# Bertinelli\_Gabriele\_Rlab01

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## 1 RLab01 - Gabriele Bertinelli (2103359)

```
[]: library(tidyverse)
     # library(qqplot2)
     # library(tibble)
     # library(dplyr)
     library(geosphere)
     Sys.setlocale("LC_TIME", "English")
[]: # read_csv is from tidyverse, faster than read.csv
     d1 <- read_csv('./Data_CitiBike/JC-201902-citibike-tripdata.csv')</pre>
     d2 <- read_csv('./Data_CitiBike/JC-201903-citibike-tripdata.csv')</pre>
     d3 <- read_csv('./Data_CitiBike/JC-201904-citibike-tripdata.csv')</pre>
     d4 <- read_csv('./Data_CitiBike/JC-201905-citibike-tripdata.csv')</pre>
     d5 <- read_csv('./Data_CitiBike/JC-201906-citibike-tripdata.csv')</pre>
[3]: str(d1)
    spc_tbl_ [18,565 x 15] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
     $ tripduration
                               : num [1:18565] 142 223 106 370 315 ...
                               : POSIXct[1:18565], format: "2019-02-01 15:35:02"
     $ starttime
    "2019-02-01 17:00:46" ...
     $ stoptime
                               : POSIXct[1:18565], format: "2019-02-01 15:37:24"
    "2019-02-01 17:04:30" ...
     $ start station id
                               : num [1:18565] 3183 3183 3183 3183 ...
                               : chr [1:18565] "Exchange Place" "Exchange Place"
     $ start station name
    "Exchange Place" "Exchange Place" ...
     $ start station latitude : num [1:18565] 40.7 40.7 40.7 40.7 40.7 ...
     $ start station longitude: num [1:18565] -74 -74 -74 -74 -74 ...
     $ end station id
                               : num [1:18565] 3639 3681 3184 3211 3273 ...
     $ end station name
                             : chr [1:18565] "Harborside" "Grand St" "Paulus Hook"
    "Newark Ave" ...
     $ end station latitude : num [1:18565] 40.7 40.7 40.7 40.7 40.7 ...
     $ end station longitude : num [1:18565] -74 -74 -74 -74 -74 ...
```

```
: chr [1:18565] "Subscriber" "Subscriber" "Subscriber"
     $ usertype
    "Subscriber" ...
     $ birth year
                              : num [1:18565] 1963 1992 1960 1976 1980 ...
     $ gender
                               : num [1:18565] 1 2 1 1 1 1 1 1 1 1 ...
     - attr(*, "spec")=
      .. cols(
           tripduration = col_double(),
           starttime = col_datetime(format = ""),
           stoptime = col_datetime(format = ""),
           `start station id` = col_double(),
           `start station name` = col_character(),
           `start station latitude` = col_double(),
           `start station longitude` = col_double(),
           `end station id` = col_double(),
           `end station name` = col_character(),
           `end station latitude` = col_double(),
           `end station longitude` = col_double(),
           bikeid = col_double(),
           usertype = col character(),
           `birth year` = col_double(),
      . .
           gender = col double()
      . .
     - attr(*, "problems")=<externalptr>
[4]: # bind rows() is useful for combining data frames wherein the columns are all,
      \rightarrow identical
     df <- as_tibble(bind_rows(d1, d2, d3, d4, d5))</pre>
[5]: str(df) # check shape of df
    tibble [150,792 × 15] (S3: tbl_df/tbl/data.frame)
     $ tripduration
                               : num [1:150792] 142 223 106 370 315 ...
                               : POSIXct[1:150792], format: "2019-02-01 15:35:02"
     $ starttime
    "2019-02-01 17:00:46" ...
     $ stoptime
                              : POSIXct[1:150792], format: "2019-02-01 15:37:24"
    "2019-02-01 17:04:30" ...
                             : num [1:150792] 3183 3183 3183 3183 3183 ...
     $ start station id
     $ start station name
                              : chr [1:150792] "Exchange Place" "Exchange Place"
    "Exchange Place" "Exchange Place" ...
     $ start station latitude : num [1:150792] 40.7 40.7 40.7 40.7 40.7 ...
     \ start station longitude: num [1:150792] -74 -74 -74 -74 -74 ...
     $ end station id
                             : num [1:150792] 3639 3681 3184 3211 3273 ...
     $ end station name
                           : chr [1:150792] "Harborside" "Grand St" "Paulus Hook"
    "Newark Ave" ...
     $ end station latitude : num [1:150792] 40.7 40.7 40.7 40.7 40.7 ...
```

: num [1:18565] 29677 26234 29588 29250 29586 ...

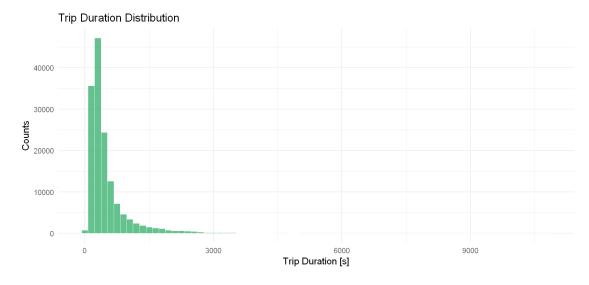
\$ bikeid

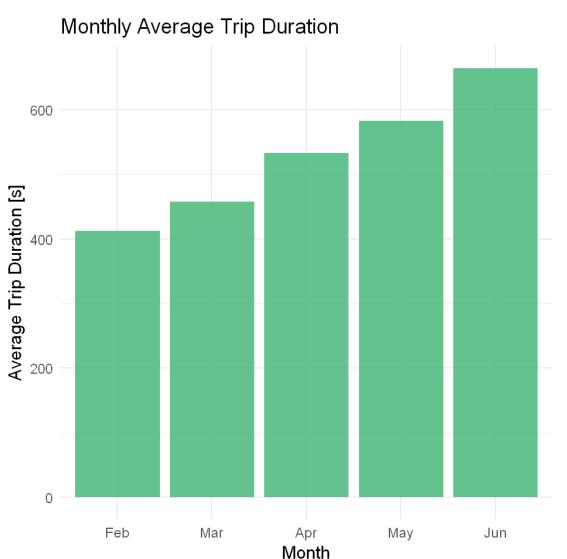
```
$ end station longitude : num [1:150792] -74 -74 -74 -74 -74 ...
                             : num [1:150792] 29677 26234 29588 29250 29586 ...
     $ bikeid
     $ usertype
                               : chr [1:150792] "Subscriber" "Subscriber"
    "Subscriber" "Subscriber" ...
     $ birth year
                             : num [1:150792] 1963 1992 1960 1976 1980 ...
     $ gender
                              : num [1:150792] 1 2 1 1 1 1 1 1 1 1 ...
    3)
[6]: df <- drop_na(df) # drop rows with NA values
[7]: str(df)
     # no rows removed due to NA values
    tibble [150,792 × 15] (S3: tbl_df/tbl/data.frame)
     $ tripduration
                              : num [1:150792] 142 223 106 370 315 ...
                               : POSIXct[1:150792], format: "2019-02-01 15:35:02"
     $ starttime
    "2019-02-01 17:00:46" ...
     $ stoptime
                              : POSIXct[1:150792], format: "2019-02-01 15:37:24"
    "2019-02-01 17:04:30" ...
     $ start station id
                             : num [1:150792] 3183 3183 3183 3183 3183 ...
     $ start station name
                             : chr [1:150792] "Exchange Place" "Exchange Place"
    "Exchange Place" "Exchange Place" ...
     $ start station latitude : num [1:150792] 40.7 40.7 40.7 40.7 40.7 ...
     $ start station longitude: num [1:150792] -74 -74 -74 -74 -74 ...
     $ end station id
                             : num [1:150792] 3639 3681 3184 3211 3273 ...
     $ end station name
                             : chr [1:150792] "Harborside" "Grand St" "Paulus Hook"
    "Newark Ave" ...
     $ end station latitude : num [1:150792] 40.7 40.7 40.7 40.7 40.7 ...
     $ end station longitude : num [1:150792] -74 -74 -74 -74 -74 ...
     $ bikeid
                               : num [1:150792] 29677 26234 29588 29250 29586 ...
     $ usertype
                               : chr [1:150792] "Subscriber" "Subscriber"
    "Subscriber" "Subscriber" ...
     $ birth year
                              : num [1:150792] 1963 1992 1960 1976 1980 ...
                              : num [1:150792] 1 2 1 1 1 1 1 1 1 1 ...
     $ gender
    4) 4.1)
[8]: avg_trip_duration <- mean(df$tripduration)
     median_trip_duration <- median(df$tripduration)</pre>
     sprintf("Average trip duration: %.2f sec", avg_trip_duration)
     sprintf("Median trip duration: %.2f sec", median_trip_duration)
    'Average trip duration: 768.64 sec'
```

'Median trip duration: 341.00 sec'

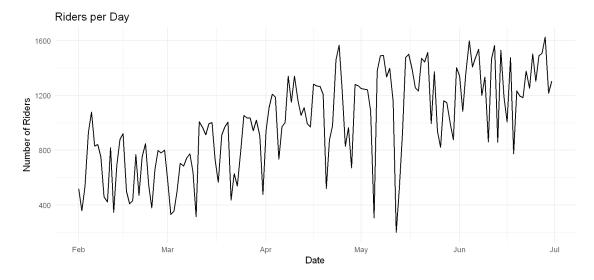
```
4.2)
```

```
[9]: min_trip_duration <- min(df$tripduration)
      max_trip_duration <- max(df$tripduration)</pre>
      sprintf("Min trip duration: %.2f sec", min_trip_duration)
      sprintf("Max trip duration: %.2f sec", max_trip_duration)
      sprintf('Max trip duration in hours: %.2f h', max trip_duration/3600)
     'Min trip duration: 61.00 sec'
     'Max trip duration: 1729020.00 sec'
     'Max trip duration in hours: 480.28 h'
     There's a trip that's really really long ahah. Almost 20 days. Maybe some long excursion... (or,
     more likely, wrong data)
     4.3)
[10]: df_clean <- df %>% filter(tripduration <= 3*3600) # filter out trips longer_
       ⇔than 3 hours
[11]: avg_trip_duration_clean <- mean(df_clean$tripduration)
      median_trip_duration_clean <- median(df_clean$tripduration)</pre>
      sprintf("Average trip duration: %.2f sec", avg_trip_duration_clean)
      sprintf("Median trip duration: %.2f sec", median_trip_duration_clean)
      min_trip_duration_clean <- min(df_clean$tripduration)</pre>
      max_trip_duration_clean <- max(df_clean$tripduration)</pre>
      sprintf("Min trip duration: %.2f sec", min_trip_duration_clean)
      sprintf("Max trip duration: %.2f sec", max_trip_duration_clean)
      sprintf('Max trip duration in hours: %.2f h', max_trip_duration_clean/3600)
     'Average trip duration: 553.38 sec'
     'Median trip duration: 340.00 sec'
     'Min trip duration: 61.00 sec'
     'Max trip duration: 10800.00 sec'
     'Max trip duration in hours: 3.00 h'
     4.4)
```





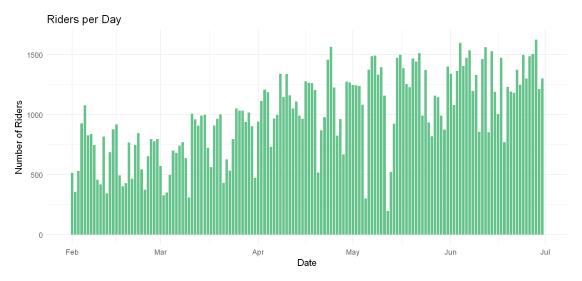
#### 6) 6.1)



```
theme_minimal(base_size⊔

c= 18)

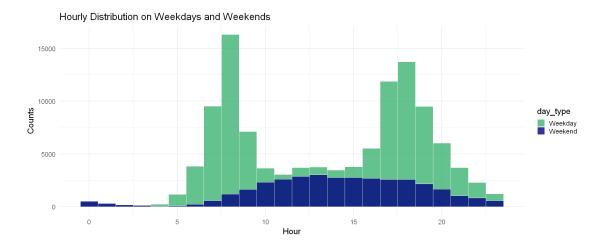
riders_per_day_barplot # barplot of ridership per day
```



#### 6.2)

```
[16]: # Extract the day of the week from the starttime col
      df_clean$weekday <- lubridate::wday(df_clean$starttime, label = TRUE)</pre>
      # Create a new col to separate between weekdays and weekends
      df_clean$day_type <- ifelse(df_clean$weekday %in% c("Sat", "Sun"), "Weekend", __
       →"Weekday")
      # Extract the hour from the starttime column
      df_clean$hour <- lubridate::hour(df_clean$starttime)</pre>
      options(repr.plot.width = 17, repr.plot.height = 7)
      hourly_distribution <- ggplot(data = df_clean, aes(x = hour, fill = day_type)) +
                                                          geom_histogram(binwidth = 1,__
       →position = "identity", alpha = 0.8, color='ivory') +
                                                          labs(title = "Hourly_
       \hookrightarrowDistribution on Weekdays and Weekends", x = "Hour", y = "Counts") +
                                                          scale_fill_manual(values =_
       →c("Weekday" = "mediumseagreen", "Weekend" = "navy")) +
                                                          theme_minimal(base_size=18)
```

#### hourly\_distribution # hourly distribution on weekdays and weekends



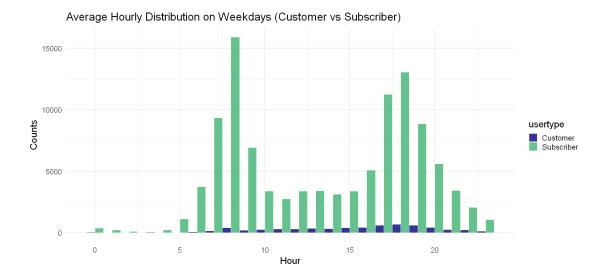
#### 6.3)

```
[17]: # Filter the data for weekdays
      df_weekdays <- df_clean %>% filter(day_type == "Weekday")
      # Group the data by hour and user type
      hourly_distribution <- df_weekdays %>%
                                                      group_by(hour, usertype) %>%
                                                       # summarize(avg_count =_
       →mean(tripduration)) %>%
                                                      ungroup()
      options(repr.plot.width = 15, repr.plot.height = 7)
      hourly_plot <- ggplot(data = hourly_distribution, aes(x = hour, fill =_u

usertype)) +
                                      geom_histogram(binwidth = 1, position =__

¬"dodge", alpha = 0.8, color='ivory') +
                                      labs(title = "Average Hourly Distribution on_
       ⇔Weekdays (Customer vs Subscriber)",
                                                      x = "Hour", y = "Counts") +
                                      scale_fill_manual(values = c("Customer" =_

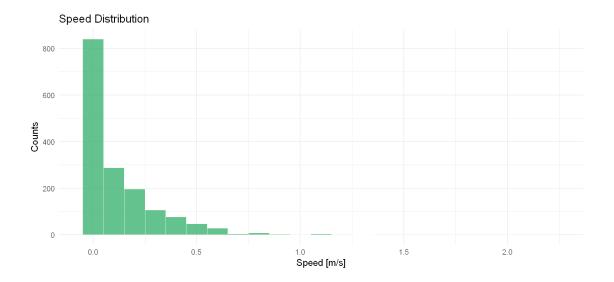
¬"navy", "Subscriber" = "mediumseagreen")) +
                                      theme_minimal(base_size = 18)
      hourly_plot # hourly distribution on weekdays, separating customer and_
       ⇒subscriber users
```



### 7) 7.1)

```
[18]: df_vel <- df_clean %>% filter(tripduration > 3600) # filter out trips longer_
       ⇔than 1 hour
      # calculate distance between start and end stations
      df_vel$dist <- distHaversine(cbind(df_vel$`start station longitude`,_</pre>
       ⇒df_vel$`start station latitude`),
                                                                  cbind(df_vel$`end_
       station longitude`, df_vel$`end station latitude`))
      df_vel$speed <- (df_vel$dist / df_vel$tripduration)#*3.6</pre>
      options(repr.plot.width = 15, repr.plot.height = 7)
      speed_hist <- ggplot(data=df_vel, aes(x=speed)) +</pre>
                                       geom_histogram(binwidth=0.1,_

→fill='mediumseagreen', color='ivory', alpha=0.8) +
                                       labs(title='Speed Distribution', x='Speed [m/
       ⇔s]', y='Counts') +
                                       theme_minimal(base_size = 18)
      speed_hist
```



### 7.2)

```
[19]: # Filter the data for distance groups
      df_group1 <- df_vel %>% filter(dist < 500)</pre>
      df_group2 <- df_vel %>% filter(dist >= 500 & dist < 1000)</pre>
      df group3 <- df vel %>% filter(dist >= 1000 & dist < 2000)</pre>
      df_group4 <- df_vel %>% filter(dist >= 2000 & dist < 3000)</pre>
      df_group5 <- df_vel %>% filter(dist >= 3000)
      avg_speed_group1 <- mean(df_group1$speed)</pre>
      avg_speed_group2 <- mean(df_group2$speed)</pre>
      avg_speed_group3 <- mean(df_group3$speed)</pre>
      avg_speed_group4 <- mean(df_group4$speed)</pre>
      avg_speed_group5 <- mean(df_group5$speed)</pre>
      avg_speed_dist <- data.frame(group_dist = c("<500m", "500m-1000m", u
       \circ"1000m-2000m", "2000m-3000m", ">3000m"),
                                                                            avg_speed =_

¬c(avg_speed_group1, avg_speed_group2, avg_speed_group3,
       →avg_speed_group5))
      options(repr.plot.width = 8, repr.plot.height = 8)
      speed_distance_plot <- ggplot(data = avg_speed_dist, aes(x = group_dist, y =__</pre>
       →avg_speed)) +
```

```
geom_bar(stat = "identity", u

fill = "mediumseagreen", color = "ivory", alpha = 0.8) +

labs(title = "Average Speed as u

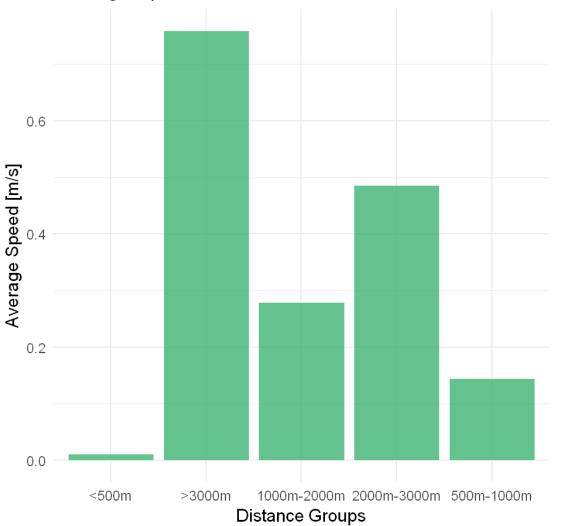
a Function of Distance",

x = "Distance Groups", y = u

theme_minimal(base_size=18)

speed_distance_plot # Plot the average speed as a function of distance
```

# Average Speed as a Function of Distance



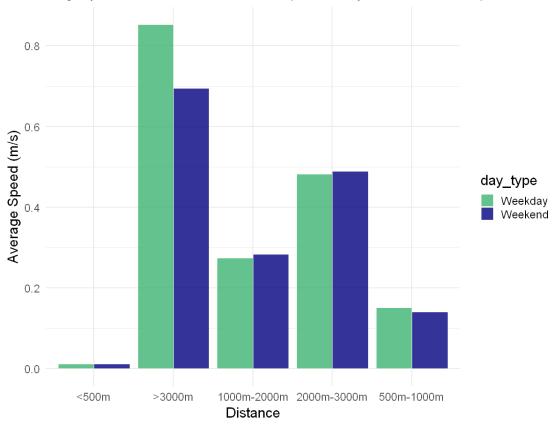
7.3)

```
[20]: # Filter the data for weekdays
      df_weekday <- df_vel %>% filter(day_type == "Weekday")
      # Filter the data for weekends
      df_weekend <- df_vel %>% filter(day_type == "Weekend")
      # same operation as before, but now for weekdays and weekends
      df group1d <- df weekday %>% filter(dist < 500)</pre>
      df_group2d <- df_weekday %>% filter(dist >= 500 & dist < 1000)</pre>
      df group3d <- df weekday \%>% filter(dist >= 1000 & dist < 2000)
      df group4d <- df weekday \%>% filter(dist >= 2000 & dist < 3000)
      df_group5d <- df_weekday %>% filter(dist >= 3000)
      df_group1w <- df_weekend %>% filter(dist < 500)</pre>
      df_group2w <- df_weekend %>% filter(dist >= 500 & dist < 1000)</pre>
      df_group3w <- df_weekend %>% filter(dist >= 1000 & dist < 2000)</pre>
      df_group4w <- df_weekend %>% filter(dist >= 2000 & dist < 3000)</pre>
      df_group5w <- df_weekend %>% filter(dist >= 3000)
      avg_speed_group1d <- mean(df_group1d$speed)</pre>
      avg_speed_group2d <- mean(df_group2d$speed)</pre>
      avg speed group3d <- mean(df group3d$speed)</pre>
      avg_speed_group4d <- mean(df_group4d$speed)</pre>
      avg speed group5d <- mean(df group5d$speed)</pre>
      avg_speed_group1w <- mean(df_group1w$speed)</pre>
      avg_speed_group2w <- mean(df_group2w$speed)</pre>
      avg_speed_group3w <- mean(df_group3w$speed)</pre>
      avg_speed_group4w <- mean(df_group4w$speed)</pre>
      avg_speed_group5w <- mean(df_group5w$speed)</pre>
      avg_speed_distd <- data.frame(group_dist = c("<500m", "500m-1000m", __
       avg_speed =
       →c(avg_speed_group1d, avg_speed_group2d, avg_speed_group3d,
       →avg_speed_group5d),
                                                                      day_type =
       avg_speed_distw <- data.frame(group_dist = c("<500m", "500m-1000m", "
       avg_speed =
       →c(avg_speed_group1w, avg_speed_group2w, avg_speed_group3w,
```

```
→avg_speed_group5w),
                                                           day_type =
df_avg_speed_dist <- rbind(avg_speed_distd, avg_speed_distw) # combine the twou
\hookrightarrow dataframes
options(repr.plot.width = 10, repr.plot.height = 8)
# Plot the average speed as a function of distance
speed_distance_plotwd <- ggplot(data = df_avg_speed_dist, aes(x = group_dist, yu
→= avg_speed, fill=day_type)) +
                                                    geom_bar(position='dodge',__
stat = "identity", alpha = 0.8, color='ivory') +
                                                    labs(title = "Avg Speed□
⇔as a Function of Dist (Weekdays vs Weekends) ",
                                                    x = "Distance", y = \Box

¬"Average Speed (m/s)") +
                                                    scale_fill_manual(values_
 theme_minimal(base_size_
 ⇒= 18)
speed_distance_plotwd
```





## 8) 8.1)

'Most common start station -> Grove St PATH'

'Least popular end station -> 1 Ave & E 16 St'

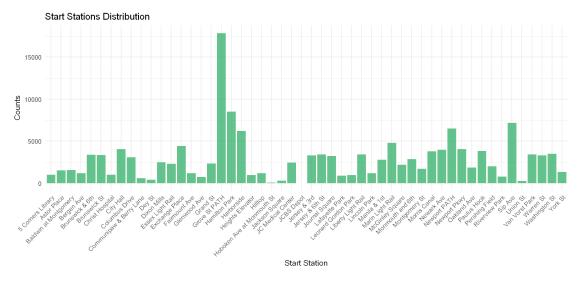
8.2)

```
[22]: start_station <- df_clean %>% group_by(`start station name`) #%>%_

$\times summarize(count = n())$
```

```
# start station
options(repr.plot.width = 15, repr.plot.height = 7)
# start_distr <- ggplot(data=start_station, aes(x=`start station name`)) +</pre>
                   geom_histogram(stat='count', binwidth = 1,__
 →fill='mediumseagreen', color='ivory', alpha=0.8) +
                   labs(title='Start Station Distribution', x='Start Station', L
 \rightarrow y = 'Frequency')
                   theme_minimal()+
                   theme(axis.text.x = element\_text(angle = 45, hjust = 1))
start_distr <- ggplot(data=start_station, aes(x=`start station name`)) +</pre>
                                   geom_bar(fill='mediumseagreen', color='ivory', __
 \rightarrowalpha=0.8) +
                                   labs(title='Start Stations Distribution', __

¬x='Start Station', y='Counts') +
                                   theme_minimal(base_size = 15) +
                                   theme(axis.text.x = element_text(angle = 45,__
 \rightarrowhjust = 1))
start_distr
```



8.3)

```
least_route <- df_clean %>% group_by(`start station name`, `end station name`)_u
      →%>% summarize(count = n()) %>% arrange(count)
     cat('\n3 most common routes:\n')
     for (i in 1:3) {
             print(sprintf('%s --> %s, counts: %d', common route$`start station,
     aname`[i], common_route$`end station name`[i], common_route$count[i]))
     }
     cat('\n3 least popular routes:\n')
     for (i in 1:3) {
             print(sprintf('%s --> %s, counts: %d', least_route$`start station_
      aname`[i], least_route$`end station name`[i], least_route$count[i]))
     }
    `summarise()` has grouped output by 'start station name'. You can
    override
    using the `.groups` argument.
    `summarise()` has grouped output by 'start station name'. You can
    override
    using the `.groups` argument.
    3 most common routes:
    [1] "Hamilton Park --> Grove St PATH, counts: 3037"
    [1] "Grove St PATH --> Hamilton Park, counts: 2318"
    [1] "Brunswick & 6th --> Grove St PATH, counts: 1915"
    3 least popular routes:
    [1] "5 Corners Library --> Dixon Mills, counts: 1"
    [1] "5 Corners Library --> Grand St, counts: 1"
    [1] "Astor Place --> Brunswick & 6th, counts: 1"
[]:
```