

Case Study of Cyclistic Bike Sharing Company

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Introduction

This case study is the Google Data Analytics Capstone Project that can be found on [coursera.org](https://www.coursera.org) and is part of the Google Data Analytics Professional Certificate course. This case study is for the fictitious Bike Sharing company called Cyclistic based on the Chicago Divvy Bike sharing company. The data analysis will follow the 6-step process of Ask, Prepare, Process, Analyze, Share, and Act. Along the way, I will be referring to the Case Study Roadmap guiding questions and key tasks.

Background

In 2016, Cyclistic launched a successful bike-share offering. Since then, the program has grown to a fleet of over 6,000 bicycles that are geo-tracked and locked into a network of over 1500 stations across Chicago. The bikes can be unlocked from one station and returned to any other station in the system anytime. Customers who purchase single-ride or full-day passes are referred to as casual riders. Customers who purchase annual memberships are Cyclistic members.

The Director of Marketing, Lily Moreno, has set a clear goal: Design marketing strategies aimed at converting casual riders into annual members. In order to do that, the marketing analyst team needs to better understand how annual members and casual riders differ, why casual riders would buy a membership, and how digital media could affect their marketing tactics. Moreno and her team are interested in analyzing the Cyclistic historical bike trip data to identify trends. Three questions will guide the future marketing program:

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

Moreno has assigned me the first question to answer: How do annual members and casual riders use Cyclistic bikes differently?

1. Ask

Following the case study roadmap's first section is "Ask", which refers to who is asking the questions and who are the major stakeholders.

Who is the Key Stakeholders: Lily Moreno

This case study asks the question: How do annual members and casual riders use the Cyclistic bikes differently?

2. Prepare

Following the case study roadmap's second section is "Prepare", which refers to the data. Where and how has the data been made available, how was it collected, and if not internal to the company, then under what license.

The data for this case study has been provided by the Motivate International Inc. under license from the Divvy Bikes and Scooters, LLC. Here are the links to the [data](#) and [license](#). I am using the data for the last 12 months - July 2022 to June 2023. For privacy all customer data was never included in the data files.

Downloaded the 12 monthly zip files and unzipped the CSV files into a separate folder, these files when unzipped ranged in size from 36MB to 141MB or 181 thousand rows to over 800 thousand rows. Each file was named with year-month-tripdata.csv. Each row in the file described a rental trip by either a member or casual customer with a starting date timestamp and ending date timestamp, along with station names and/or geo tracking Latitude and Longitude. At first look these files they were too large to combine into one spreadsheet with some missing data, like station names and geo data.

3. Process

Following the case study roadmap's third section is "Process", which refers to cleaning the data, including what steps taken and using what tools.

To start, I used Excel to do a preliminary pass through the data.

Created a column "ride length" by subtracting the ending at date timestamp from the starting at date timestamp. I noticed that some of these ride lengths were either zero or a negative number, so I eliminated those rows.

I also noticed that some of the rows were missing both the ending station names and the geo data, so I assumed that these bikes were taken out of service or stolen, I removed those rows.

Next, I wanted to fill in the missing station names by using the Geo data to locate the name of the station. To do this, created a separate sheet with the list of only station names and their Geo data (longitude and latitude). What I found was that each station did not have a single geo data location. It appears that the unit collecting the geo data was not exact even though some of the data went out 9 decimal points, so I reduced the data width to only 5 decimal places which helped align some of the data. I also collected multiple geo data for each station to allow for better matching of geo data to station names. I used a VLOOKUP function to fill in the missing station names from my master station name sheet to each of the month's spreadsheets.

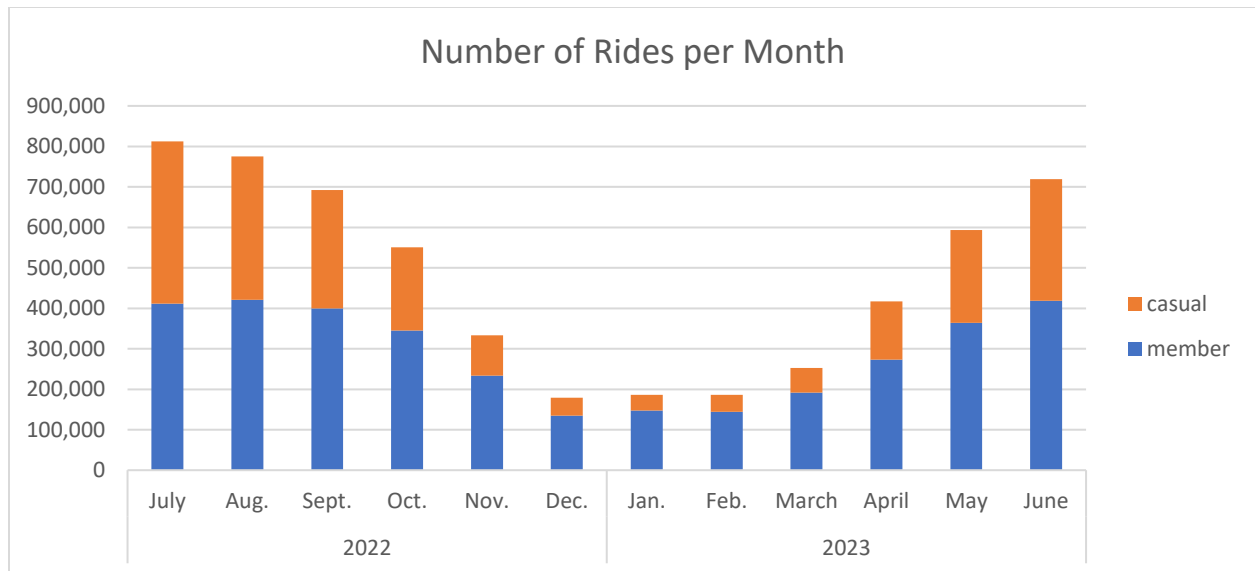
The spreadsheet was getting very large so I removed unnecessary columns, all the longitude and latitude fields, the station ID fields, and ride ID columns. I added a column describing the day of the week each ride took place on using the =WEEKDAY(C2,1) function for number day of week "nday_of_week". Saved the spreadsheet and repeated for the other 11 months of files.

Now to combine all the months together I used the software R Studio which was able to read in all twelve spreadsheets and combine them into one data frame which allowed me to run calculations on the full year's worth of data. Creating reports of how many rides took place by the two customer categories (member and casual) breaking out the by days of the week and hour of the day. Placing this data into other data frames and then saving them into CSV files to be analyzed in the next section.

For complete list of these steps used and those in the next analyze phase, please refer to separate document "Cyclistic Bike Sharing - Technical Steps .docx"

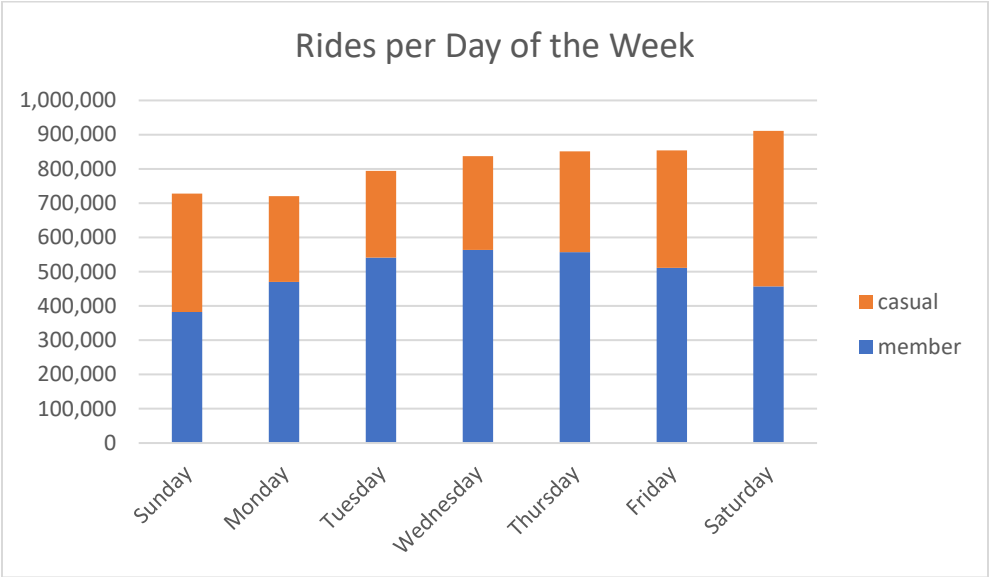
4. Analyze

When looking at the ridership you can see that both types of customers (Casual and Members) take less rides during the cold winter months.



Rides per Month			
	Casual	Member	Total
2022	1395804	1945634	3341438
July	400381	411877	812258
Aug.	353684	421225	774909
Sept.	292561	399359	691920
Oct.	205876	344738	550614
Nov.	99188	233761	332949
Dec.	44114	134674	178788
2023	816356	1538265	2354621
Jan.	39206	147157	186363
Feb.	42177	144370	186547
March	60861	191871	252732
April	144017	272902	416919
May	229677	363717	593394
June	300418	418248	718666
Grand Total	2212160	3483899	5696059

The Customers use the bikes differently during the week, with Member customers taking the most rides mid-week and the least over the weekend. The Casual customers are the opposite with the weekend (Friday, Saturday, and Sunday) having the most rides.



Rides per Day of the Week			
	Casual	Member	Totals
Sunday	346057	382167	728224
Monday	249541	470690	720231
Tuesday	253302	540953	794255
Wednesday	273294	563678	836972
Thursday	294188	557418	851606
Friday	342291	511378	853669
Saturday	453487	457615	911102
Grand Total	2212160	3483899	5696059

Heatmap Legend

Most

In Between

Least

Looking at the Grand Totals of 5,696,059 trips of which Casual customers took 2,212,160 rides (39%) and Member customers took 3,483,899 rides (61%).

My data has found that the busiest time for Casual customers is Saturday at 3pm and Saturday is also the busiest day.

Casual Rides per Hour

	Sun.	Mon.	Tues.	Wed.	Thurs.	Fri.	Sat.
0am	13527	3454	2597	2789	3522	5228	11349
1am	8945	2021	1509	1554	1942	3221	8541
2am	5949	1355	891	925	1186	1861	4769
3am	3124	1013	646	665	802	1180	2407
4am	1813	915	720	729	674	841	1163
5am	1663	1693	1862	1958	1798	1864	1322
6am	2499	4127	5352	5254	5363	6334	2476
7am	3960	6863	9675	9873	9808	8698	4568
8am	6647	9045	11912	12475	12700	10073	9121
9am	12263	8340	8386	8813	9122	9798	15369
10am	18622	9804	8716	8629	9563	11833	22872
11am	24069	12475	10719	11361	11898	15763	29385
12pm	27075	15485	13179	13205	14659	19863	33952
1pm	28436	16270	13515	13680	15239	21058	35911
2pm	29820	17104	14356	15004	16720	23131	36872
3pm	31256	18916	17420	17949	20213	27362	37926
4pm	29213	21908	23185	24618	26039	30578	37221
5pm	25160	25337	29033	31821	32963	33216	34581
6pm	21173	22247	24698	27135	29564	30055	31020
7pm	16001	17012	18050	20676	21775	22568	23948
8pm	11484	12395	12744	14970	15428	16346	18344
9pm	9671	9654	11144	13247	13321	13803	16955
10pm	8208	7322	8198	9947	12243	14345	17235
11pm	5479	4786	4795	6017	7646	13272	16180
Total	346057	249541	253302	273294	294188	342291	453487

Heatmap Legend



Most



In Between



Least

The busiest time for Member customers is Wednesday at 5pm and is also the busiest day of the week.

It is also noticeable that a lot of the Members are using the bike to commute to and from work with rush hours of 7am and 8am along with 4, 5, 6 pm.

Member Rides per Hour

	Sun.	Mon.	Tues.	Wed.	Thurs.	Fri.	Sat.
0am	9822	2979	2580	3019	3619	4786	9219
1am	6634	1647	1374	1456	1790	2702	6431
2am	4143	1066	753	785	1062	1504	3443
3am	2311	784	613	577	702	1012	1994
4am	1485	1204	1287	1154	1205	1299	1156
5am	1892	4933	6319	6277	5733	5138	1922
6am	4156	15351	20416	19365	18359	15942	5003
7am	7160	27323	38449	37866	35807	26176	9084
8am	10710	33011	45627	44424	43756	30049	16156
9am	16883	19679	23921	24584	25056	20814	22893
10am	23234	16378	17698	18477	18691	18716	28355
11am	28006	19835	20444	21750	21887	22891	32317
12pm	29665	23380	24475	25809	26347	27786	33772
1pm	30190	23488	23475	25566	25593	29016	33508
2pm	30042	23917	23239	25295	25481	30433	33930
3pm	31203	29790	32673	33464	34174	38258	33644
4pm	30373	43465	52118	53670	50723	46581	33667
5pm	28459	55918	65708	68147	64693	51103	33695
6pm	25877	43783	48576	51002	50455	42828	31098
7pm	19812	30862	33609	36085	35273	30585	25090
8pm	14759	21261	22803	25047	23815	21062	18484
9pm	11522	15552	17599	19374	19303	16272	15813
10pm	8596	9690	11280	13288	15057	14686	14380
11pm	5233	5394	5917	7197	8837	11739	12561
Total	382167	470690	540953	563678	557418	511378	457615

Heatmap Legend



Most



In Between



Least

The busiest time for all customers is Wednesday at 5pm, but Saturdays are the busiest day.

Total Rides per Hour

	Sun.	Mon.	Tues.	Wed.	Thurs.	Fri.	Sat.
0am	23349	6433	5177	5808	7141	10014	20568
1am	15579	3668	2883	3010	3732	5923	14972
2am	10092	2421	1644	1710	2248	3365	8212
3am	5435	1797	1259	1242	1504	2192	4401
4am	3298	2119	2007	1883	1879	2140	2319
5am	3555	6626	8181	8235	7531	7002	3244
6am	6655	19478	25768	24619	23722	22276	7479
7am	11120	34186	48124	47739	45615	34874	13652
8am	17357	42056	57539	56899	56456	40122	25277
9am	29146	28019	32307	33397	34178	30612	38262
10am	41856	26182	26414	27106	28254	30549	51227
11am	52075	32310	31163	33111	33785	38654	61702
12pm	56740	38865	37654	39014	41006	47649	67724
1pm	58626	39758	36990	39246	40832	50074	69419
2pm	59862	41021	37595	40299	42201	53564	70802
3pm	62459	48706	50093	51413	54387	65620	71570
4pm	59586	65373	75303	78288	76762	77159	70888
5pm	53619	81255	94741	99968	97656	84319	68276
6pm	47050	66030	73274	78137	80019	72883	62118
7pm	35813	47874	51659	56761	57048	53153	49038
8pm	26243	33656	35547	40017	39243	37408	36828
9pm	21193	25206	28743	32621	32624	30075	32768
10pm	16804	17012	19478	23235	27300	29031	31615
11pm	10712	10180	10712	13214	16483	25011	28741
Total	728224	720231	794255	836972	851606	853669	911102

Heatmap Legend



Most



In Between



Least

The next question is which stations are the busiest? (For both starting and ending trip locations)

Top Ten Locations								Grand Total
	Sunday	Monday	Tuesday	Wednesday	Thursday	Friday	Saturday	
Grand Total	1330675	1326507	1464228	1538439	1560678	1555840	1658209	10434576
Streeter Dr & Grand Ave*	24714	16428	14418	14697	15969	20365	33686	140277
DuSable Lake Shore Dr & Monroe St*	15258	8869	7382	7572	8427	11238	20699	79445
DuSable Lake Shore Dr & North Blvd*	13294	9533	8833	9915	9982	9616	18124	79297
Michigan Ave & Oak St*	13593	8707	8560	8935	9671	11354	17472	78292
Wells St & Concord Ln*	10263	8804	9100	10063	10543	11264	14926	74963
Clark St & Elm St	9423	9126	10075	10420	10286	10288	10711	70329
Kingsbury St & Kinzie St	6504	10472	11455	11206	10716	10021	7773	68147
Millennium Park*	11164	8521	7130	7037	9040	10246	14221	67359
Theater on the Lake*	11944	7381	6763	6827	7208	8187	15383	63693
Halsted St & Fulton St	7031	7001	8521	9631	10343	9859	9255	61641

It is interesting that 7 of the top 10 locations are all along the edge of Lake Michigan*.

(Not near the lake: Clark St & Elm St, Kingsbury St & Kinzie St. Halsted St & Fulton St)

Where is everyone going?

Top Ten Casual Ending Locations	
Streeter Dr & Grand Ave*	54925
DuSable Lake Shore Dr & Monroe St*	28132
Michigan Ave & Oak St*	25122
Millennium Park*	24959
DuSable Lake Shore Dr & North Blvd*	23994
Theater on the Lake*	18319
Shedd Aquarium*	17262
Wells St & Concord Ln*	14592
Dusable Harbor*	13116
Clark St & Armitage Ave*	12683

For the casual riders the top ten ending station are all near Lake Michigan. Compared to the member riders few of their locations are near the lake.

Top Ten Member Ending Locations	
Kingsbury St & Kinzie St	25300
Clark St & Elm St	23270
Wells St & Concord Ln	23247
Clinton St & Washington Blvd	23154
University Ave & 57th St	21186
Loomis St & Lexington St	21080
Clinton St & Madison St	21054
Ellis Ave & 60th St	20364
Wells St & Elm St	19393
Broadway & Barry Ave	19162

In evaluating the data, I found a number of rides (274,715) that started and ended at the same stations, I call these rides “joy” rides. When compared to the total number of rides (5,696,059) this amounts to about 5 % of the total, the casual riders spent more time riding the bikes.

Breakdown of Joy Rides				
		Number of Joy Ride	Percentage of Total Rides	Average Ride Length in Minutes
casual		163,856	3%	34.7
member		110,859	2%	15.3
Totals	5,696,059	274,715	5%	26.9

Top Ten Locations for Casual Joy Rides

	Number of Ride	Average Ride Length in Minutes
Streeter Dr & Grand Ave*	9324	41.2
DuSable Lake Shore Dr & Monroe St*	6360	34.6
Michigan Ave & Oak St*	4148	46.1
Millennium Park*	3725	39.2
Montrose Harbor*	2640	49.6
Dusable Harbor*	2369	35.3
Shedd Aquarium*	2284	22.2
DuSable Lake Shore Dr & North Blvd*	2146	38.8
Theater on the Lake*	1949	42.7
Adler Planetarium*	1803	36.0

Top Ten Locations for Member Joy Rides

	Number of Ride	Average Ride Length in Minutes
Halsted St & Fulton St	1548	4.2
Streeter Dr & Grand Ave*	1285	23.1
Loomis St & Lexington St	1139	12.1
Michigan Ave & Oak St*	886	24.4
N Clark St & W Elm St	840	11.7
State St & 33rd St	766	12.5
DuSable Lake Shore Dr & North Blvd*	758	17.6
Clark St & Elm St	706	16.1
N Sheffield Ave & W Wellington Ave	697	4.4
W Armitage Ave & N Sheffield Ave	697	4.0

It appears that both riders like to go near the lake parks and go for a relaxing spin through the parks along the waterfront.

5. Share

This analysis discovered three differences between the casual customers and the member customers of the Cyclistic bike sharing community.

- Days used:
 - Casual riders used the bikes more on Friday, Saturday, and Sunday with Saturday being the busiest.
 - Member riders used the bikes more mid-week on Tuesday, Wednesday, and Thursday with Wednesday being the busiest.
- Locations used:
 - Casual riders used the bikes more going to and along the waterfront parks of Lake Michigan.
 - Member riders used the bikes in a mix of commuting around the city and at the waterfront parks.
- Time spent on the bikes:
 - Casual riders used the bikes for longer trips on average 21 minutes.
 - Member riders used the bikes for shorter trips on average 12 minutes.

6. Act

While this analysis uncovered some key insights, further analysis would be beneficial.

- Such as track riders to discover usage patterns of individual customers. The question is of the casual customers which ones are “tourists” enjoying the city for a short period of time and then never use the service again. By eliminating the tourist this would uncover the casual customers that reside in the city and would be the prime targets of any ad campaign. The difficulty of this added level of information would need to be done in a way to adhere with the privacy license.
- With most of the Casual customers using the bike sharing on the weekends, the marketing department could consider a weekend only membership. Along the same line of thought, offer a membership for 9 months instead of the full year allowing customers to not feel like they are wasting their membership during the winter months in Chicago.
- Continue tracking of the busiest stations to see if customers usage changes over time. For example, how busy are the stations near the Universities to see if new classes of students change the bike usage.