

## Cyclistic

```
installed.packages("tidyverse")
install.packages("tidyverse")
library("tidyverse")
install.packages("lubridate")
bike_share_202107 <- read.csv('202107-divvy-tripdata.csv')
library("lubridate")
install.packages("ggplot2")
library("ggplot2")
install.packages("dplyr")
library("dplyr")
```

Cleaned and renamed .CSV files locally. trip\_data\_jan, trip\_data\_feb, trip\_data\_mar, , trip\_data\_april, trip\_data\_may, trip\_data\_jun, trip\_data\_july, trip\_data\_aug, trip\_data\_sept, trip\_data\_oct, trip\_data\_nov, trip\_data\_dec.

Merged into one data set named: trip\_data:

```
trip_data <- bind_rows(trip_data_jan, trip_data_feb, trip_data_mar, trip_data_april,
trip_data_may, trip_data_jun, trip_data_july, trip_data_aug, trip_data_sept, trip_data_oct,
trip_data_nov, trip_data_dec)
```

Removed Duplicates:

```
> trip_data_no_dups <- trip_data[!duplicated(trip_data$ride_id), ]
> print(paste("Removed", nrow(trip_data) - nrow(trip_data_no_dups), "duplicated rows"))
[1] "Removed 209 duplicated rows"
```

Parsing Datetime Columns:

```
> trip_data_no_dups$started_at <- as.POSIXct(trip_data_no_dups$started_at, "%Y-%m-%d
%H:%M:%S")
> trip_data_no_dups$ended_at <- as.POSIXct(trip_data_no_dups$ended_at, "%Y-%m-%d
%H:%M:%S")
```

Create new column for ride length in minutes:

```
> trip_data_no_dups <- trip_data_no_dups %>%
+   mutate(ride_time_m = as.numeric(trip_data_no_dups$ended_at -
trip_data_no_dups$started_at) / 60)
> summary(trip_data_no_dups$ride_time_m)
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
-29049.97   7.32   13.17   23.46   24.08 55944.15
*notice outliers in min and max
```

Combine year and month into one column:

```
> trip_data_no_dups <- trip_data_no_dups %>%
+   mutate(year_month = paste(strftime(trip_data_no_dups$started_at, "%Y"),
+                                 "-",
+                                 strftime(trip_data_no_dups$started_at, "%m"),
+                                 paste("(", strftime(trip_data_no_dups$started_at, "%b"), ") ", sep="")))
> unique(trip_data_no_dups$year_month)
[1] "2020 - 08 (Aug)" "2020 - 07 (Jul)" "2020 - 09 (Sep)" "2020 - 10 (Oct)" "2020 - 11 (Nov)"
"2020 - 12 (Dec)" "2021 - 01 (Jan)"
[8] "2021 - 02 (Feb)" "2021 - 03 (Mar)" "2021 - 04 (Apr)" "2021 - 05 (May)" "2021 - 06 (Jun)"
"2021 - 07 (Jul)"
```

Create weekday column:

```
> trip_data_no_dups <- trip_data_no_dups %>%
+   mutate(weekday = paste(strftime(trip_data_no_dups$ended_at, "%u"), "-",
strftime(trip_data_no_dups$ended_at, "%a")))
> unique(trip_data_no_dups$weekday
+ )
[1] "4 - Thu" "3 - Wed" "2 - Tue" "1 - Mon" "5 - Fri" "7 - Sun" "6 - Sat"
```

Create start hour column:

```
> trip_data_no_dups <- trip_data_no_dups %>%
+   mutate(start_hour = strftime(trip_data_no_dups$ended_at, "%H"))
> unique(trip_data_no_dups$start_hour)
[1] "14" "15" "17" "08" "12" "16" "18" "11" "04" "05" "19" "13" "09" "20" "21" "01" "10" "07" "06"
"22" "03" "00" "02" "23"
```

Create new cleaned .csv file:

```
> trip_data_no_dups %>%  
+ write.csv("trip_data_clean.csv")
```

Analyze:

```
> trip_data <- trip_data_no_dups  
> head(trip_data)
```

	ride_id	rideable_type	started_at	ended_at	start_station_name	start_station_id
1	322BD23D287743ED	docked_bike	2020-08-20 18:08:14	2020-08-20 18:17:51	Lake Shore Dr & Diversey Pkwy	329
2	2A3AEF1AB9054D8B	electric_bike	2020-08-27 18:46:04	2020-08-27 19:54:51	Michigan Ave & 14th St	168
3	67DC1D133E8B5816	electric_bike	2020-08-26 19:44:14	2020-08-26 21:53:07	Columbus Dr & Randolph St	195
4	C79FBBBD412E578A7	electric_bike	2020-08-27 12:05:41	2020-08-27 12:53:45	Daley Center Plaza	81
5	13814D3D661ECADB	electric_bike	2020-08-27 16:49:02	2020-08-27 16:59:49	Leavitt St & Division St	658
6	56349A5A42F0AE51	electric_bike	2020-08-27 17:26:23	2020-08-27 18:07:50	Leavitt St & Division St	658

	end_station_name	end_station_id	start_lat	start_lng	end_lat	end_lng	member_casual
1	Clark St & Lincoln Ave	141	41.93259	-87.63643	41.91569	-87.63460	member
2	Michigan Ave & 14th St	168	41.86438	-87.62368	41.86422	-87.62344	casual
3	State St & Randolph St	44	41.88464	-87.61955	41.88497	-87.62757	casual
4	State St & Kinzie St	47	41.88409	-87.62964	41.88958	-87.62754	casual
5	Leavitt St & Division St	658	41.90299	-87.68377	41.90300	-87.68384	casual
6	Leavitt St & Division St	658	41.90302	-87.68373	41.90309	-87.68363	casual

	ride_length	ride_time_m	year_month	weekday	start_hour
1	577 secs	9.616667	2020 - 08 (Aug)	4 - Thu	14
2	4127 secs	68.783333	2020 - 08 (Aug)	4 - Thu	15
3	7733 secs	128.883333	2020 - 08 (Aug)	3 - Wed	17
4	2884 secs	48.066667	2020 - 08 (Aug)	4 - Thu	08

```

5 647 secs 10.783333 2020 - 08 (Aug) 4 - Thu 12
6 2487 secs 41.450000 2020 - 08 (Aug) 4 - Thu 14

```

Create Summary of Data:

```

> summary(trip_data)
  ride_id      rideable_type   started_at      ended_at
start_station_name
Length:4730872  classic_bike :1785514  Min.   :2020-08-01 00:00:01  Min.   :2020-08-01
00:04:41  Length:4730872
Class :character  docked_bike :1558141  1st Qu.:2020-10-03 08:45:45  1st Qu.:2020-10-03
09:08:11  Class :character
Mode :character  electric_bike:1387217  Median :2021-04-05 13:52:15  Median :2021-04-05
14:16:31  Mode :character
                        Mean   :2021-02-17 10:26:13  Mean   :2021-02-17 10:49:41
                        3rd Qu.:2021-06-15 05:56:21  3rd Qu.:2021-06-15 06:21:59
                        Max.   :2021-07-31 23:59:58  Max.   :2021-08-12 17:45:41

start_station_id  end_station_name  end_station_id   start_lat   start_lng   end_lat
end_lng
Length:4730872  Length:4730872  Length:4730872  Min.   :41.64  Min.   : -87.87  Min.
:41.51  Min.   : -88.07
Class :character  Class :character  Class :character  1st Qu.:41.88  1st Qu.: -87.66  1st
Qu.:41.88  1st Qu.: -87.66
Mode :character  Mode :character  Mode :character  Median :41.90  Median : -87.64
Median :41.90  Median : -87.64
                        Mean   :41.90  Mean   : -87.64  Mean   :41.90  Mean
: -87.64
                        3rd Qu.:41.93  3rd Qu.: -87.63  3rd Qu.:41.93  3rd
Qu.: -87.63
                        Max.   :42.08  Max.   : -87.52  Max.   :42.16  Max.   : -87.44
                        NA's   :5246  NA's   :5246

member_casual    date      month      day      year      day_of_week
ride_length
casual:2102054  Min.   :2020-08-01  Length:4730872  Length:4730872  Length:4730872
Sunday :723892  Length:4730872
member:2628818  1st Qu.:2020-10-03  Class :character  Class :character  Class :character
Monday :579158  Class :difftime
                        Median :2021-04-05  Mode :character  Mode :character  Mode :character
Tuesday :604193  Mode :numeric
                        Mean   :2021-02-16                        Wednesday:631066
                        3rd Qu.:2021-06-15                        Thursday :612640
                        Max.   :2021-07-31                        Friday  :692953

```

Saturday :886970

```
ride_time_m    year_month    weekday    start_hour
Min.   :-29049.97 Length:4730872 Length:4730872 Length:4730872
1st Qu.:  7.32 Class :character Class :character Class :character
Median : 13.17 Mode  :character Mode  :character Mode  :character
Mean   : 23.46
3rd Qu.: 24.08
Max.   : 55944.15
```

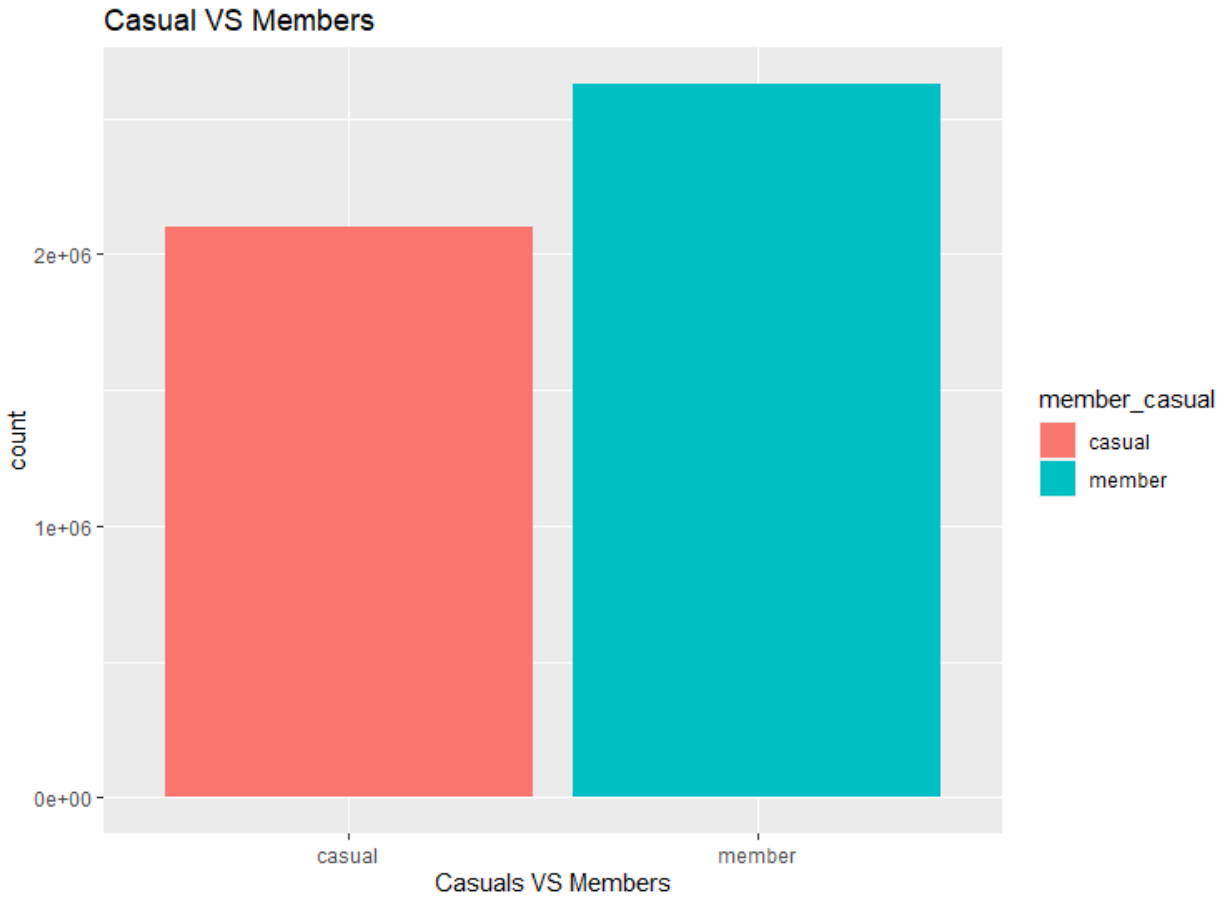
Visualize:

Casual vs Member Usage:

```
> trip_data %>%
+   group_by(member_casual) %>%
+   summarise(count = length(ride_id),
+             '%' = (length(ride_id) / nrow(trip_data)) * 100)
# A tibble: 2 x 3
  member_casual count  `%`
  <fct>        <int> <dbl>
1 casual      2102054 44.4
2 member      2628818 55.6
```

Count of Rides Taken per Type of User:

```
> fig(16,8)
> ggplot(trip_data, aes(member_casual, fill=member_casual)) +
+   geom_bar() +
+   labs(x="Casuals x Members", title="Chart 01 - Casuals x Members distribution")
> fig(16,8)
> ggplot(trip_data, aes(member_casual, fill=member_casual)) +
+   geom_bar() +
+   labs(x="Casuals x Members", title="Casuals VS Members")
> fig(16,8)
> ggplot(trip_data, aes(member_casual, fill=member_casual)) +
+   geom_bar() +
+   labs(x="Casuals VS Members", title="Casual VS Members")
> #Members account for 59% of rides
```



Usage by User per Month:

```
> trip_data %>%
+   group_by(year_month) %>%
+   summarise(count = length(ride_id),
+             '%' = (length(ride_id) / nrow(trip_data)) * 100,
+             'members_p' = (sum(member_casual == "member") / length(ride_id)) * 100,
+             'casual_p' = (sum(member_casual == "casual") / length(ride_id)) * 100,
+             'Member VS Casual Perc Difer' = members_p - casual_p)
```

# A tibble: 13 x 6

year_month	count	`%`	members_p	casual_p	`Member VS Casual Perc Difer`
<chr>	<int>	<dbl>	<dbl>	<dbl>	<dbl>
1 2020 - 07 (Jul)	1383	0.0292	20.7	79.3	-58.6
2 2020 - 08 (Aug)	621136	13.1	53.5	46.5	7.05
3 2020 - 09 (Sep)	532946	11.3	56.7	43.3	13.4
4 2020 - 10 (Oct)	389046	8.22	62.7	37.3	25.4
5 2020 - 11 (Nov)	259230	5.48	66.1	33.9	32.2
6 2020 - 12 (Dec)	131482	2.78	77.1	22.9	54.2

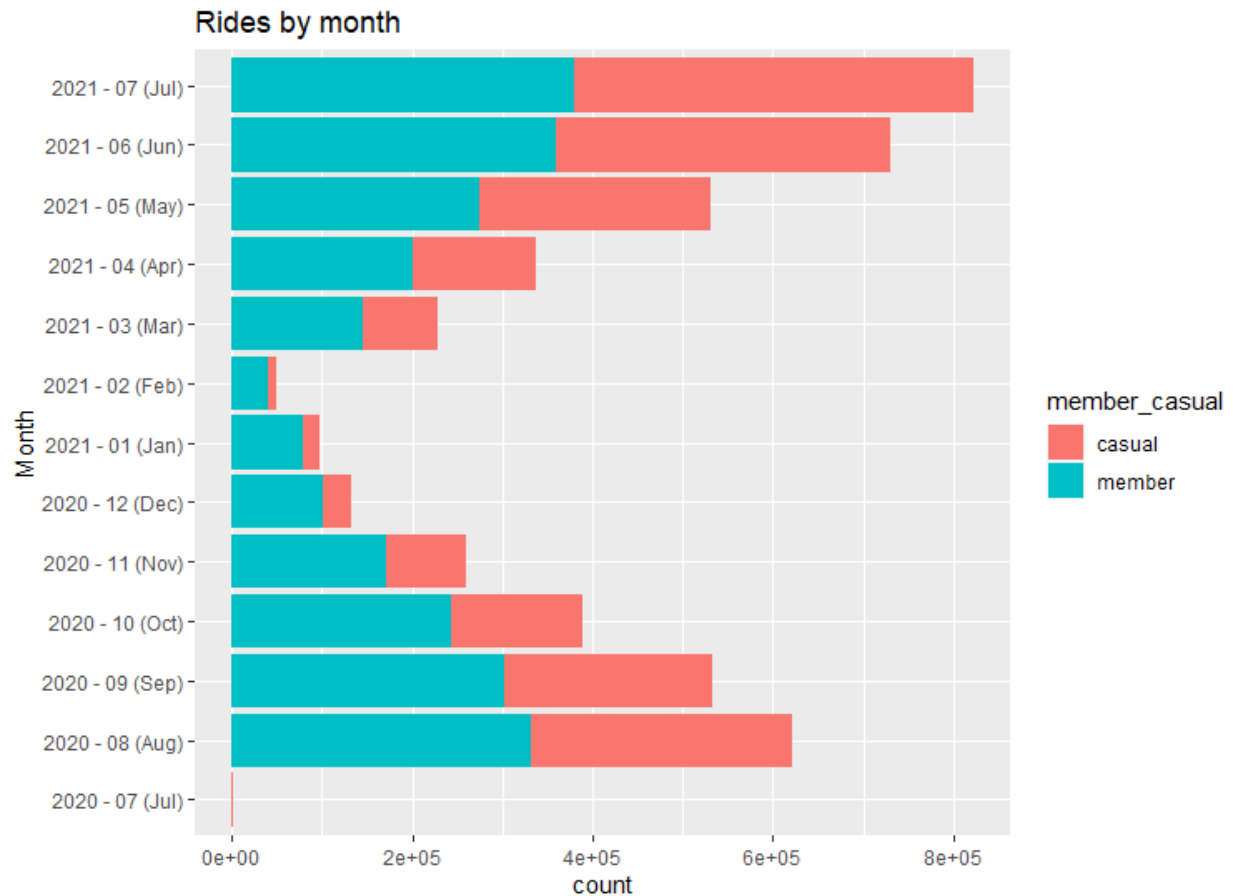
7	2021 - 01 (Jan)	96673	2.04	81.3	18.7	62.7
8	2021 - 02 (Feb)	49648	1.05	79.6	20.4	59.2
9	2021 - 03 (Mar)	228526	4.83	63.2	36.8	26.4
10	2021 - 04 (Apr)	337814	7.14	59.5	40.5	18.9
11	2021 - 05 (May)	531266	11.2	51.7	48.3	3.35
12	2021 - 06 (Jun)	729876	15.4	49.2	50.8	-1.63
13	2021 - 07 (Jul)	821846	17.4	46.3	53.7	-7.48

>

```

> trip_data %>%
+   ggplot(aes(year_month, fill=member_casual)) +
+   geom_bar() +
+   labs(x="Month", title="Chart 02 - Distribution by month")
> trip_data %>%
+   ggplot(aes(year_month, fill=member_casual)) +
+   geom_bar() +
+   labs(x="Month", title="Chart 02 - Distribution by month") +
+   coord_flip()
> trip_data %>%
+   ggplot(aes(year_month, fill=member_casual)) +
+   geom_bar() +
+   labs(x="Month", title="Rides by month") +
+   coord_flip()

```



\*Takeaway: Weather in Chicago impacts Usage

Rides by Day of Week:

```
> trip_data %>%
+   group_by(weekday) %>%
+   summarise(count = length(ride_id),
+             '%' = (length(ride_id) / nrow(trip_data)) * 100,
+             'members_p' = (sum(member_casual == "member") / length(ride_id)) * 100,
+             'casual_p' = (sum(member_casual == "casual") / length(ride_id)) * 100,
+             'Member x Casual Perc Difer' = members_p - casual_p)
# A tibble: 7 x 6
  weekday count  `%' members_p casual_p `Member x Casual Perc Difer`
  <chr>   <int> <dbl>   <dbl>   <dbl>         <dbl>
1 1 - Mon 575507 12.2    60.6    39.4         21.1
```



2	2 - Tue	604943	12.8	63.0	37.0	26.0
3	3 - Wed	632701	13.4	63.2	36.8	26.5
4	4 - Thu	616709	13.0	62.0	38.0	24.0
5	5 - Fri	721004	15.2	55.3	44.7	10.6
6	6 - Sat	889065	18.8	44.7	55.3	-10.5
7	7 - Sun	690943	14.6	46.4	53.6	-7.30

>

> #use coord\_flip again to view axis titles.

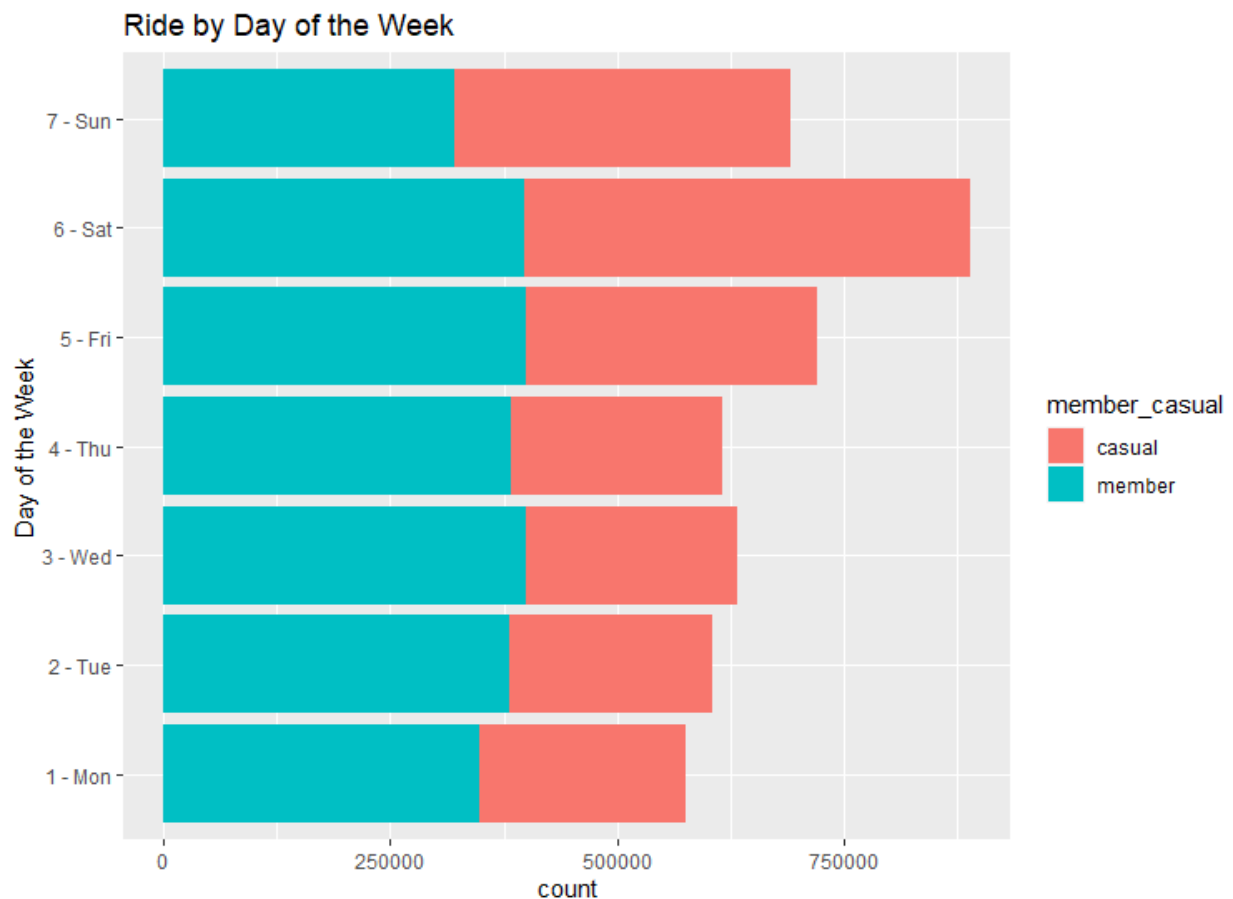
>

> ggplot(trip\_data, aes(weekday, fill=member\_casual)) +

+ geom\_bar() +

+ labs(x="Day of the Week", title="Ride by Day of the Week") +

+ coord\_flip()

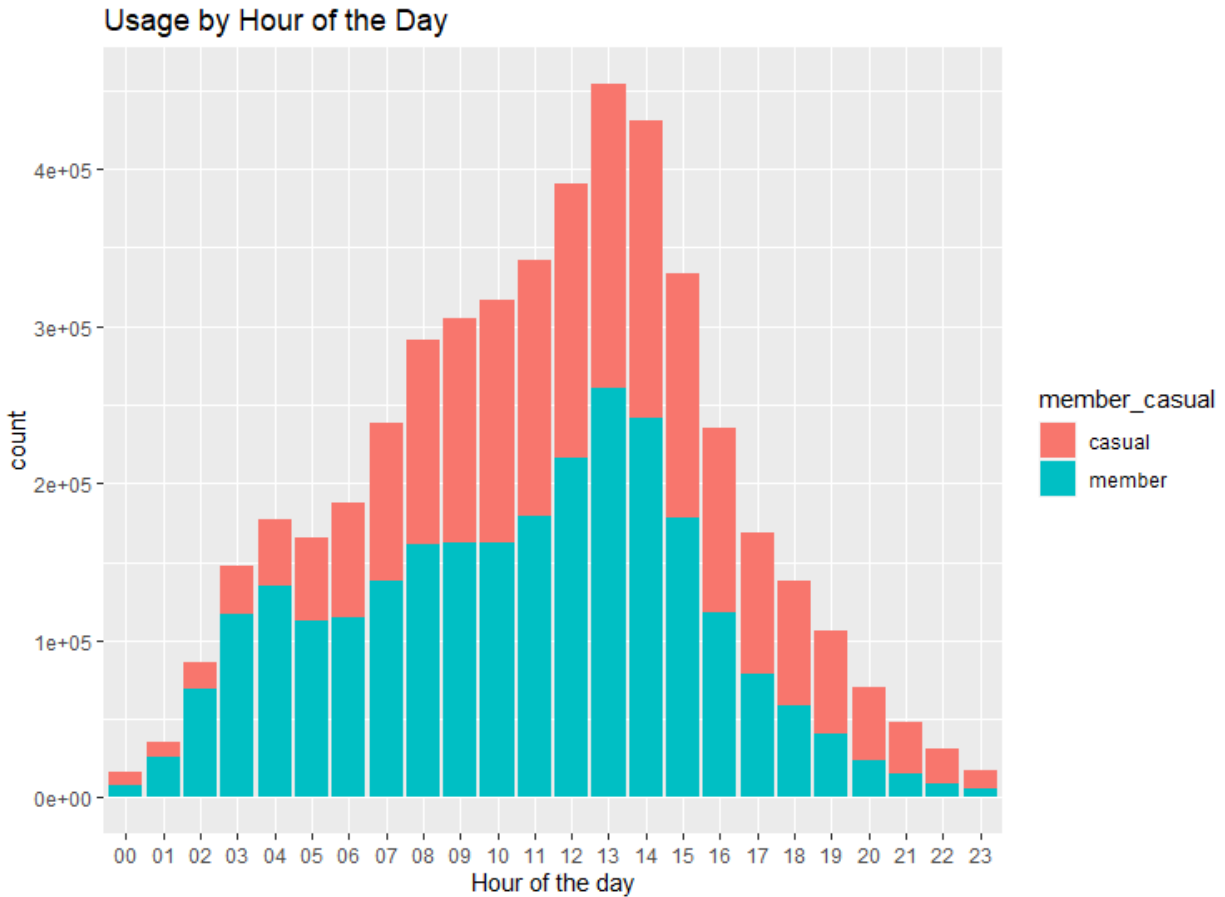


\*Takeaway: Members are more consistent across the week, use bikes more on weekdays. Saturday is by far the most popular day.

## Usage by Hour of the Day:

```
> trip_data %>%
+   group_by(start_hour) %>%
+   summarise(count = length(ride_id),
+             '%' = (length(ride_id) / nrow(trip_data)) * 100,
+             'members_p' = (sum(member_casual == "member") / length(ride_id)) * 100,
+             'casual_p' = (sum(member_casual == "casual") / length(ride_id)) * 100,
+             'member_casual_perc_difer' = members_p - casual_p)
# A tibble: 24 x 6
  start_hour count ` % ` members_p casual_p member_casual_perc_difer
  <chr>      <int> <dbl>   <dbl>   <dbl>          <dbl>
1 00        16087 0.340    49.5    50.5          -1.06
2 01        34851 0.737    75.1    24.9           50.3
3 02        86553 1.83     79.8    20.2           59.7
4 03       146921 3.11     79.6    20.4           59.1
5 04       177474 3.75     75.7    24.3           51.3
6 05       165275 3.49     68.3    31.7           36.6
7 06       187341 3.96     61.4    38.6           22.7
8 07       238752 5.05     57.8    42.2           15.6
9 08       290672 6.14     55.5    44.5           10.9
10 09      305313 6.45     53.0    47.0            5.95
# ... with 14 more rows
>

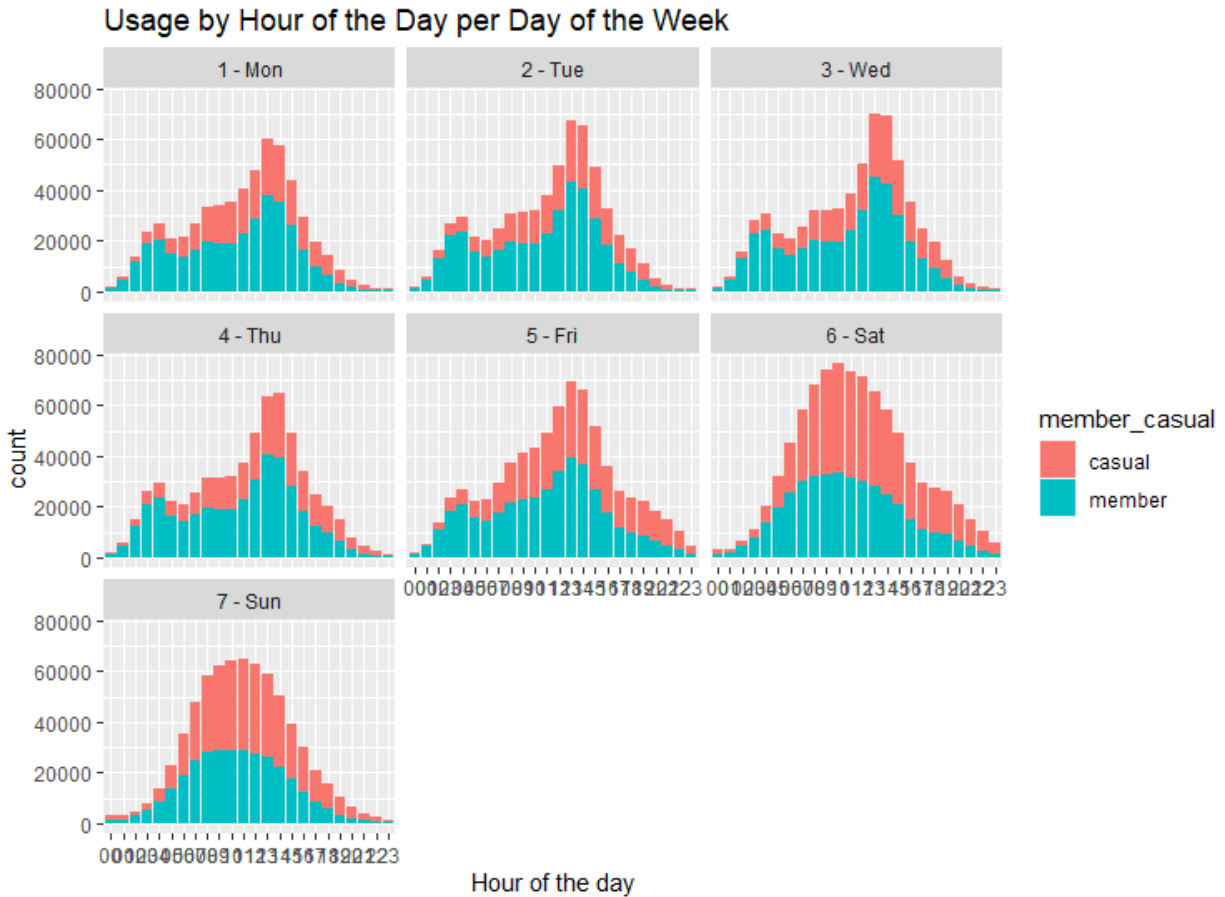
> trip_data %>%
+   ggplot(aes(start_hour, fill=member_casual)) +
+   geom_bar() +
+   labs(x="Hour of the day", title="Usage by Hour of the Day per Day of the Week") +
+   facet_wrap(~ weekday)
```



\*Takeaway: Members use bikes a bit at the start of workday, but definitely more at the end of the workday, assuming for commuting home or elsewhere, also perhaps for exercise.

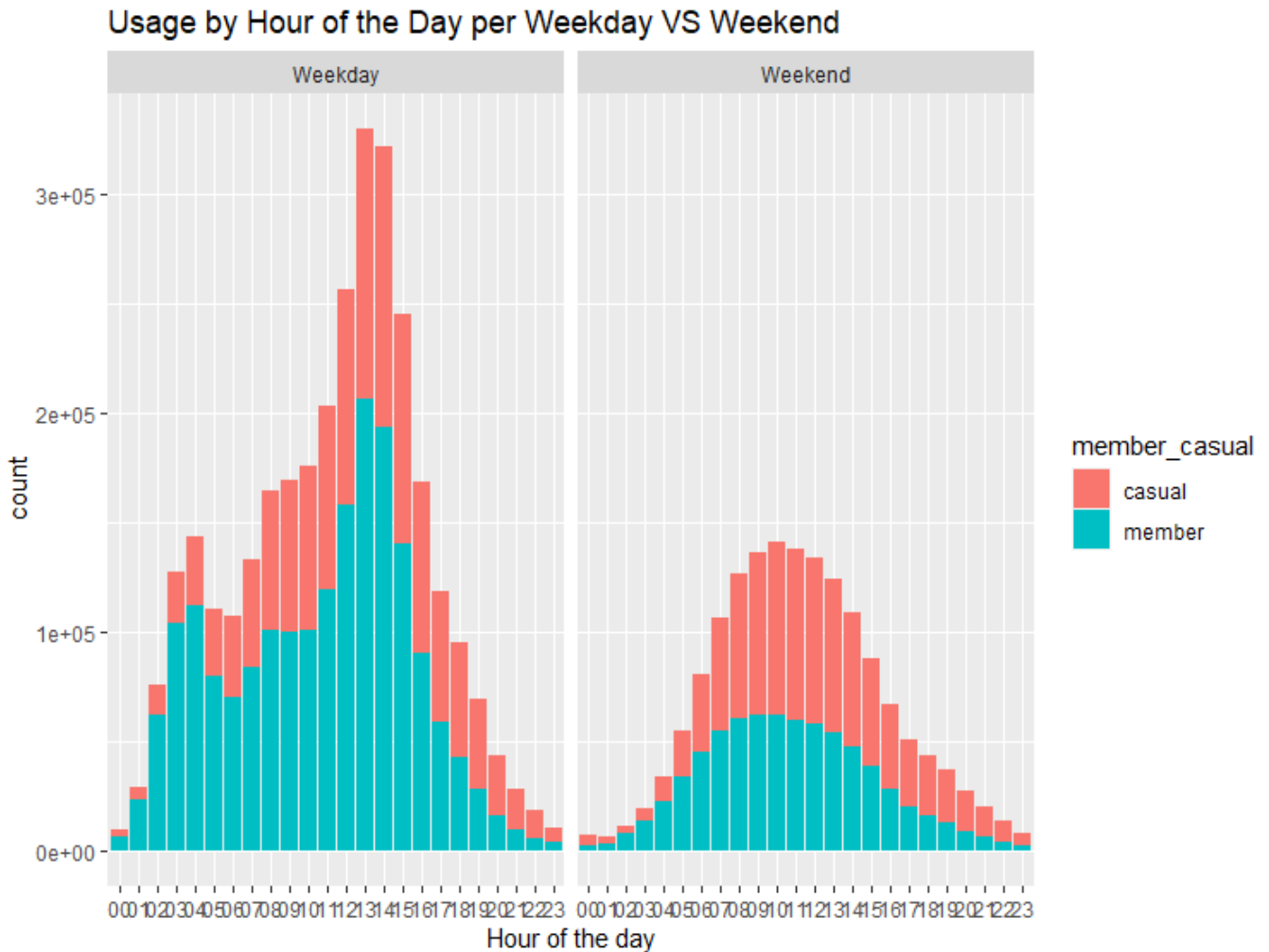
Usage per Hour of the Day per Day of the Week:

```
> trip_data %>%
+   ggplot(aes(start_hour, fill=member_casual)) +
+   geom_bar() +
+   labs(x="Hour of the day", title="Usage by Hour of the Day per Day of the Week") +
+   facet_wrap(~ weekday)
```



Comparing Usage Per Hour of the Day on Weekdays VS Weekends:

```
> trip_data %>%
+   mutate(type_of_weekday = ifelse(weekday == '6 - Sat' | weekday == '7 - Sun',
+                                   'Weekend',
+                                   'Weekday')) %>%
+   ggplot(aes(start_hour, fill=member_casual)) +
+   labs(x="Hour of the day", title="Usage by Hour of the Day per Weekday VS Weekend") +
+   geom_bar() +
+   facet_wrap(~ type_of_weekday)
>
```



[Apologies for the clustered data points on the X axis. A better storyboard will be added at a later date.]

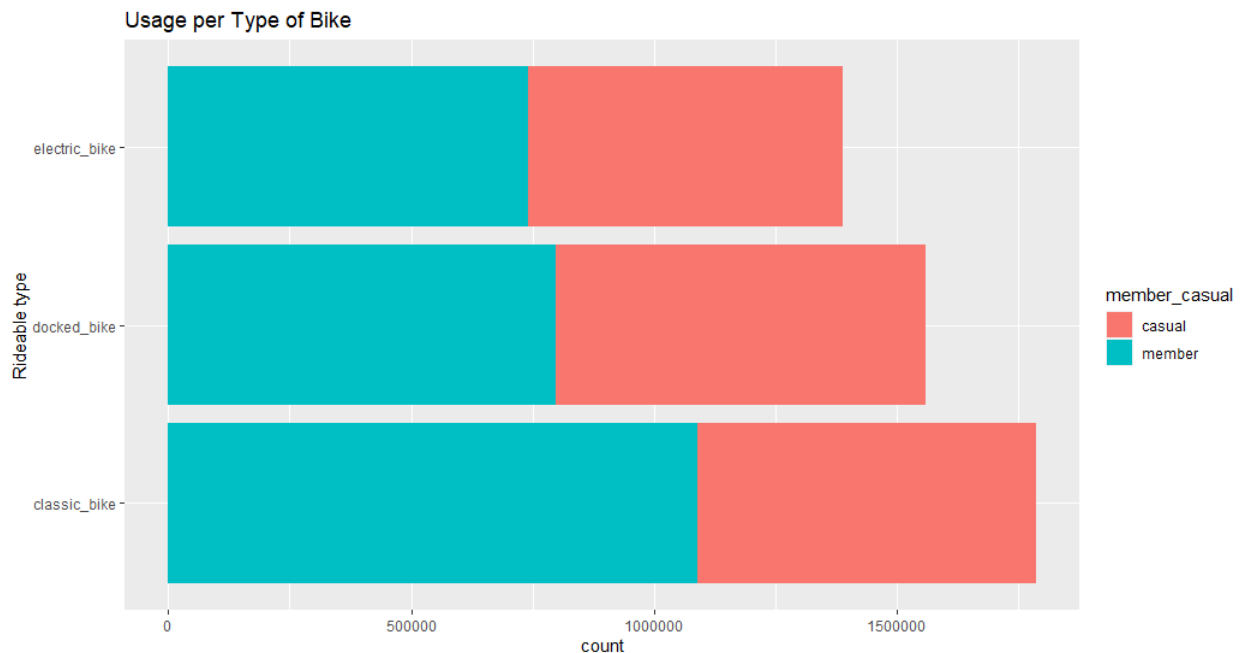
Types of Bikes Used:

```
> trip_data %>%
+   group_by(rideable_type) %>%
+   summarise(count = length(ride_id),
+             '%' = (length(ride_id) / nrow(trip_data)) * 100,
+             'members_p' = (sum(member_casual == "member") / length(ride_id)) * 100,
+             'casual_p' = (sum(member_casual == "casual") / length(ride_id)) * 100,
+             'member_casual_perc_difer' = members_p - casual_p)
# A tibble: 3 x 6
  rideable_type count  `%` members_p casual_p member_casual_perc_difer
<fct>         <int> <dbl>   <dbl>   <dbl>         <dbl>
1 classic_bike 1785514 37.7    61.1    38.9         22.1
```

```

2 docked_bike 1558141 32.9 51.2 48.8 2.41
3 electric_bike 1387217 29.3 53.4 46.6 6.80
>
>
> ggplot(trip_data, aes(rideable_type, fill=member_casual)) +
+ labs(x="Rideable type", title="Usage per Type of Bike") +
+ geom_bar() +
+ coord_flip()

```



Finally, I'll be honest here. The difficult part of taking online classes is not having anyone to ask for help. I felt I went above what was asked for a junior analyst role I stopped attempting to fix the outlier problem of the ride time. When I look at the summary it shows a negative number for the minimum (-29049.97 in minutes) and a huge number for the maximum (55944.15 in minutes). I spent a lot of time on StackOverflow attempting to fixed this and my best an final attempt still showed an error:

```

> #outliers
>
> summary(trip_data$ride_time_m)
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
-29049.97  7.32   13.17   23.46  24.08 55944.15
>
>
> #remove 5%
>

```

```

> trip_data_without_outliers <- trip_data %>%
+   filter(ride_time_m > as.numeric(ventiles['5%'])) %>%
+   filter(ride_time_m < as.numeric(ventiles['95%']))
Error: Problem with `filter()` input `..1`.
i Input `..1` is `ride_time_m > as.numeric(ventiles["5%"])`.
x object 'ventiles' not found
Run `rlang::last_error()` to see where the error occurred.

```

### Main Takeaways:

- Members use the bikes more often, but for a lesser amount of time. This shows they are using it for the purpose of commuting or exercising during the week at commute time and after work time.
- Members like the classic bike more, perhaps showing the desire for exercise.
- Casual riders use it for fun on weekends and near tourist locations (I spend a lot of time in Chicago). Will include that data in a nicer presentation.
- Members make up approximately 59% of the usage.
- July '21 was the most popular month
- Weather greatly affects usage. Not much can be done about that.
- Saturday is the most popular day
- More usage overall on weekends
- More usage in the afternoon.

Act:

In order to switch casual user to members:

Show a cost savings for members.

Target advertising for "being green".

Target advertising on the benefits of regular exercise.

Show hip, young, and healthy professionals using the bikes to avoid being stuck in traditional car traffic or waiting for the 'L' train.