R file

Processing data

Install and load packages to set the environment for processing and analysing Cyclistic data.

```
install.packages(c("tidyverse","dplyr","lubridate","ggplot2","skimr","janitor","rmarkdown"))
library(tidyverse)
library(dplyr)
library(lubridate)
library(ggplot2)
library(skimr)
library(janitor)
library(rmarkdown)
```

Import individual csv files and check for similarity in structure of each table.

```
jul22 <- read_csv("~/extracted files/202207-divvy-tripdata.csv")
jun22 <- read_csv("~/extracted files/202206-divvy-tripdata.csv")
may22 <- read_csv("~/extracted files/202205-divvy-tripdata.csv")
apr22 <- read_csv("~/extracted files/202204-divvy-tripdata.csv")
mar22 <- read_csv("~/extracted files/202203-divvy-tripdata.csv")
feb22 <- read_csv("~/extracted files/202202-divvy-tripdata.csv")
jan22 <- read_csv("~/extracted files/202201-divvy-tripdata.csv")
dec21 <- read_csv("~/extracted files/202112-divvy-tripdata.csv")
nov21 <- read_csv("~/extracted files/202111-divvy-tripdata.csv")
oct21 <- read_csv("~/extracted files/202110-divvy-tripdata.csv")
sep21 <- read_csv("~/extracted files/202109-divvy-tripdata.csv")
aug21 <- read_csv("~/extracted files/202108-divvy-tripdata.csv")</pre>
```

Union all the csv files into one data frame.

```
Yearly_Raw_Data <-
bind_rows(aug21,sep21,oct21,nov21,dec21,jan22,feb22,mar22,apr22,may22,jun22,jul22)
dim(Yearly_Raw_Data)</pre>
```

The output shows 13 columns with 5,901,463 rows in total.

Now adding new columns in order to make it easy to breakdown data to analyze. Use the date in order to break down the started_at datetime to date, month, year and hour of day.

```
Yearly_Raw_Data$date <- as.Date(Yearly_Raw_Data$started_at)
Yearly_Raw_Data$weekday <- format(as.Date(Yearly_Raw_Data$date),"%a")
Yearly_Raw_Data$month <- format(as.Date(Yearly_Raw_Data$date),"%m")
Yearly_Raw_Data$year <- format(as.Date(Yearly_Raw_Data$date),"%y")
Yearly_Raw_Data$starthour <- strftime(Yearly_Raw_Data$started_at,"%H")</pre>
```

Checking for any null values in dataframe.

```
sum(is.na(Yearly_Raw_Data))
```

~33% of data has NULL values.

```
colsums(is.na(Yearly_Raw_Data))
```

Output shows that majority of the NULLS are in start_station_name, start_station_id, end_station_name, end_station_id, end_lat, end_lng

Now, moving to remove NA from the Yearly_Raw_Data.

```
Yearly_Cleaned <- na.omit(Yearly_Raw_Data)
dim(Yearly_Cleaned)</pre>
```

Output has 4,629,230 rows.

Adding a new column - rideduration to the dataset. The ridedurationshall be in minutes.

```
Yearly_Cleaned <- Yearly_Cleaned %>%
  mutate(rideduration=difftime(ended_at,started_at,units="mins"))
```

Now checking rideduraiton data type

```
str(Yearly_Cleaned$rideduration)
summary(Yearly_Cleaned$rideduration)
is.numeric(Yearly_Cleaned$rideduration)
```

Output is FALSE for is.numeric

```
\label{thm:continuous} Yearly\_Cleaned\$rideduration <- as.numeric(as.character(Yearly\_Cleaned\$rideduration)) \\ is.numeric(YEarly\_Cleaned\$rideduration)
```

Output is TRUE and ready for use

```
summary(Yearly_Cleaned$rideduration)
```

Minimum is -129 and Maximum is 41,629. So, we need to omit rideduration <=0 and greater than 16 hours (i.e. 960 minutes)

```
Yearly_Cleaned <- Yearly_Cleaned[!(Yearly_Cleaned$rideduration<=0),]
Yearly_Cleaned <- Yearly_Cleaned[!(Yearly_Cleaned$rideduration>960),]
summary(Yearly_Cleaned$rideduration)
```

4,627,392 rows are still in the data Yearly_Cleaned after filtering data.

Analyzing data

Now, stepping into analysis of the cleaned data without visualizations

• Summary of ride count, mean ride duration and median ride duration

• Summary of ride count, mean ride duration and median ride duration by month of year

Output is a tibble 24 x 5.

• Keeping a proper order for the days of the week

```
Yearly_Cleaned$weekday <- ordered(Yearly_Cleaned$weekday,
levels=c("Mon","Tue","Wed","Thu","Fri","Sat","Sun"))
```

• Summary of ride count, mean ride duration and median ride duration based on day of week

Output is a tibble: 14 x 5

• Summary of ride count, mean ride duration and median ride duration based on hour of day

```
Yearly_Cleaned %>%
  group_by(member_casual,starthour) %>%
summarize(Number_of_rides = length(member_casual)
      , meanridedurationinsec = mean(rideduration)
      , medianridedurationinsec = median(rideduration)) %>%
print(n=48)
```

Output is a tibble 48 x 5

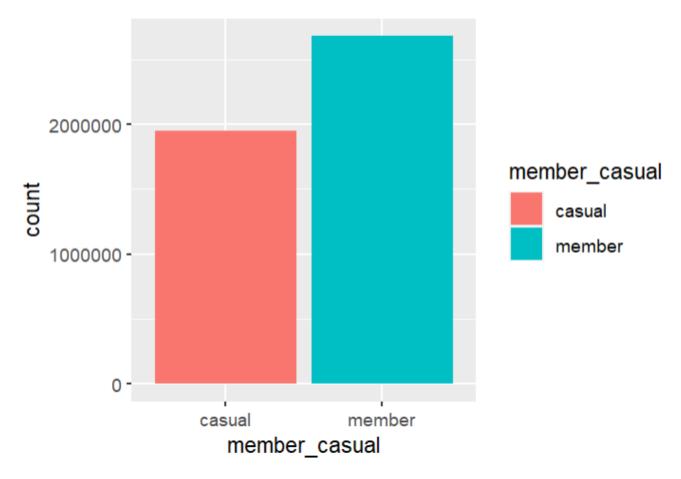
• Checking for rideable_type count, meanrideduration

```
Yearly_Cleaned %>%
    group_by(member_casual,rideable_type) %>%
    summarize(Number_of_rides_by_type = length(rideable_type)
               , meanridedurationinmin = mean(rideduration))
# A tibble: 5 × 4
# Groups: member_casual [2]
  member_casual rideable_type Number_of_rides_by_type meanridedurationinmin
  <chr>
            <chr>
                                                   <int>
                                                                           <dbl>
              classic_bike
docked_bike
electric_bike
classic_bike
electric_bike
1 casual2 casual
                                                1128313
                                                                            24.3
                                                  224026
                                                                           45.7
3 casual
                                                 595506
                                                                           18.2
4 member
                                                1919882
                                                                           13.0
5 member
                                                 759665
                                                                            11.3
```

Visualizing data in R

Based on number of rides

```
ggplot(data=Yearly_Cleaned) +
  geom_bar(mapping=aes(x=member_casual, fill=member_casual)) +
  scale_y_continuous(labels = function(x) format(x, scientific = FALSE))
```

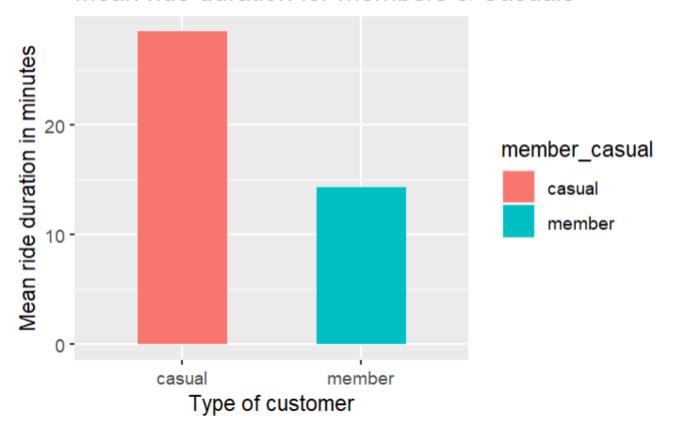


• Based on ride duration

```
Yearly_Cleaned %>%
  group_by(member_casual,weekday) %>%
  summarise(Ride_Duration=mean(rideduration),.groups="drop") %>%
  ggplot(aes(x=member_casual,y=Ride_Duration,fill=member_casual))+
  geom_col(width = 0.5,position = position_dodge(width = 0.5))+
```

```
labs(title="Mean ride duration for members & Casuals") +
xlab("Type of customer") +
ylab("Mean ride duration in minutes") +
scale_y_continuous(labels = function(x) format(x, scientific=FALSE))
```

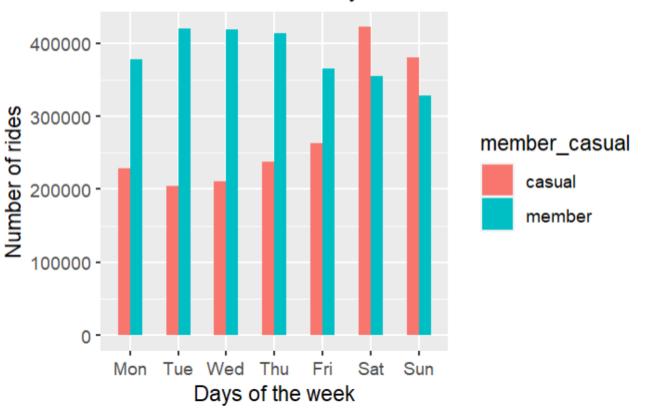
Mean ride duration for members & Casuals



• Break up of number of rides over the day of the week

```
Yearly_Cleaned %>%
  group_by(member_casual,weekday) %>%
  summarise(Number_of_rides=length(ride_id),.groups="drop") %>%
  arrange(member_casual,weekday) %>%
  ggplot(aes(x=weekday,y=Number_of_rides,fill=member_casual))+
  geom_col(width = 0.5,position = position_dodge(width = 0.5))+
  labs(title="Number of rides in day of week for members & Casuals") +
  xlab("Days of the week") +
  ylab("Number of rides") +
  scale_y_continuous(labels = function(x)format(x,scientific=FALSE))
```

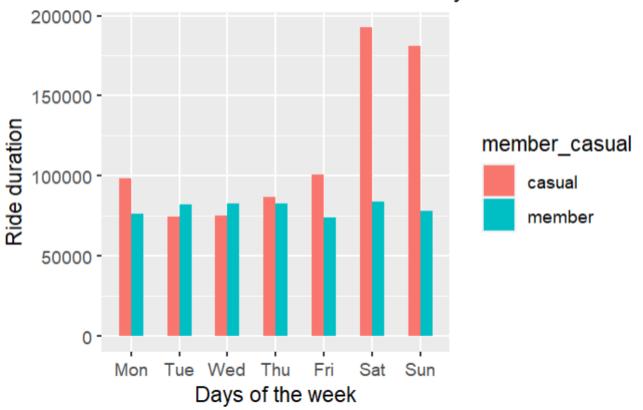
Number of rides in day of week for members & C



• Break up of ride duration based on the day of the week

```
Yearly_Cleaned %>%
   group_by(member_casual,weekday) %>%
   summarise(Total_ride_duration_hrs=sum(rideduration)/60,.groups="drop") %>%
   arrange(member_casual,weekday) %>%
   ggplot(aes(x=weekday,y=Total_ride_duration_hrs,fill=member_casual))+
   geom_col(width = 0.5,position = position_dodge(width = 0.5))+
   labs(title="Total ride duration in hrs for day of week : Members & Casuals") +
   xlab("Days of the week") +
   ylab("Ride duration") +
   scale_y_continuous(labels = function(x)format(x,scientific=FALSE))
```

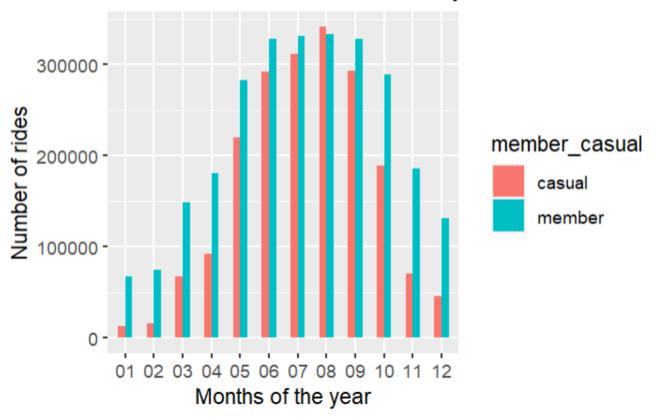
Total ride duration in hrs for day of week: Memb



• Break up of number of rides over the month of the year

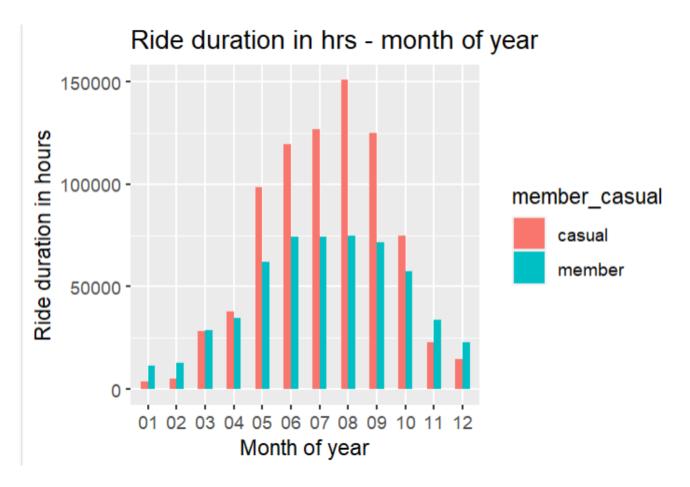
```
Yearly_Cleaned %>%
  group_by(member_casual,month) %>%
  summarise(Number_of_rides = length(ride_id)) %>%
  arrange(member_casual,month) %>%
  ggplot(aes(x=month,y=Number_of_rides,fill=member_casual))+
   geom_col(width = 0.5,position = position_dodge(width = 0.5))+
   labs(title="Number of rides in months of year for members & Casuals") +
   xlab("Months of the year")+ ylab("Number of rides")+
   scale_y_continuous(labels = function(x)format(x,scientific=FALSE))
```

Number of rides in months of year for members



• Break up of ride duration over the month of the year

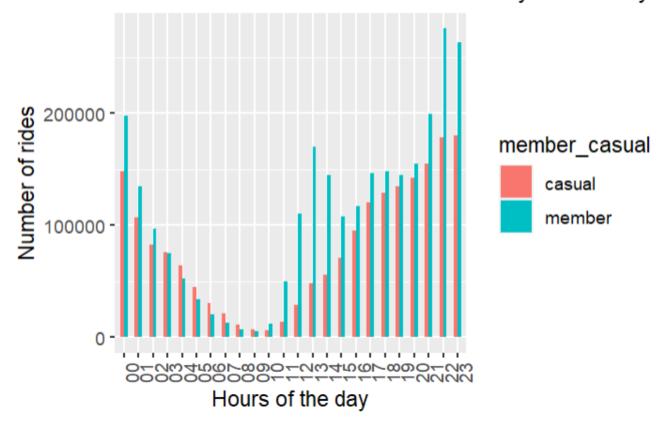
```
Yearly_Cleaned %>%
  group_by(member_casual,month) %>%
  arrange(member_casual,month) %>%
  summarize(Number_of_hours = sum(rideduration)/60) %>%
  ggplot(aes(x=month,y=Number_of_hours,fill=member_casual))+
  geom_col(width = 0.5,position = position_dodge(width = 0.5))+
  labs(title="Ride duration in hrs - month of year") +
  xlab("Month of year") +
  ylab("Ride duration in hours") +
  scale_y_continuous(labels = function(x)format(x,scientific=FALSE))
```



• Break up of number of rides over the hour of the day

```
Yearly_Cleaned %>%
  group_by(member_casual,starthour) %>%
  summarise(Number_of_rides = length(ride_id)) %>%
  arrange(member_casual,starthour) %>%
  ggplot(aes(x=starthour,y=Number_of_rides,fill=member_casual))+
  geom_col(width = 0.5,position = position_dodge(width = 0.5))+
  labs(title="Number of rides in each hour of day over the year") +
  xlab("Hours of the day")+ ylab("Number of rides")+
  theme(axis.text.x = element_text(angle = 90))+
  scale_y_continuous(labels = function(x)format(x,scientific=FALSE))
```

Number of rides in each hour of day over the year



• Break up of ride duration over the hour of the day

```
Yearly_Cleaned %>%
  group_by(member_casual,starthour) %>%
  arrange(member_casual,starthour) %>%
  summarize(Number_of_hours = sum(rideduration)/60) %>%
  ggplot(aes(x=starthour,y=Number_of_hours,fill=member_casual))+
  geom_col(width = 0.5,position = position_dodge(width = 0.5))+
  labs(title="Ride duration (hrs) in hour of the day") +
  xlab("Hour of the day")+ ylab("Ride duration in hours")+
  theme(axis.text.x = element_text(angle = 90)) +
  scale_y_continuous(labels = function(x)format(x,scientific=FALSE))
```

Ride duration (hrs) in hour of the day

