. Tools that I am going to use are SQL for data cleaning and data analyzing and Tableau for data visualization. Please click the following link to see the dashboard that I created for this project.

First, after ensuring that all columns names are the same, then I merged 12 datasets into a unionized dataset by using SQL with Microsoft SQL on my laptop. I excluded 4 columns Start\_lat, start\_lng, end\_lat, and end\_lng because they are not necessarily used during the analyzing process.

Some notes about the data:

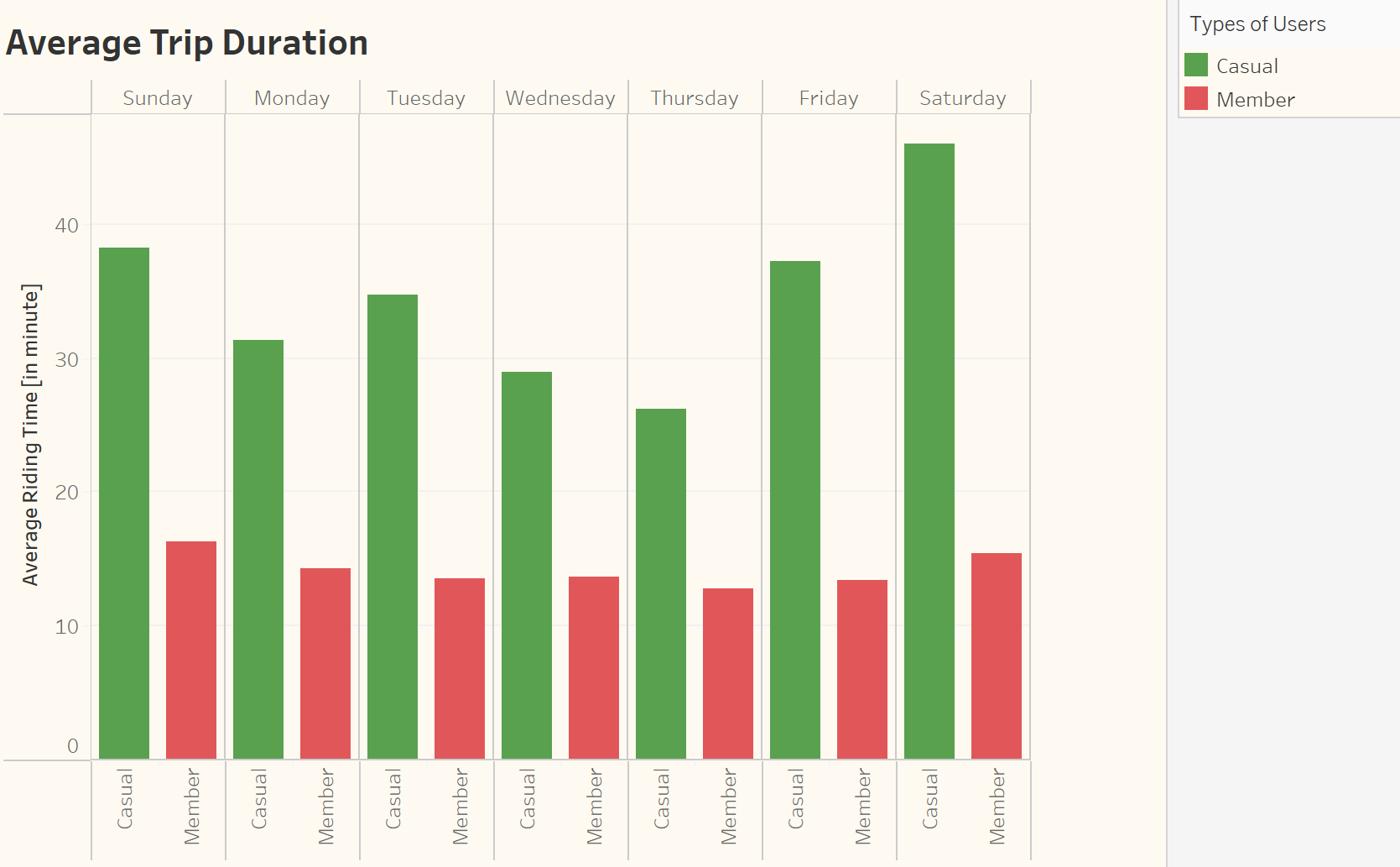
1. The dataset contains 4,987,545 rows.
2. 564,456 rows have null values in the start\_station\_name and 606,485 rows have null values in the end\_station\_name. These rows will be still be used in the overall analysis but will be filtered out when analyzing the top 5 start and end stations’ names.

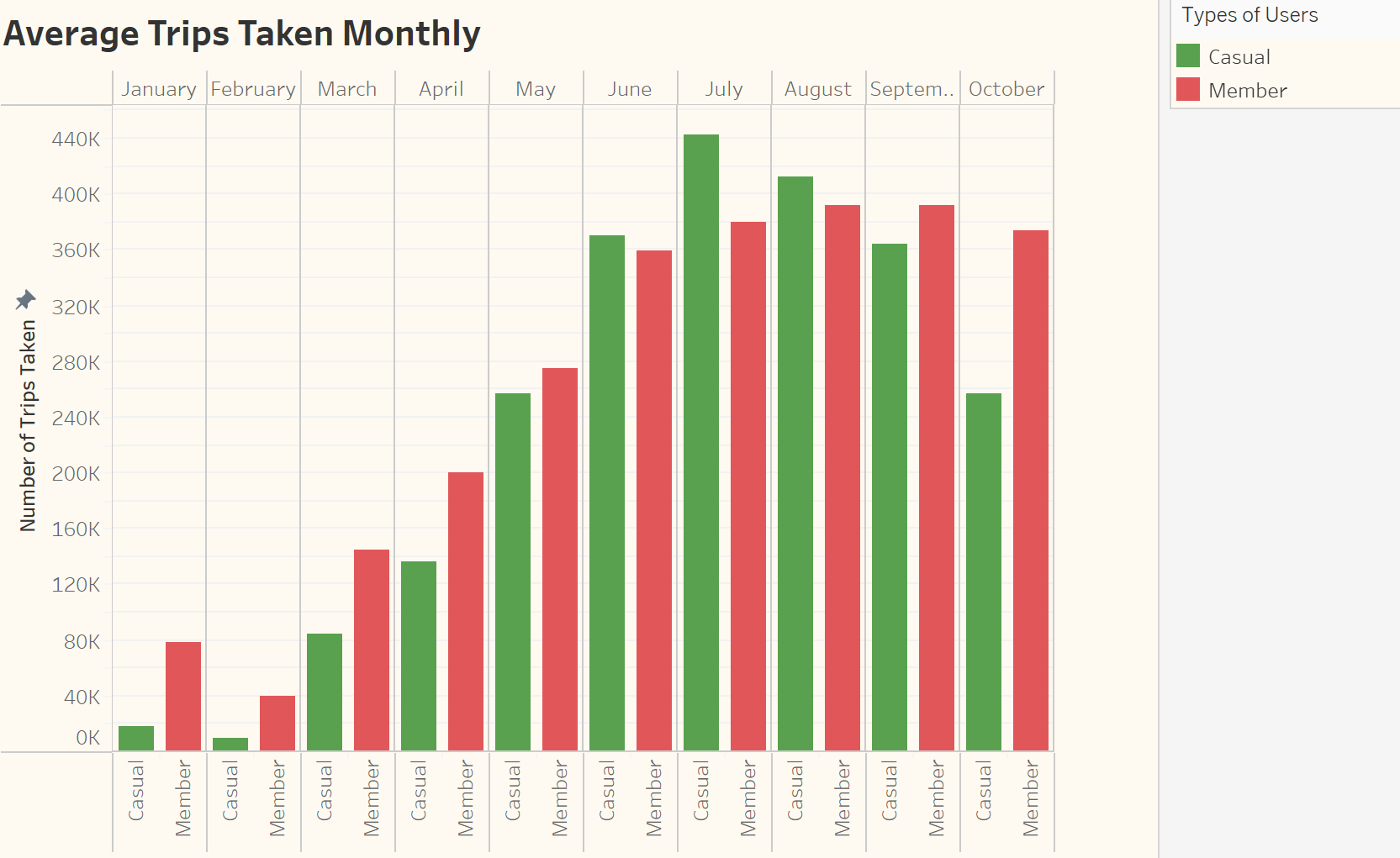
There are a total of 4,987,545 trips taken during the period Jan-Oct 2021. We can see the pie chart below showing the total number of trips taken by annual members is a little more than a half of the total number of trips taken by casual riders.



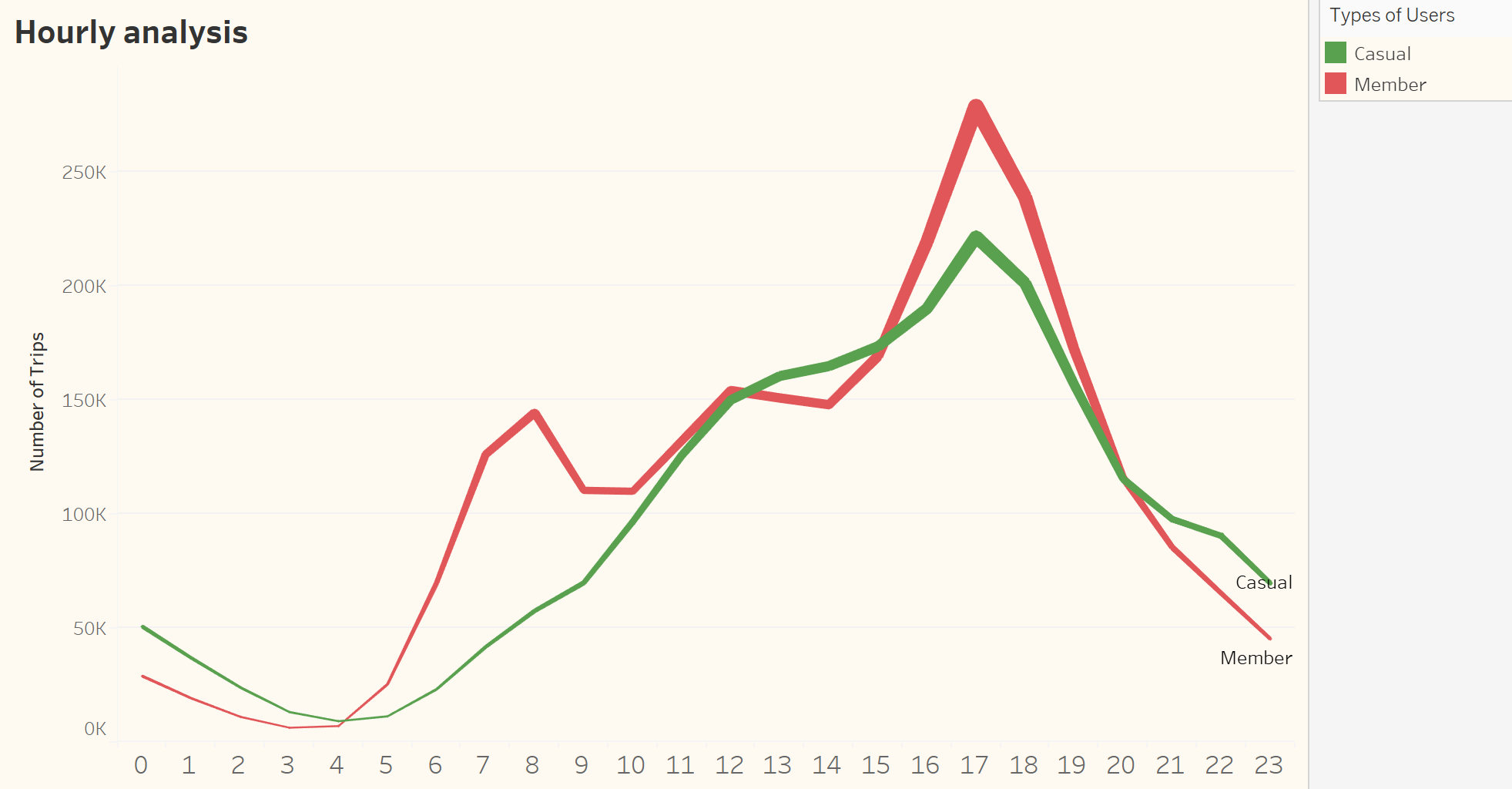
The observation here is that members tend to take short trips than casuals. In addition, casuals tend to rent the bikes mostly on weekends (Saturdays and Sundays) while members’ usage is evenly distributed throughout the week.

There is a significant increase in trips taken during summertime when the weather is warmer.

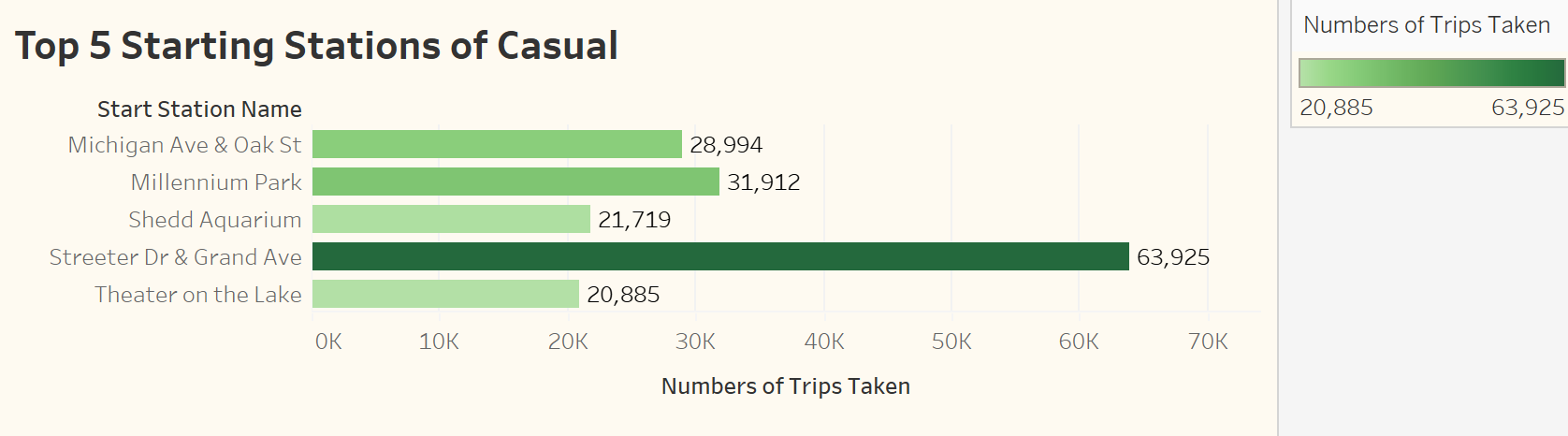


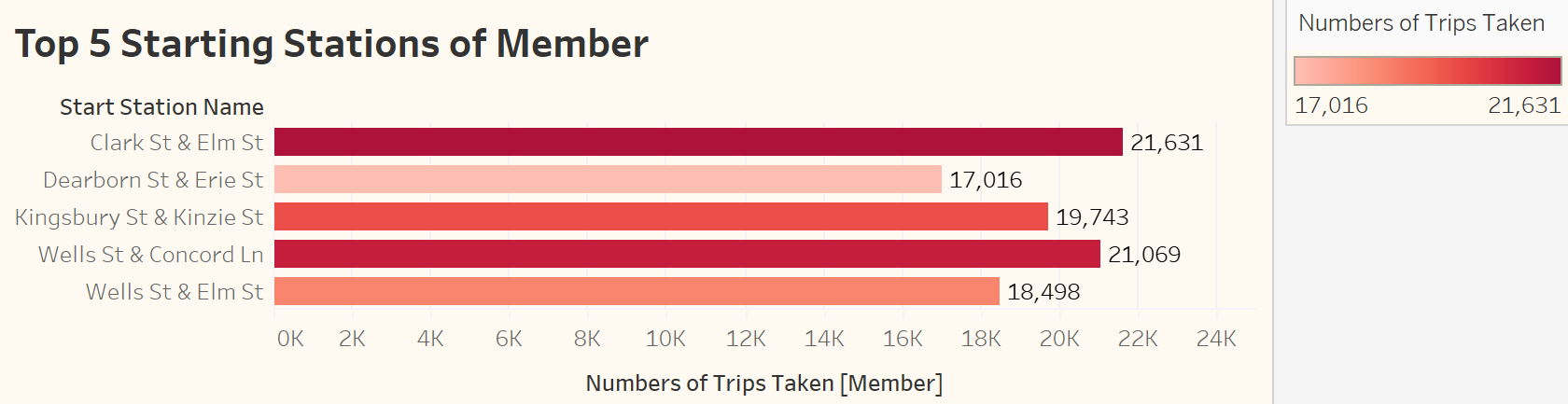


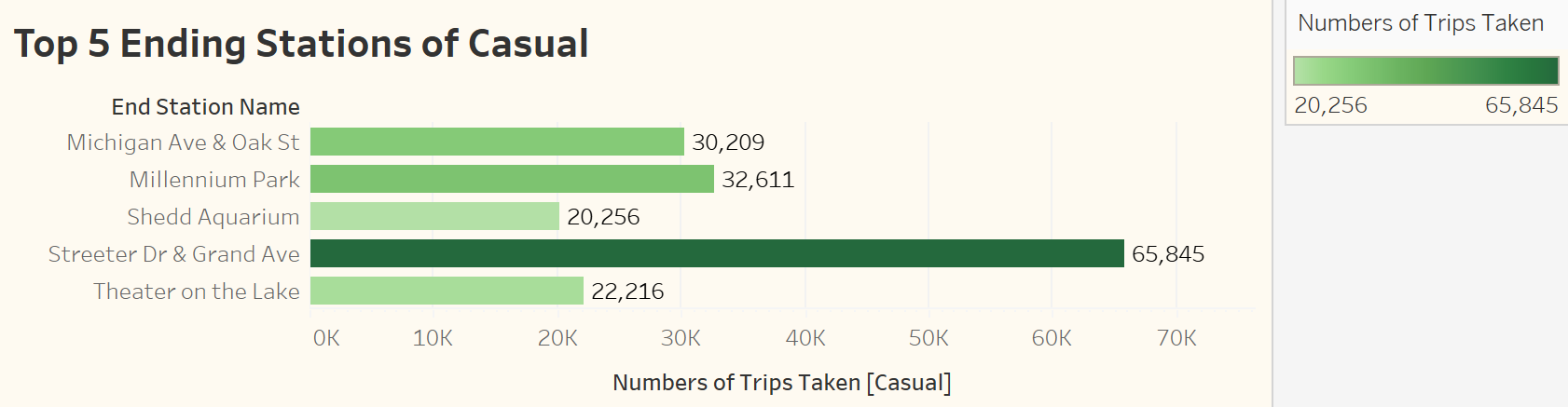
Both member and casual users’ lines display peak hours during the day at 5:00 PM. The busy hours normally occur in the afternoon between 4:00 PM and 7:00 PM. The thickness of the lines increases when the number of trips increases.

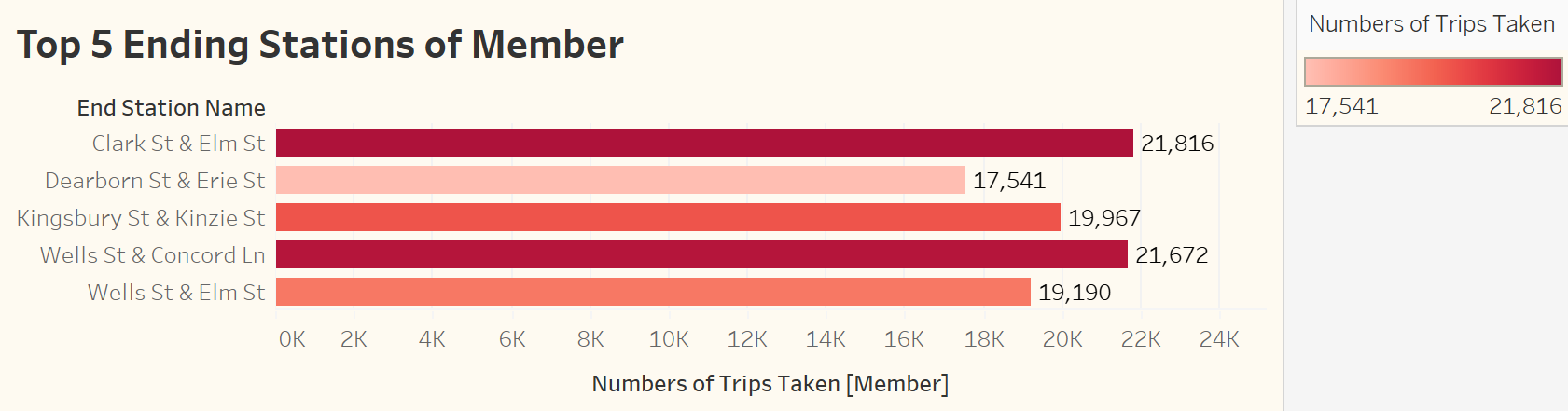


The four graphs below show the top 5 starting/ending stations of Casual and Member. The locations visited by casual users mostly are nearby downtown and are tourist locations.

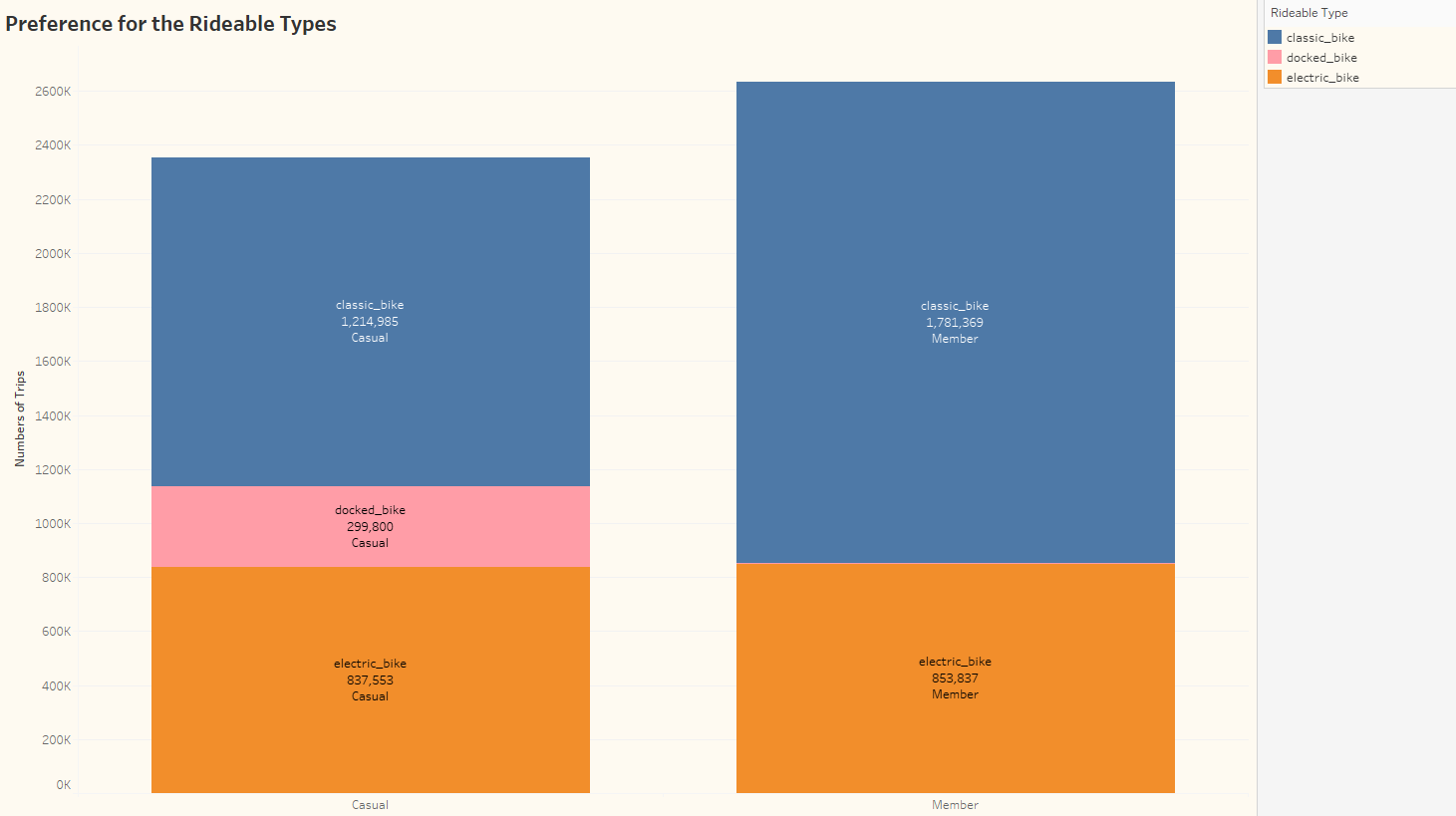








Now, let’s plot a bar graph that shows the likely distribution of the member and casual users with bike types. The classic bike is the most popular one among the 3 types. In contrast, the docked bike is the least popular one.



***Share and Act***

After using SQL to communicate and analyze the dataset and Tableau for data storytelling, it is time to share the result and answer the questions “How can we convert casuals to members?”

1. We see that the season affects the total trip duration for casual riders and annual members. The temperature is very low during the winter season, people are not willing to go out as normal as spring or summer, and people who need to commute to work would rather choose other public transports like bus or train. Therefore, the idea of **lowering the cost of the annual membership plan during winters will lure the riders to use for bike share over public transport.**

2. Members tend to take short trips than casuals. However, casuals tend to rent the bikes mostly on weekends (Saturdays and Sundays) while members’ usage is evenly distributed throughout the week. Therefore, we can increase renting a bike for the weekend to target casual riders into having a membership, also we should launch more special deals for annual members only to motivate casual riders to switch their plans like members have the option to book weekday rides at a lower rate.

3. Both annual member and casual riders prefer classic bikes. The docked bike is the least popular one, especially there is only one docked bike rented during all the period of analysis.

4. We know the most popular start/end station names and routes for casual users, we should put banners or special discount advertisements in those areas or routes that would target casual users.

5. The company can also run a survey to collect the data from current casual riders to understand what kind of benefits they look for when considering subscribing to annual members, and then create a plan to act based on the result of the survey.

*For interactive Tableau Dashboard, please click* [*here*](https://public.tableau.com/app/profile/khanh.pham4251/viz/BikeShareAnalysisDashboard/Dashboard1?publish=yes)*.*

*For SQL queries, please click* [*here*](https://github.com/kpham2104/Portfolio/blob/main/BikeShare_SQLQuery.sql)*.*