

Cyclistic Customer Analysis

Stefanus Yudi Irwan – September 11th 2022

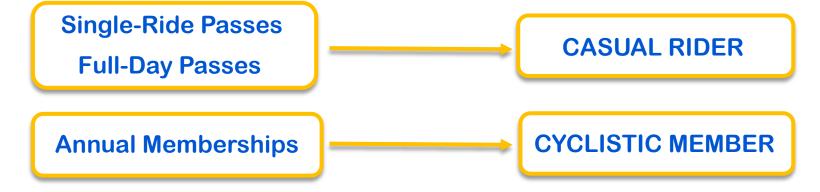




About Cyclistic



- Cyclistic is a bike-share program that features more than 5,800 bicycles and 692 docking stations in Chicago. The bikes can be unlocked from one station and returned to any other station in the system anytime.
- Cyclistic apply flexible pricing plans: single-ride passes, full-day passes, and annual memberships.















Outline

→ Output Steps -



PROCESS ——— Clean Data







ANALYZE ————







ACT ----- Marketing Strategy <a>©





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Project Stakeholders



CYCLISTIC Executive Team

Marketing program recommendation approval



CYCLISTIC Director of Marketing

Responsible for bike-share program campaigns and initiatives development





Problems



Company want more profit to expand the business



Annual members are much more profitable than casual riders



Annual members will be the key to company's future growth.



Project Goal

"Increase the number of Annual members by at least 25% within the Q1 of the year, through conversion from casual rider to annual member"

Business Task

"Analyze CYCLISTIC historical bike trip data to identify the behavior of the casual rider and annual member to enhance the number of annual members through marketing strategies to increase profit and grow the company."







Data Collection

Data Requirements.

- Bike user time data
- Bike user place data
- Bike station data
- Type of bike used
- User payment type

Scope of Data.

One year of historical data from August 2021 till July 2022 was used for this analysis

Data Source.

Raw data was stored within <u>this website</u>. The data was provided by Google Data Analytics Professional Certificate. This is an open data under <u>this license</u>.





Dataset Overview



Raw Data comprises of **12 CSV files**, represent each month bike-share historical data



Size of the raw data is 1,01 Giga-Byte



Raw Data comprises of 5,901,463 rows, and 13 columns



Raw Data Composition

Time Data

- Started_Time
- Ended_Time

- 2 Spatial Data
- Start_latitude
- Start_longitude
- End_latitude
- End_longitude
- 3 Category Data
- Rideable_type
- Payment_type

- 4 ID Data
- Ride_id
- Start_station_id
- Start_station_name
- End_station_id
- End_station_name

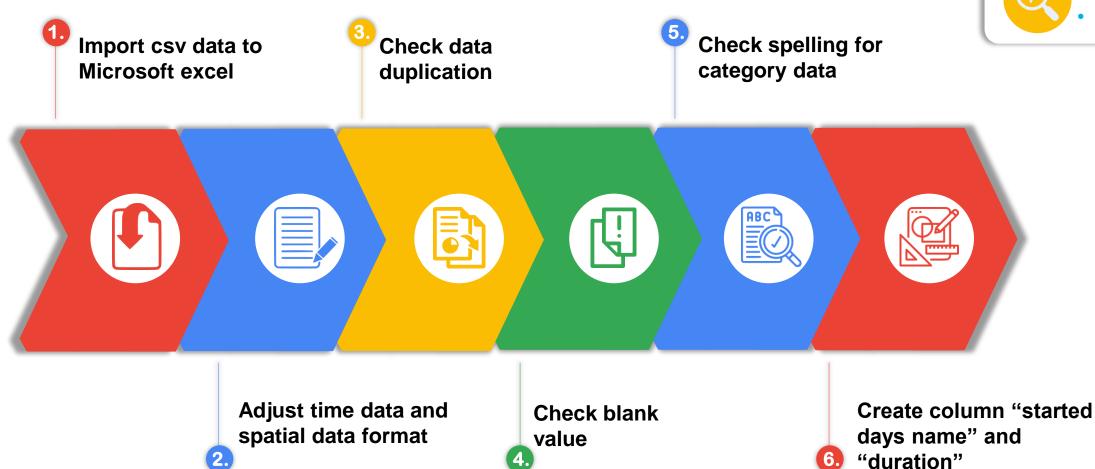




DATA CLEANING STEPS





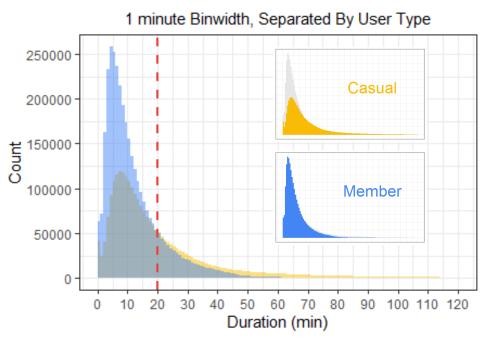


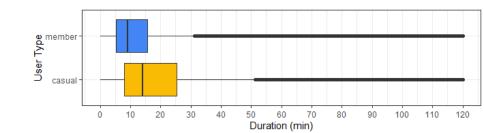




Analysis By Duration

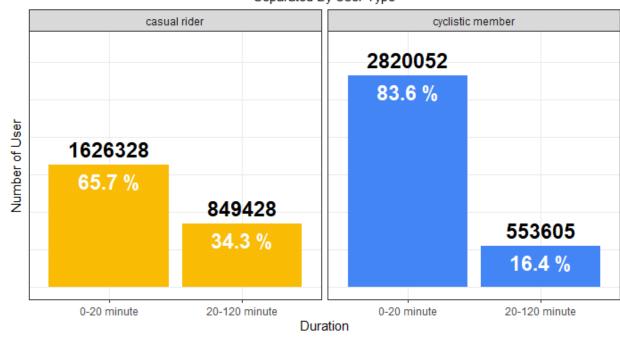
Duration Data Distribution





Number Of User By Duration

Separated By User Type



💡 Insight.

User Type

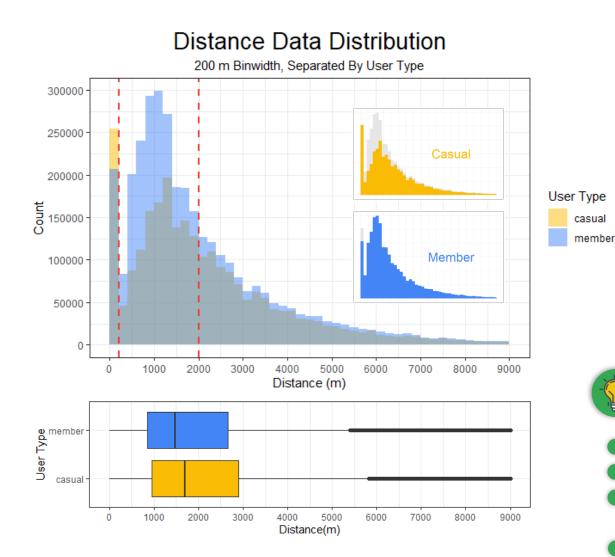
casual

member

- casual rider tend to use the bike longer than cyclistic member
- cyclistic member and casual rider most likely uses the bike within time 0-20 minutes.
- more casual rider at time frame 20-120 minutes than cyclistic member
- casual rider at time frame 0-20 minutes will be potential to be cyclistic member

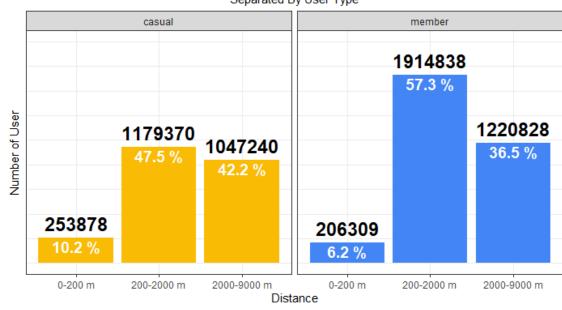


Analysis By Distance



Number Of User By Distance





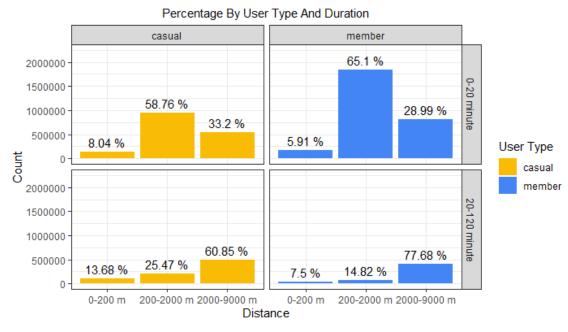


- most cyclistic member use the bike for distance 200-2000 m.
- more casual rider at distance 0-200 m than cyclistic member.
- cyclistic member has more difference between distance 200-2000m and 2000-9000m than casual rider.
- casual rider at distance 200-2000 m will be potential to be cyclistic member.

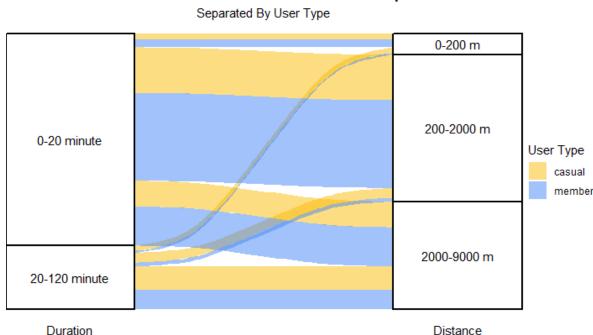


Analysis By Duration and Distance

Rider Duration and Distance Count



Rider Duration and Distance Proportion



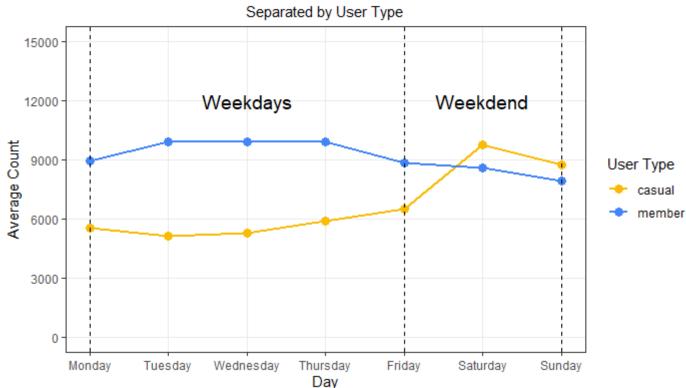


- Cyclistic member most likely use the bike for duration 0-20 minute and distance 200-9000m, also for duration 20-120 minute and distance 2000-9000 m
- The difference between the most dominant count and the rest of the count in every group is larger at cyclistic member than at casual rider
- Casual rider with duration 0-20 minute and distance 200-9000 m or with duration 20-120 minute and distance 2000-9000 m will be a good candidate to be cyclistic member



Analysis By Days

Average Rider Count by Day



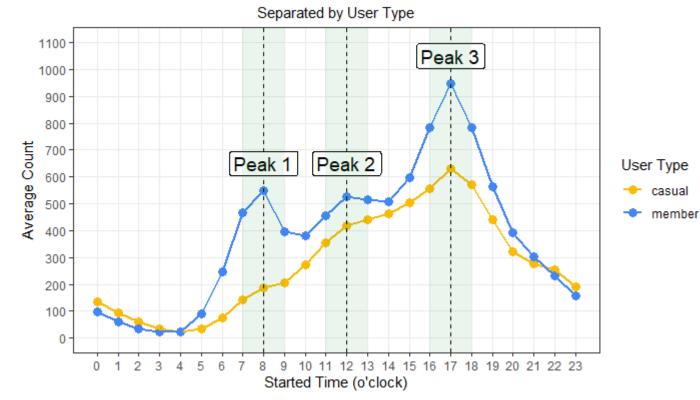
Insight.

- Average cyclistic member per day is stable at weekdays and going down at weekend.
- Average casual rider per day raise at the weekend.
- Casual rider at weekdays will be good potential to be cyclistic member
- Weekend is a good time for cyclistic member promotion activity because there are more casual rider



Analysis By Started Time

Average Rider Count by Started Time

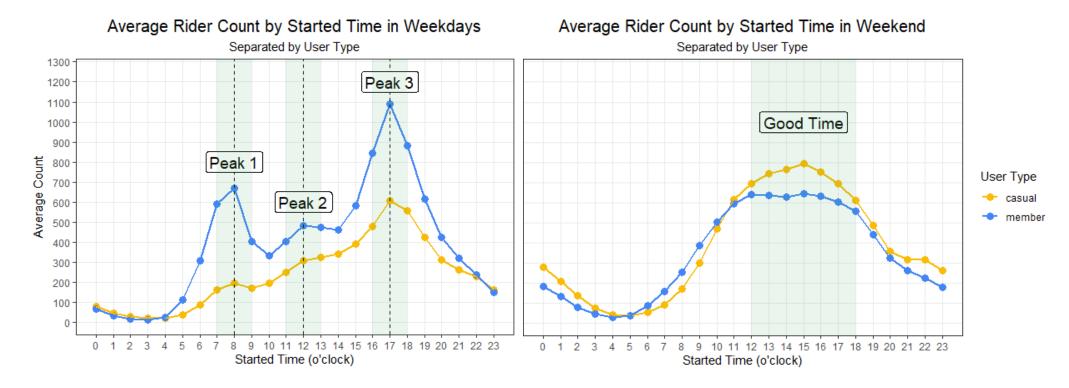


🌹 Insight.

- There are 3 peak time for cyclistic member, 8, 12, and 17 o'clock. This is might be the time when commuting to work, lunch break, and commuting to home.
- Casual rider only has obvious peak at 17 o'clock.
- Casual rider at time 7-9, 11-13, and 16 18 o'clock could be good potential candidate to be cyclistic member.



Analysis By Days and Time



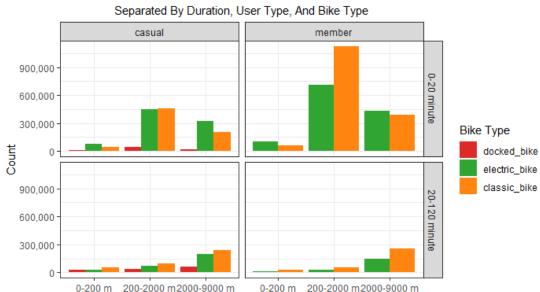


- in weekdays there are still **3 peak time** for cyclistic member, **8**, **12**, and **17** o'clock.
- casual member at weekdays now has 2 obvious peak at 8 and 17 o'clock.
- more casual rider at weekend 12 18 o'clock (afternoon) this could be a good time to promote the cyclistic member conversion program
- casual rider at time 7-9, 11-13, and 16 18 o'clock at weekdays could be good potential candidate to be cyclistic member.



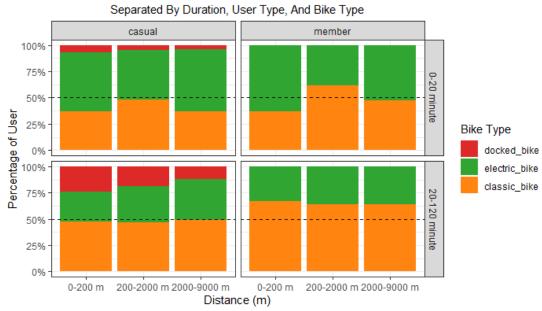
Analysis By Bike Type

User Type Bike Count by Distance



Distance (m)

User Type Bike Proportion by Distance

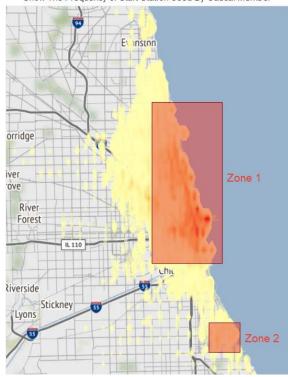




- Cyclistic members prefer to use a classic bike over an electric bike for distances 200-2000 m and duration 0-20 minutes, while casual riders are still 50:50 for classic bikes or electric bikes.
- For distances 2000 9000 m and duration 0-20 minutes either cyclistic members or casual riders prefer an electric bike to a classic bike
- There are no cyclistic members who used docked bikes, then there is no use to promote cyclistic members to docked bike users.
 - Casual rider with a short duration (0-20 minute) and medium to long distances (200-9000 m) who use the classic or electric bike is a good group to be cyclistic member



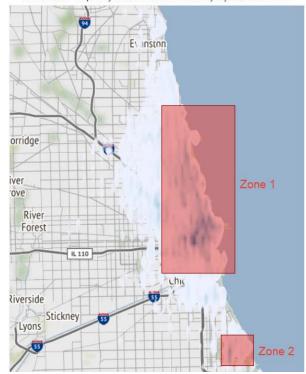
Casual Rider Start Station Density Plot Show The Frequency of Start Station Used By Casual Member



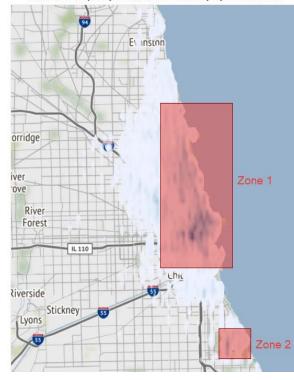
Casual Rider End Station Density Plot Show The Frequency of End Station Used By Casual Member



Cyclistic Member Start Station Density Plot Show The Frequency of Start Station Used By Cyclistic Member



Cyclistic Member End Station Density Plot Show The Frequency of End Station Used By Cyclistic Member

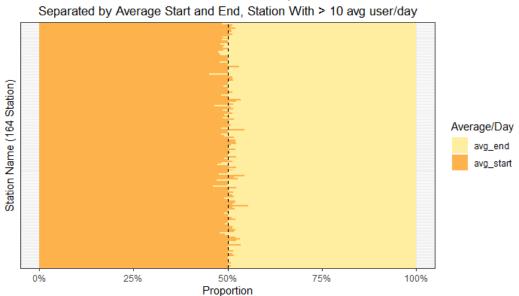




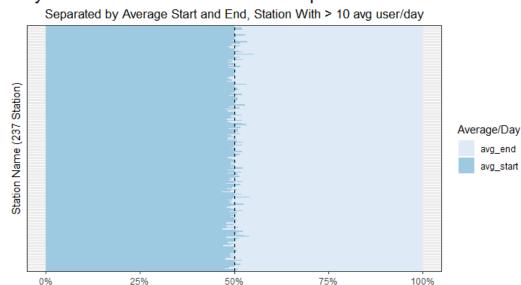
- There are high number of bike user for casual rider and cyclistic member at Zone 1 and 2
- From this density map, we can infer that between casual riders and cyclistic members there are similarities of the station used
 - **Zone 1** posses a good opportunity for cyclistic member marketing campaign



Casual Rider Start and End Proportion at Zone 1



Cyclistic Member Start and End Proportion at Zone 1



Proportion

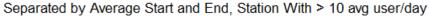


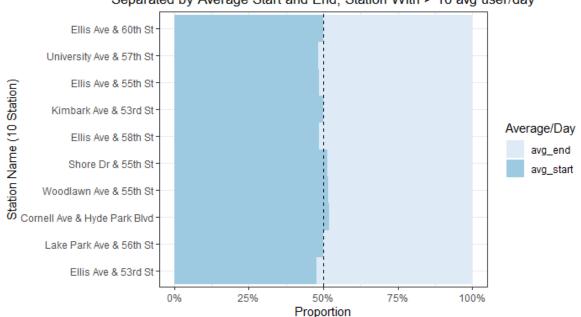
- Station with an average user more than 10 user/day in zone 1, has a 50:50 proportion between the user who starts using the bike and the user who finish using the bike in the same station
- In zone 1, behaviour of casual riders and cyclistic member was similar, they use the same station to start using the bike and end using the bike
- From this chart we could assume that there are fix number of casual rider and cyclistic member whom used the bikes for commuting in zone 1
 - Casual rider use **less station** compare to cyclistic member, **164 : 237 station**, for **commuting in zone 1**





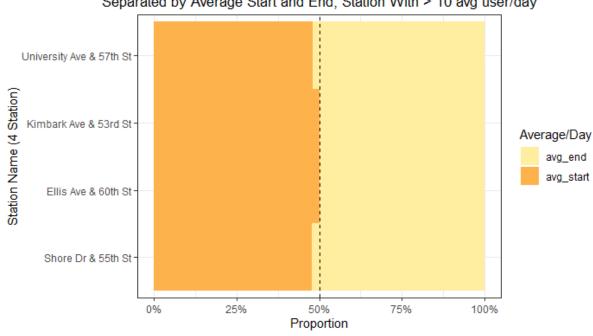
Cyclistic Member Start and End Proportion at Zone 2





Casual Rider Start and End Proportion at Zone 2





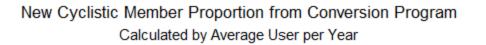


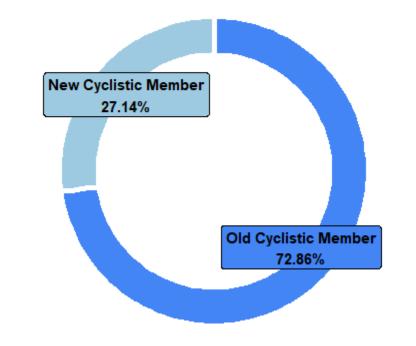
- Station with an average user more than 10 user/day in zone 2, also has a 50:50 proportion between the user who starts using the bike and the user who finish using the bike in the same station
- In zone 2, behaviour of casual riders and cyclistic member was similar, they use the same station to start using the bike and end using the bike
 - Casual rider use less station compare to cyclistic member, 4:10 station, for commuting in zone 2



	EDA
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Station Target Zone 1 (160 Station)	Station Target Zone 2 (4 Station)
Stretter Dr & Grand Ave	University Ave & 57th St
DuSable Lake Shore Dr & Monroe	Kimbark Ave & 53rd St
Millenium Park	Ellis Ave & 60th St
Michigan Ave & Oak St	Shore Dr & 55th St
DuSable Lake Shore Dr & North Blvd	
Shedd Aquarium	
Theater on the Lake	
Wells St & Concord Ln	







- For efficient marketing campaign, we could target 160 most visited station in zone 1 and 4 most visited station in zone 2 by commuted casual rider
- If we could target commuted casual rider at those station, We could get a new Cyclistic Member scheme comprises of 27,14% new member from casual rider and 72,86% Old Cyclistic Member. There are 37,3% increase in Cyclistic Member head count

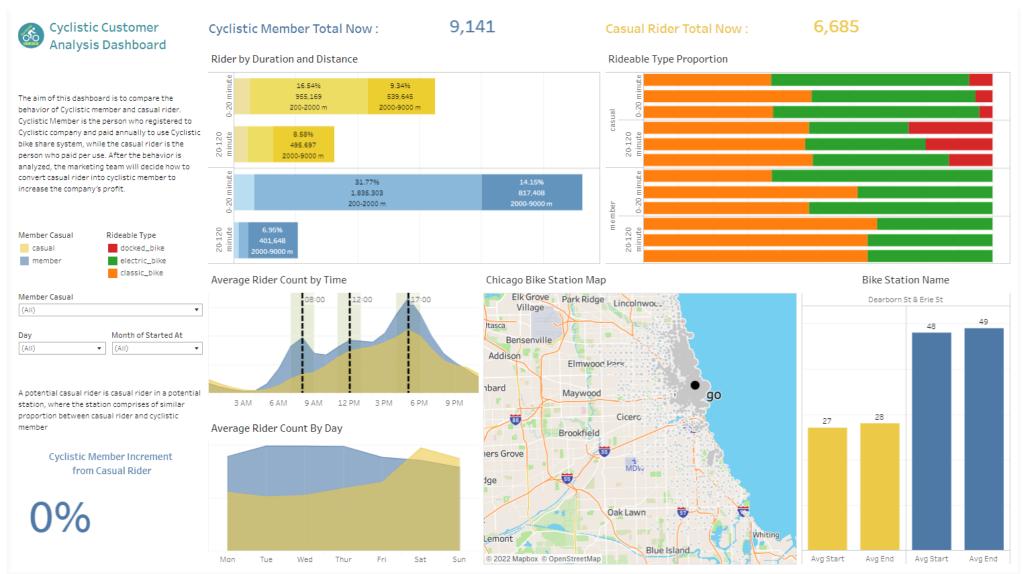


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Tableau Dashboard







A C T



Marketing Strategy

Stations

1

To increase the number of cyclistic members by the end of this quarter, there are **160 most-used stations in zone 1**, and **4 most-used stations in zone 2**. If we could convert all of the cyclistic members who usually use the bike for commuting at those stations, there will be a **37.3% increase in the amount of cyclistic members**.

Time

2

Run promotion program on weekdays when people commute in the morning (7:00-9:00 A.M), in lunchtime (11:00 AM - 1:00 PM), or in the afternoon (4:00-6:00 PM). Conduct also the promotion program on weekends every 12:00-6:00 PM because at this time there is much casual rider in Chicago.

Bike Type



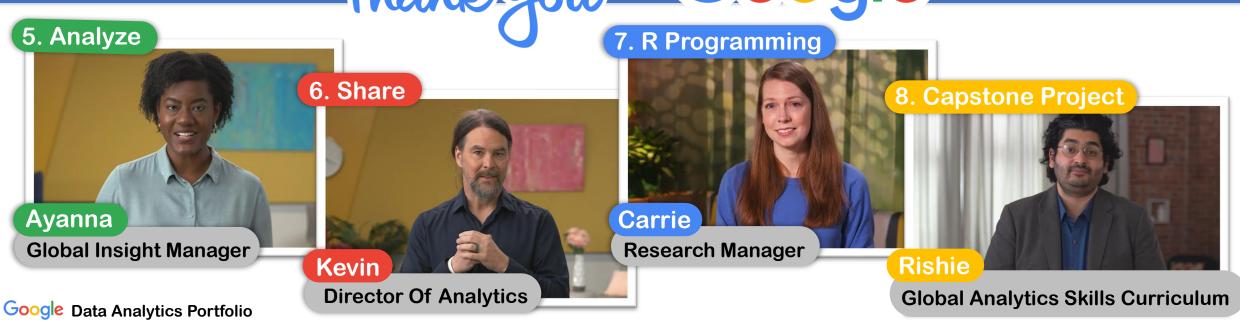
Never target **docked bike** user, just target casual riders who use **electric bikes** or **classic bikes**. First target the classic bike users, then target the electric bike users.

Duration and Distance



Target the casual rider who use the bike with a duration of 0-20 minute and distance 200-2000 m first, then the second target is the group with a duration 0-20 minutes and distance of 2000-9000 m, and the last target is the group with a duration of 20-120 minutes and a distance of 2000-9000 m.









8 Courses



Ask Questions to Make Data-Driven Decisions

Prepare Data for Exploration

Process Data from Dirty to Clean

Analyze Data to Answer Questions

Share Data Through the Art of Visualization

Data Analysis with R Programming

Google Data Analytics Capstone: Complete a Case Study



Oct 13, 2022

Stefanus Yudi Irwan

has successfully completed the online, non-credit Professional Certificate

Google Data Analytics

Those who earn the Google Data Analytics Professional Certificate have completed eight courses, developed by Google, that include hands-on, practice-based assessments and are designed to prepare them for introductory-level roles in Data Analytics. They are competent in tools and platforms including spreadsheets, SQL, Tableau, and R. They know how to prepare, process, analyze, and share data for thoughtful action.

The online specialization named in this certificate may draw on material from courses taught on-campus, but the included courses are not equivalent to on-campus courses. Participation in this online specialization does not constitute enrollment at this university. This certificate does not confer a University grade, course credit or degree, and it does not verify the identity of the learner.

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