

Google Data Analytics Professional Certification: Capstone Project - Cyclistic

Case Study: How does a Bike-Share Navigate Speedy Success?

### Introduction

Cyclistic is a fictional company that runs a bike-share program in Chicago that features more than 5,800 bicycles and 600 docking stations. As a junior data analyst working in the marketing analyst team at Cyclistic, we want to understand how casual riders and annual members use Cyclistic bikes differently. The senior management believes that the company's future success lies in maximizing the number of annual members. Therefore, a thorough analysis of the customer data and identifying possible trends will facilitate the decision-making process upon developing a new marketing strategy to attract more casual riders to purchase an annual membership.

The structure of this analysis follows the six-step data analysis process:

- 1) Ask
- 2) Prepare
- 3) Process
- 4) Analyze
- 5) Share
- 6) Act

For this project, I employ **Microsoft Excel spreadsheets** for the data cleaning process (Prepare, Process), **MySQL** to combine and analyze data (Analyze), and **Tableau** for the data visualization.

#### Ask

As instructed by management, three questions will guide the future marketing program, and this project's purpose is to answer the first question.

- 1. How do annual members and casual riders use Cyclistic bikes differently?
- 2. Why would casual riders buy Cyclistic annual memberships?
- 3. How can Cyclistic use digital media to influence casual riders to become members?

Business task: Analyze the historical bike trip data of the last twelve (12) months of Cyclistic, discover trends and insights regarding the bike usage of annual and casual riders. The key findings must be presented in visualizations to the stakeholders to facilitate the development of new digital marketing strategies to increase annual members.

Primary stakeholders: Director of marketing, Cyclistic executive team.

Secondary stakeholders: Cyclistic marketing analytics team

### **Prepare**

Dataset: 12 zipped ".csv" files from November 2020 to October 2021.

Data Source: Link (AWS cloud storage)

License: Available by Motivate International Inc.

The data files were downloaded, unzipped, and saved in a subfolder of the project's folder. Then, they opened and saved as ".xls" files in a separate subfolder to keep original and edited data files in different places. Each of the initial data files consisted of 13 columns, and the total rows of the dataset were 5,378,834.

Check the credibility and potential bias in data: ROCCC

- Reliability: The dataset contains complete and accurate data regarding bike rides as recorded by Divvy, a bicycle-sharing service of Chicago that collects data from the city of Chicago and makes it available to the public.
- Original: Data is collected directly from people in Chicago and is available to the public (second-party data).
- Comprehensive: The data includes for each ride its id (column A), bike type (B), starting time (C), ending time (D), starting station (E) and id (F), ending station (G), and id (H), starting latitude (I) longitude (J), ending latitude (K) longitude (L), and customer's category, casual OR member (M).
- Current: Dataset dates from November 2020 to October 2021.
- Cited: Data is appropriately cited, as we know who the author is, the data source, and the provided data license agreement.

Data Limitations: The data-privacy issues deter using riders' personally identifiable information, so it is impossible to connect bike passes purchased to credit card numbers to determine if casual riders live in the Cyclistic service area or if they have purchased multiple single passes. Thus, single, or numerous rides cannot be linked to individuals. In addition, the dataset contains some missing values within most of the files, which are located in the starting or ending stations name and id and ending latitude or longitude columns.

#### **Process**

Tool: Microsoft Excel Spreadsheet

The following columns were added to each file:

Column	Description	Formula
N: ride_length	Transacts ending and	=D2-C2
	starting time	
O: ride_length_m	Calculated total ride length	=(T2*24*60)+(HOUR(N2)*60)+MINUTE(N2)+SECOND(N2)/60
	in minutes	
T: ride_days	Calculates the days'	=DAYS(D2,C2)
	difference between ending	
	and starting time	
P: day_of_week	Returns 1 for Sunday and 7	=WEEKDAY(C2,1)
	for Saturday	
Q: day_name	Transforms integers into	=IF(P2=1,"Sunday",IF(P2=2,"Monday",IF(P2=3,"Tuesday",IF(P2=4,"W
	day names	ednesday",IF(P2=5,"Thursday",IF(P2=6,"Friday","Saturday"))))))
R: hour_of_day	Hour in a day from 0 to 23	=HOUR(C2)
S: part_of_day	Based on hour	=IF(AND(R2>=5,R2<=11),"Morning",IF(AND(R2>11,R2<=16),"Afterno
		on",IF(AND(R2>16,R2<=20),"Evening","Night")))
U: month_num	From 1 to 12	=MONTH(C2)
V: month	Name of month based on	=IF(U2=1,"January",IF(U2=2,"February",IF(U2=3,"March",IF(U2=4,"A
	integer	pril",IF(U2=5,"May",IF(U2=6,"June",IF(U2=7,"July",IF(U2=8,"August",
		IF(U2=9,"September",IF(U2=10,"October",IF(U2=11,"November","D
		ecember"))))))))))
W: season	Based on month	=IF(OR(V2="September",V2="October",V2="November"),"Autumn",I
		F(OR(V2="December",V2="January",V2="February"),"Winter",IF(OR(
		V2="March",V2="April",V2="May"),"Spring","Summer")))

I resaved all ".xls" files into ".csv" files in a different subfolder to keep track of all the steps during the data preparation and cleaning process in case of iterating any step. I checked all files for errors that could lead to problems during the

analysis phase. Then, I checked for errors or flaws within all the data files and removed all incomplete or problematic rows as shown below:

- ### value: Filter data in column "ride\_length\_m", type "#", select only the values that contain "#" from the dropdown list, and delete all problematic rows that appear in each file.
- O value: Filter data in column "ride\_length\_m", select only the values that contain "0" from the dropdown list, and delete all problematic rows that appear in each file.
- Missing values: Filter data in columns "end\_lat" or "end\_lng", type "" (blank), and delete all problematic rows from the dropdown list that appear in each file.

Then, I removed the following columns to filter out unnecessary data for my analysis:

ride\_id, started\_at, ended\_at, start\_station\_name, start\_station\_id, end\_station\_name, end\_station\_id, ride\_length, day\_of\_week, ride\_days, month\_num

(Note: I should have also removed columns with latitude and longitude as they hardly provided any insights for this particular project. Geographical data had no impact on this project to support developing a new digital media strategy, and I also knew all rides are referring to Chicago)

After the cleaning process took place, the data files ended up as follows:

File	Initial Count of rows	Rows with errors	Final Count of rows
202011	259716	1171	258545
202012	131573	551	131022
202101	96834	108	96726
202102	49622	218	49404
202103	228496	179	228317
202104	337230	305	336925
202105	531633	506	531127
202106	729595	783	728812
202107	822410	813	821597
202108	804352	813	803539
202109	756147	702	755445
202110	631226	554	630672
Total	5378834	6703	5372131

## **Analyze**

Tool: MySQL Workbench 8.0.27

I created a database named "cyclistic\_project" in MySQL Workbench.

## Create database

CREATE DATABASE cyclistic\_project;

USE cyclistic\_project;

Create table "trips" to combine all files into one table to increase speed

CREATE TABLE trips (

ride\_type VARCHAR(255),

start\_lat DECIMAL(10,8),

start\_Ing DECIMAL(11,8),

end\_lat DECIMAL(10,8),

end\_lng DECIMAL(11,8),

customer\_type VARCHAR(255),

ride\_minutes DECIMAL(10,2),

ride\_day VARCHAR(255),

ride\_hour INT,

part\_day VARCHAR(255),

ride\_month VARCHAR(255),

ride\_season VARCHAR(255));

Import each file into the table

LOAD DATA INFILE "C:/ProgramData/MySQL/MySQL Server 8.0/Uploads/trips\_202011.csv"

**INTO TABLE trips** 

FIELDS TERMINATED BY "," ENCLOSED BY ""

LINES TERMINATED BY "\n"

IGNORE 1 ROWS;

(Same for the rest of the ".csv" files, just replacing the name of the file: trips\_202012, trips\_202101, trips\_202102, trips\_202103, trips\_202104, trips\_202105, trips\_202106, trips\_202107, trips\_202108, trips\_202109, trips\_202110)

Check if all imported data is correct by comparing COUNT(\*) with csv's rows and the sum of ride\_length\_m with SUM(ride\_min). All query's results are exported as a ".csv" file and displayed as a table after each query.

SELECT COUNT(\*) AS Total\_Rows,

SUM(ride\_min) AS Total\_Ride\_Length\_Minutes

FROM trips;

Imported Data from clean CSV files			
File	Number of Rows	Sum of ride_length_m	
202011	258,545	6,193,115.54	
202012	131,022	2,509,246.32	
202101	96,726	1,790,979.15	
202102	49,404	1,386,560.88	
202103	228,317	6,650,809.39	
202104	336,925	11,282,440.56	
202105	531,127	20,124,525.89	
202106	728,812	28,948,667.37	
202107	821,597	29,566,472.18	
202108	803,539	24,948,362.54	
202109	755,445	22,043,442.89	
202110	630,672	15,712,215.39	
Total	5,372,131	171,156,838.10	

**Summary statistics** 

**SELECT** 

customer\_type,

COUNT(\*) AS total\_num,

MAX(ride\_minutes) AS Longest\_ride,

MIN(ride\_minutes) AS Shortest\_ride,

AVG(ride\_minutes) AS Average\_ride

**FROM trips** 

**GROUP BY 1** 

ORDER BY 2 DESC;

customer_type	total_num	Longest_ride	Shortest_ride	Average_ride
member	2905532	2873.73	0.02	18.229471
casual	2466599	57131.68	0.02	47.916393

**SELECT** 

ride\_type,

COUNT(\*) AS total\_type,

MAX(ride\_minutes) AS Longest\_ride,

MIN(ride\_minutes) AS Shortest\_ride,

AVG(ride\_minutes) AS Average\_ride

FROM trips

**GROUP BY 1** 

ORDER BY 5 DESC;

ride_type	total_type	Longest_ride	Shortest_ride	Average_ride
docked_bike	462744	57131.68	0.02	84.075066
classic_bike	3062390	2939.93	0.02	27.633536
electric_bike	1846997	1927.27	0.02	25.78615

## Top Days by Customer's category

(SELECT

customer\_type,

ride\_day,

COUNT(ride\_day) AS num\_days

FROM trips

**GROUP BY 1,2** 

HAVING customer\_type = "casual"

ORDER BY 3 DESC LIMIT 1)

UNION

(SELECT

customer\_type,

ride\_day,

COUNT(ride\_day) AS num\_days

**FROM trips** 

**GROUP BY 1,2** 

HAVING customer\_type = "member"

ORDER BY 3 DESC LIMIT 1);

customer_type	ride_day	num_days
casual	Saturday	550971
member	Wednesday	444026

### Top bike type per customer's category

(SELECT

customer\_type,

ride\_type,

COUNT(ride\_type) AS top\_ride\_type

FROM trips

GROUP BY 1,2

HAVING customer\_type = "casual"

ORDER BY 3 DESC LIMIT 1)

UNION

(SELECT

customer\_type,

ride\_type,

COUNT(ride\_type) AS top\_ride\_type

FROM trips

**GROUP BY 1,2** 

HAVING customer\_type = "member"

ORDER BY 3 DESC LIMIT 1);

customer_type	ride_type	top_ride_type
casual	classic_bike	1223122
member	classic_bike	1839268

# Top ride day per customer type

(SELECT

customer\_type,

ride\_day,

COUNT(ride\_day) AS top\_ride\_day

FROM trips

**GROUP BY 1,2** 

HAVING customer\_type = "casual"

ORDER BY 3 DESC LIMIT 1)

UNION

(SELECT

customer\_type,

ride\_day,

COUNT(ride\_day) AS top\_ride\_day

FROM trips

GROUP BY 1,2

HAVING customer\_type = "member"

ORDER BY 3 DESC LIMIT 1);

customer_type	ride_day	top_ride_day
casual	Saturday	550971
member	Wednesday	444026

## Top part of day per customer type

(SELECT

customer\_type,

part\_day,

COUNT(part\_day) AS top\_part\_day

FROM trips

GROUP BY 1,2

HAVING customer\_type = "casual"

ORDER BY 3 DESC LIMIT 1)

UNION

(SELECT

customer\_type,

part\_day,

COUNT(part\_day) AS top\_part\_day

FROM trips

**GROUP BY 1,2** 

HAVING customer\_type = "member"

ORDER BY 3 DESC LIMIT 1);

customer_type	part_day	top_part_day
casual	Afternoon	895184
member	Afternoon	952018

# Top month

(SELECT

customer\_type,

ride\_month,

COUNT(ride\_month) AS top\_ride\_month

FROM trips

GROUP BY 1,2

HAVING customer\_type = "casual"

ORDER BY 3 DESC LIMIT 1)

UNION

(SELECT

customer\_type,

ride\_month,

COUNT(ride\_month) AS top\_ride\_month

FROM trips

**GROUP BY 1,2** 

HAVING customer\_type = "member"

ORDER BY 3 DESC LIMIT 1);

customer_type	ride_month	top_ride_month
casual	July	441428
member	September	392028

Author: Evangelos Bikis Date: November 20, 2021 Top season per customer type (SELECT customer\_type, ride\_season, COUNT(ride\_season) AS top\_season FROM trips GROUP BY 1,2 HAVING customer\_type = "casual" ORDER BY 3 DESC LIMIT 1) **UNION** (SELECT customer\_type, ride\_season, COUNT(ride\_season) AS top\_season

FROM trips

GROUP BY 1,2

HAVING customer\_type = "member"

ORDER BY 3 DESC LIMIT 1);

customer_type	ride_season	top_season
casual	Summer	1223586
member	Summer	1130362

## Export the table with all data into one file

SELECT \*

INTO OUTFILE 'C:/ProgramData/MySQL/MySQL Server 8.0/Uploads/Tripsclean.csv'

FIELDS TERMINATED BY ','

ENCLOSED BY ""

ESCAPED BY '\\'

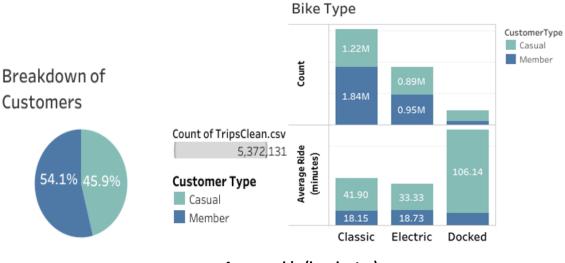
LINES TERMINATED BY '\n'

FROM trips;

#### **Share**

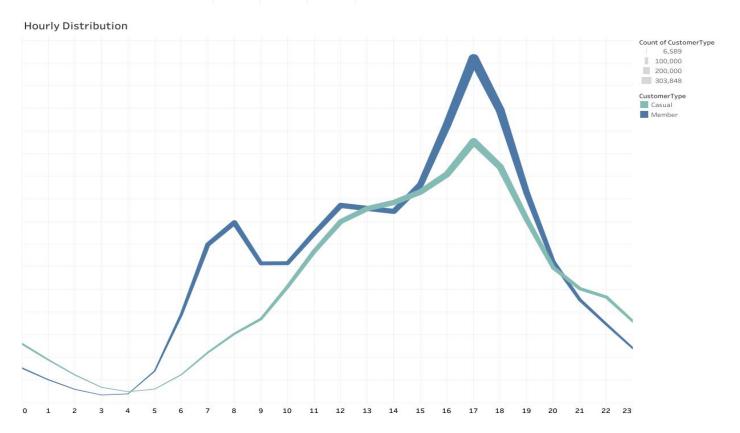
Tool: Tableau Public

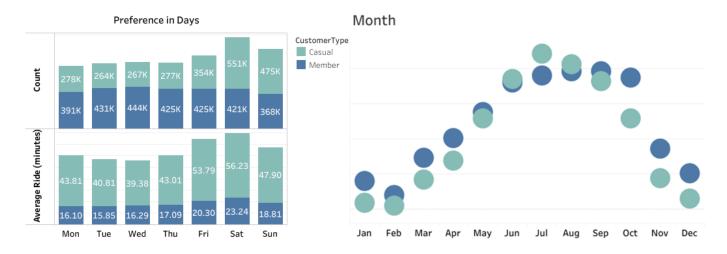
Since I created the ".csv" file with all the cleaned and ready to go for analysis data, I imported it to Tableau and created data visualizations.

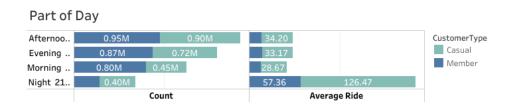


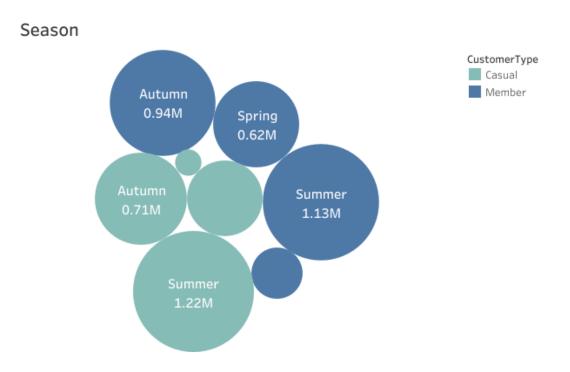
# **Average ride (in minutes)**



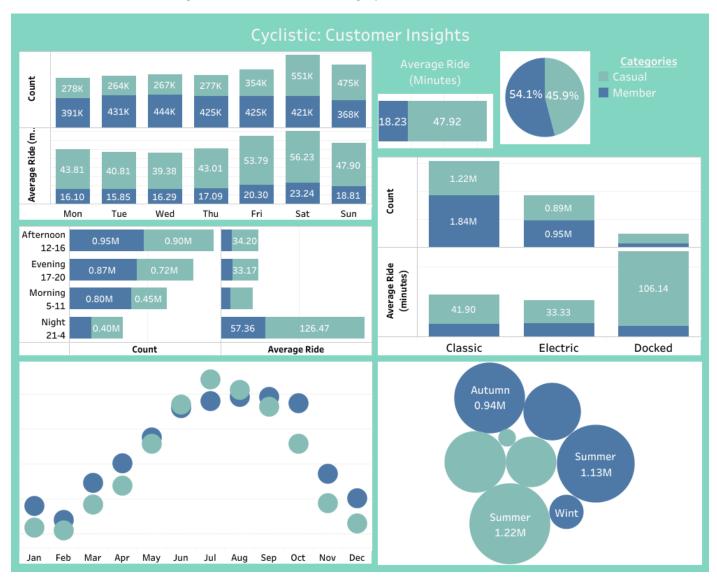




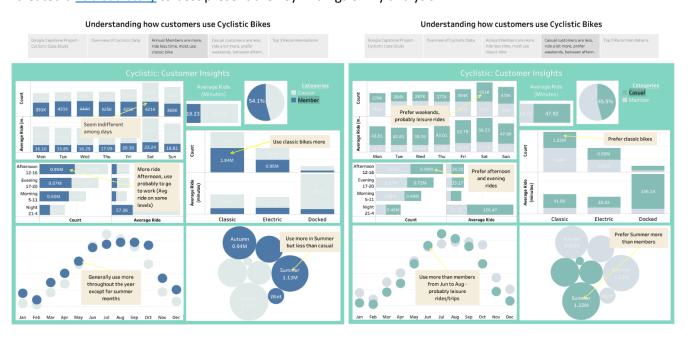




## Tableau Dashboard combining relevant data visualization graphs



I created a <u>Tableau story</u> to best present the key findings of my analysis.



#### Act

Based on the key findings of the analysis, the decision-makers are to form a new digital media marketing strategy to convert casual riders to annual members.

My top three recommendations are:

- 1) Give one more bike free during weekends to motivate leisure rides with friends or family.
- 2) Set milestones to members for long rides and reward with discounts or benefits.
- 3) Introduce seasonal memberships only for summer months to attract casual riders that don't want an annual membership.

# Understanding how customers use Cyclistic Bikes

Google Capstone Project - Overview of Cyclistic Data Annual Members are more, Cyclistic Case Study

Annual Members are more, ride less time, most use classic bike

Casual customers are less, ride a lot more, prefer weekends, between aftern..

