

analysis of smart device usage

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Ask

1. What are some trends in smart device usage?
2. How could these trends apply to Bellabeat customers?
3. How could these trends help influence Bellabeat marketing strategy?

Prepare

For this analysis we use an open source dataset available on [kaggle.link](https://www.kaggle.com/arashnic/fitbit) (<https://www.kaggle.com/arashnic/fitbit>). There are 18 csv files in the datasets. After carefully check the dataset we decide to use the `dailyActivity_merged.csv`, `sleepDay_merged.csv` and `dailyCalories_merged.csv`.

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.1 --
```

```
## v ggplot2 3.3.5      v purrr   0.3.4
## v tibble  3.1.1      v dplyr   1.0.6
## v tidyr   1.1.3      v stringr 1.4.0
## v readr   1.4.0      v forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
daily_activity <- read.csv("D:/FitabaseData/dailyActivity_merged.csv")
head(daily_activity)
```

```
##           Id ActivityDate TotalSteps TotalDistance TrackerDistance
## 1 1503960366    4/12/2016     13162           8.50           8.50
## 2 1503960366    4/13/2016     10735           6.97           6.97
## 3 1503960366    4/14/2016     10460           6.74           6.74
## 4 1503960366    4/15/2016      9762           6.28           6.28
## 5 1503960366    4/16/2016     12669           8.16           8.16
## 6 1503960366    4/17/2016      9705           6.48           6.48
##   LoggedActivitiesDistance VeryActiveDistance ModeratelyActiveDistance
## 1                        0                1.88                   0.55
## 2                        0                1.57                   0.69
## 3                        0                2.44                   0.40
## 4                        0                2.14                   1.26
## 5                        0                2.71                   0.41
## 6                        0                3.19                   0.78
##   LightActiveDistance SedentaryActiveDistance VeryActiveMinutes
## 1                6.06                    0                25
## 2                4.71                    0                21
## 3                3.91                    0                30
## 4                2.83                    0                29
## 5                5.04                    0                36
## 6                2.51                    0                38
##   FairlyActiveMinutes LightlyActiveMinutes SedentaryMinutes Calories
## 1                13                328                728       1985
## 2                19                217                776       1797
## 3                11                181               1218       1776
## 4                34                209                726       1745
## 5                10                221                773       1863
## 6                20                164                539       1728
```

```
sleep_day <- read.csv("D:/FitabaseData/sleepDay_merged.csv")
head(sleep_day)
```

```
##           Id           SleepDay TotalSleepRecords TotalMinutesAsleep
## 1 1503960366 4/12/2016 12:00:00 AM                1                327
## 2 1503960366 4/13/2016 12:00:00 AM                2                384
## 3 1503960366 4/15/2016 12:00:00 AM                1                412
## 4 1503960366 4/16/2016 12:00:00 AM                2                340
## 5 1503960366 4/17/2016 12:00:00 AM                1                700
## 6 1503960366 4/19/2016 12:00:00 AM                1                304
## TotalTimeInBed
## 1                346
## 2                407
## 3                442
## 4                367
## 5                712
## 6                320
```

```
daily_calories <- read.csv("D:/FitabaseData/dailyCalories_merged.csv")
head(daily_calories)
```

```
##           Id ActivityDay Calories
## 1 1503960366 4/12/2016    1985
## 2 1503960366 4/13/2016    1797
## 3 1503960366 4/14/2016    1776
## 4 1503960366 4/15/2016    1745
## 5 1503960366 4/16/2016    1863
## 6 1503960366 4/17/2016    1728
```

Process

in dataframe `sleep_day` the column `sleepDay` is in datetime format. we convert the datetime format to data format to be consistent with the other two dataframes.

```
sleep_day$SleepDay = as.Date(sleep_day$SleepDay, format = "%m/%d/%Y %I:%M:%S %p")
head(sleep_day)
```

```
##           Id SleepDay TotalSleepRecords TotalMinutesAsleep TotalTimeInBed
## 1 1503960366 2016-04-12                1                327                346
## 2 1503960366 2016-04-13                2                384                407
## 3 1503960366 2016-04-15                1                412                442
## 4 1503960366 2016-04-16                2                340                367
## 5 1503960366 2016-04-17                1                700                712
## 6 1503960366 2016-04-19                1                304                320
```

We rename the column sleepDay to ActivityDate in dataframe sleep_day. rename the column ActivityDay to ActivityDate. We then convert the column ActivityDate to %Y-%m-%d format. After that we convert the column ActivityDay in dataframe daily_calories to %Y-%m-%d format and rename the column ActivityDay to ActivityDate. The three dataframe has consistent data format after these operation.

```
sleep_day<-sleep_day%>% rename(ActivityDate=SleepDay)
head(sleep_day)
```

##		Id	ActivityDate	TotalSleepRecords	TotalMinutesAsleep	TotalTimeInBed
## 1	1503960366	2016-04-12	1	327	34	
6						
## 2	1503960366	2016-04-13	2	384	40	
7						
## 3	1503960366	2016-04-15	1	412	44	
2						
## 4	1503960366	2016-04-16	2	340	36	
7						
## 5	1503960366	2016-04-17	1	700	71	
2						
## 6	1503960366	2016-04-19	1	304	32	
0						

```
daily_activity$ActivityDate = as.Date(format(daily_activity$ActivityDate), "%m/%d/%Y")
head(daily_activity)
```

```
##           Id ActivityDate TotalSteps TotalDistance TrackerDistance
## 1 1503960366 2016-04-12      13162          8.50          8.50
## 2 1503960366 2016-04-13      10735          6.97          6.97
## 3 1503960366 2016-04-14      10460          6.74          6.74
## 4 1503960366 2016-04-15       9762          6.28          6.28
## 5 1503960366 2016-04-16      12669          8.16          8.16
## 6 1503960366 2016-04-17       9705          6.48          6.48
##   LoggedActivitiesDistance VeryActiveDistance ModeratelyActiveDistance
## 1                        0                1.88                   0.55
## 2                        0                1.57                   0.69
## 3                        0                2.44                   0.40
## 4                        0                2.14                   1.26
## 5                        0                2.71                   0.41
## 6                        0                3.19                   0.78
##   LightActiveDistance SedentaryActiveDistance VeryActiveMinutes
## 1                  6.06                      0                 25
## 2                  4.71                      0                 21
## 3                  3.91                      0                 30
## 4                  2.83                      0                 29
## 5                  5.04                      0                 36
## 6                  2.51                      0                 38
##   FairlyActiveMinutes LightlyActiveMinutes SedentaryMinutes Calories
## 1                   13                   328                728    1985
## 2                   19                   217                776    1797
## 3                   11                   181               1218    1776
## 4                   34                   209                726    1745
## 5                   10                   221                773    1863
## 6                   20                   164                539    1728
```

```
daily_calories$ActivityDay = as.Date(daily_calories$ActivityDay, format = "%m/%
d/%Y")
daily_calories<-daily_calories%>% rename(ActivityDate=ActivityDay)
head(daily_calories)
```

```
##           Id ActivityDate Calories
## 1 1503960366 2016-04-12    1985
## 2 1503960366 2016-04-13    1797
## 3 1503960366 2016-04-14    1776
## 4 1503960366 2016-04-15    1745
## 5 1503960366 2016-04-16    1863
## 6 1503960366 2016-04-17    1728
```

merge the three dataframe into a final dataframe and select only the columns we are interested.

```
final_data <- merge(daily_activity, sleep_day, daily_calories, by.x=c("Id", "ActivityDate"), by.y=c("Id", "ActivityDate"), by.z=c("Id", "ActivityDate"))
head(final_data)
```

```
##           Id ActivityDate TotalSteps TotalDistance TrackerDistance
## 1 1503960366 2016-04-12      13162           8.50           8.50
## 2 1503960366 2016-04-13      10735           6.97           6.97
## 3 1503960366 2016-04-15       9762           6.28           6.28
## 4 1503960366 2016-04-16      12669           8.16           8.16
## 5 1503960366 2016-04-17       9705           6.48           6.48
## 6 1503960366 2016-04-19      15506           9.88           9.88
##   LoggedActivitiesDistance VeryActiveDistance ModeratelyActiveDistance
## 1                        0                1.88                   0.55
## 2                        0                1.57                   0.69
## 3                        0                2.14                   1.26
## 4                        0                2.71                   0.41
## 5                        0                3.19                   0.78
## 6                        0                3.53                   1.32
##   LightActiveDistance SedentaryActiveDistance VeryActiveMinutes
## 1                  6.06                      0                25
## 2                  4.71                      0                21
## 3                  2.83                      0                29
## 4                  5.04                      0                36
## 5                  2.51                      0                38
## 6                  5.03                      0                50
##   FairlyActiveMinutes LightlyActiveMinutes SedentaryMinutes Calories
## 1                   13                   328                728    1985
## 2                   19                   217                776    1797
## 3                   34                   209                726    1745
## 4                   10                   221                773    1863
## 5                   20                   164                539    1728
## 6                   31                   264                775    2035
##   TotalSleepRecords TotalMinutesAsleep TotalTimeInBed
## 1                   1                327            346
## 2                   2                384            407
## 3                   1                412            442
## 4                   2                340            367
## 5                   1                700            712
## 6                   1                304            320
```

```
data<-final_data%>%select(Id,ActivityDate,TotalSteps,TotalDistance,VeryActiveMinutes,Calories,TotalMinutesAsleep)
head(data)
```

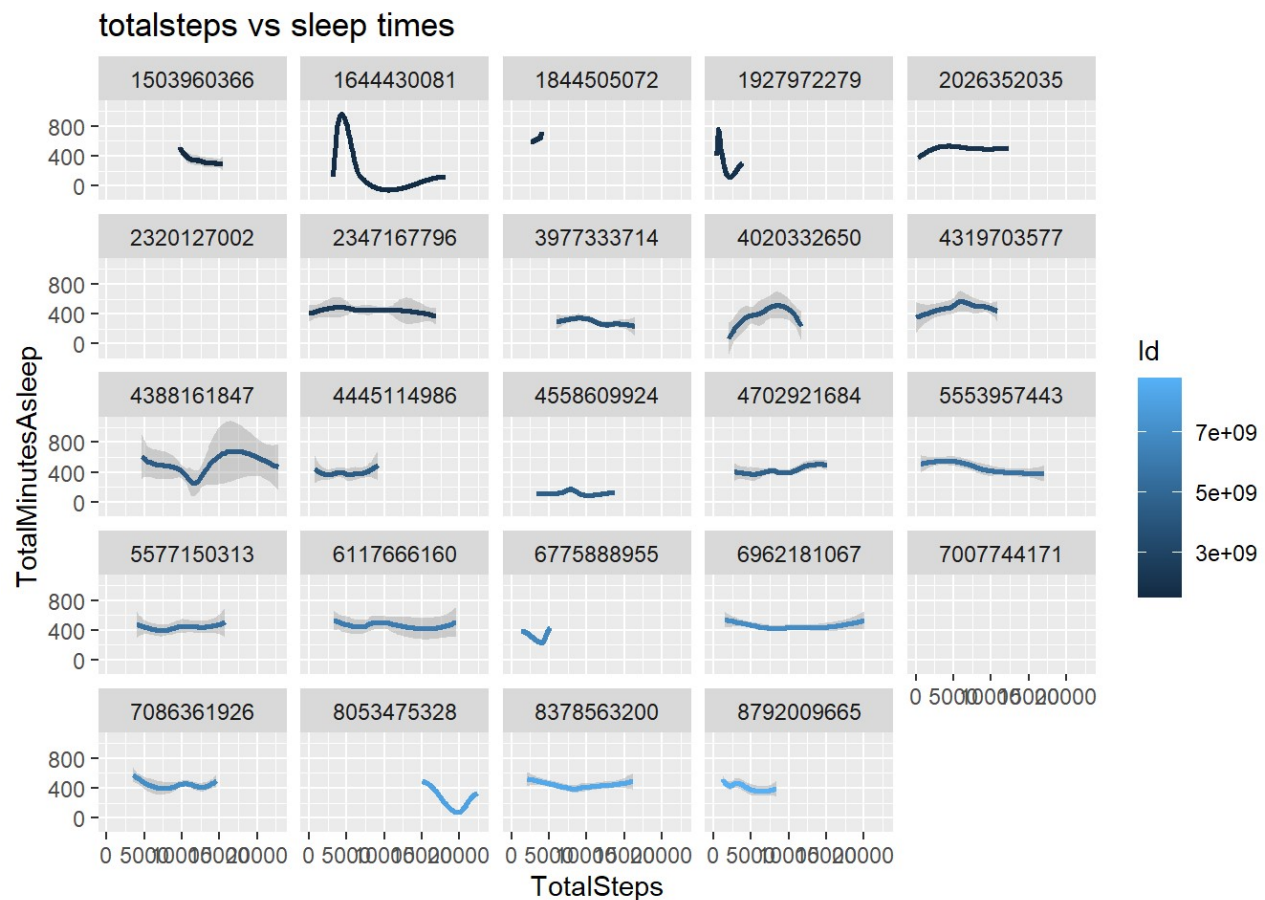
```
##           Id ActivityDate TotalSteps TotalDistance VeryActiveMinutes Calorie
s
## 1 1503960366 2016-04-12      13162          8.50             25        198
5
## 2 1503960366 2016-04-13      10735          6.97             21        179
7
## 3 1503960366 2016-04-15       9762          6.28             29        174
5
## 4 1503960366 2016-04-16      12669          8.16             36        186
3
## 5 1503960366 2016-04-17       9705          6.48             38        172
8
## 6 1503960366 2016-04-19      15506          9.88             50        203
5
## TotalMinutesAsleep
## 1              327
## 2              384
## 3              412
## 4              340
## 5              700
## 6              304
```

Analyze

first we check the relationships between TotalSteps and TotalMinutesAsleep

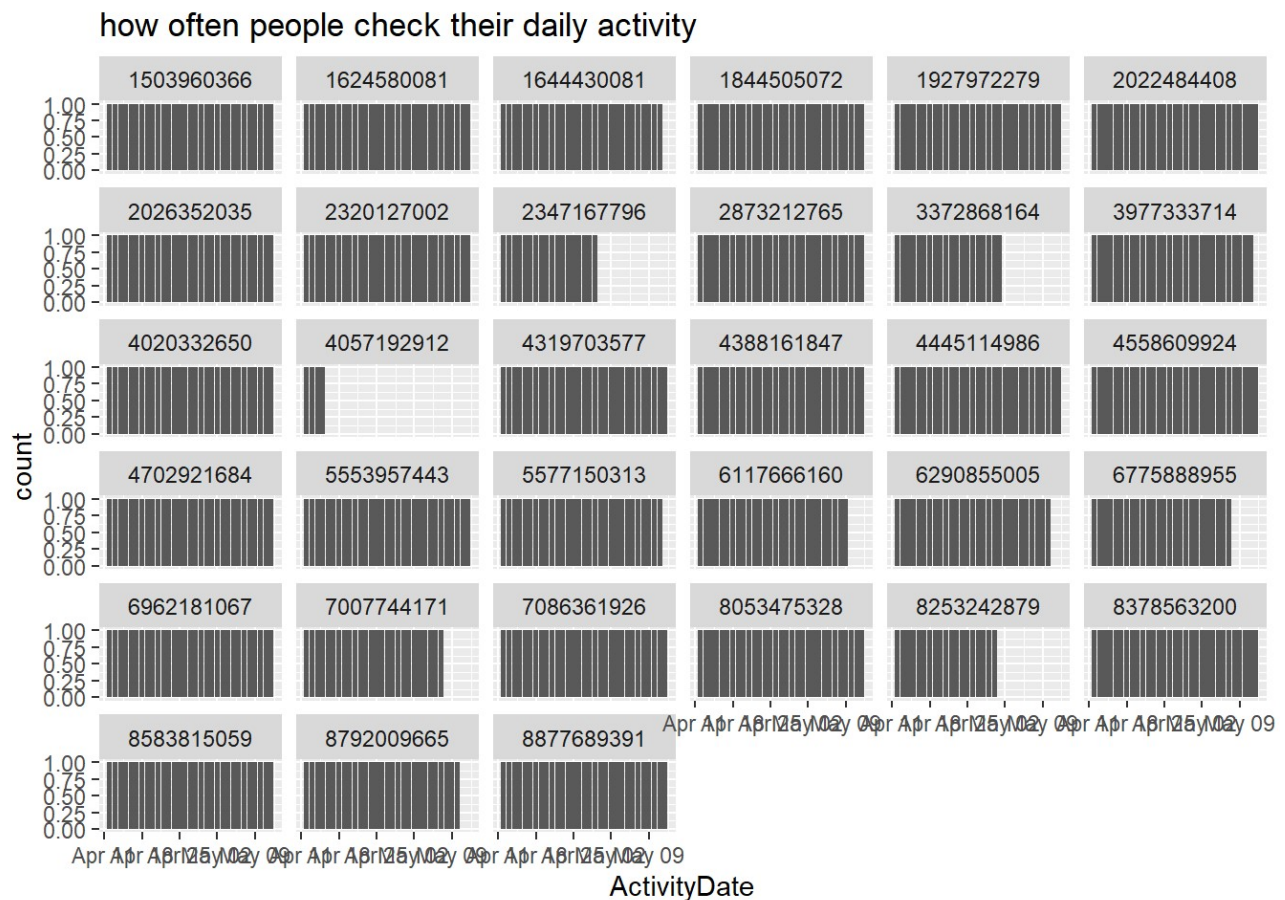
```
ggplot(data=data)+geom_smooth(mapping=aes(x=TotalSteps,y=TotalMinutesAsleep,col
or=Id))+facet_wrap(~Id)+labs(title="totalsteps vs sleep times")
```

```
## `geom_smooth()` using method = 'loess' and formula 'y ~ x'
```



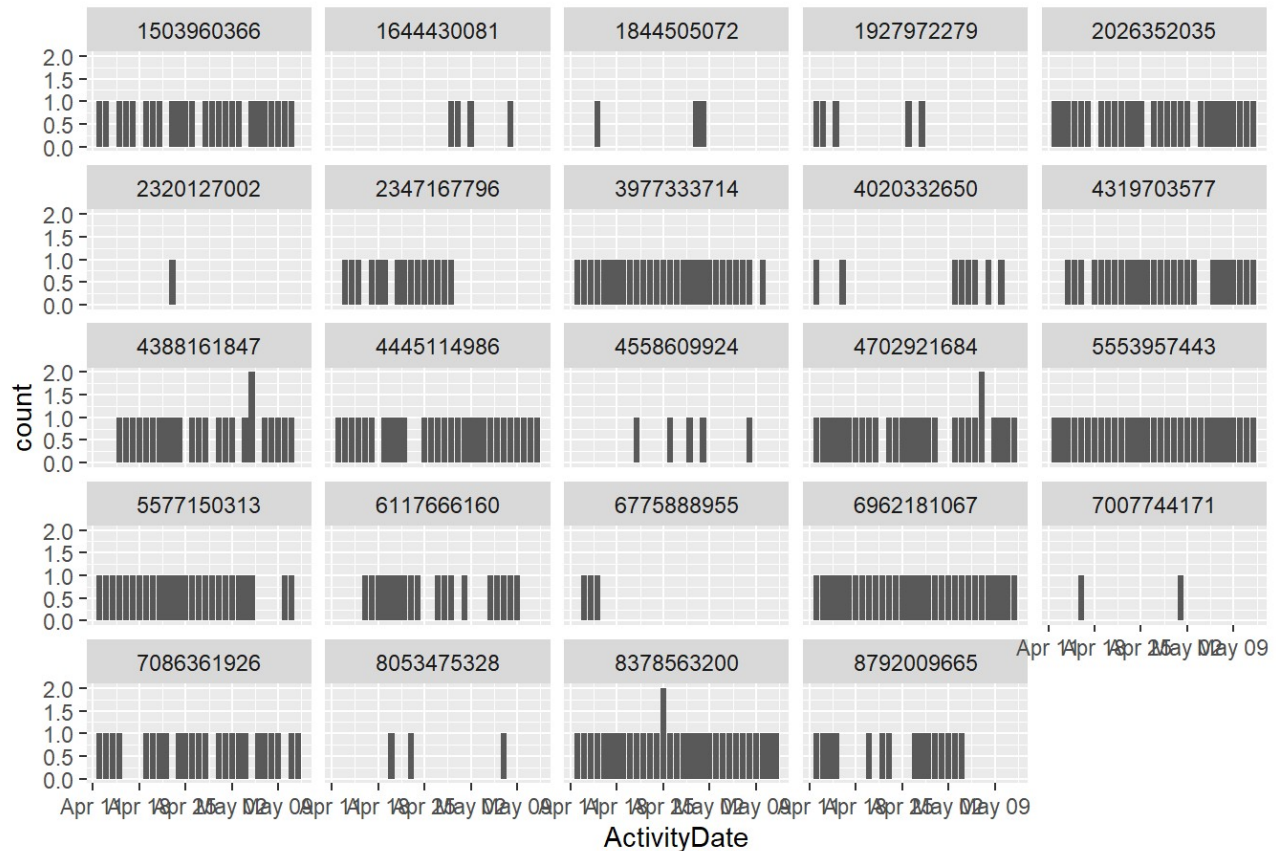
these graphs show that there are no clear relationship between totalsteps and sleeptimes. There are some peaks and valleys in the plots. Indicate appropriate excecies is helpful for some people.

```
ggplot(data=daily_activity)+geom_bar(mapping=aes(x=ActivityDate))+facet_wrap(~Id)+labs(title="how often people check their daily activity")
```

```
ggplot(data=sleep_day)+geom_bar(mapping=aes(x=ActivityDate))+facet_wrap(~Id)+labs(title="how often people check their sleeping tims")
```

how often people check their sleeping times

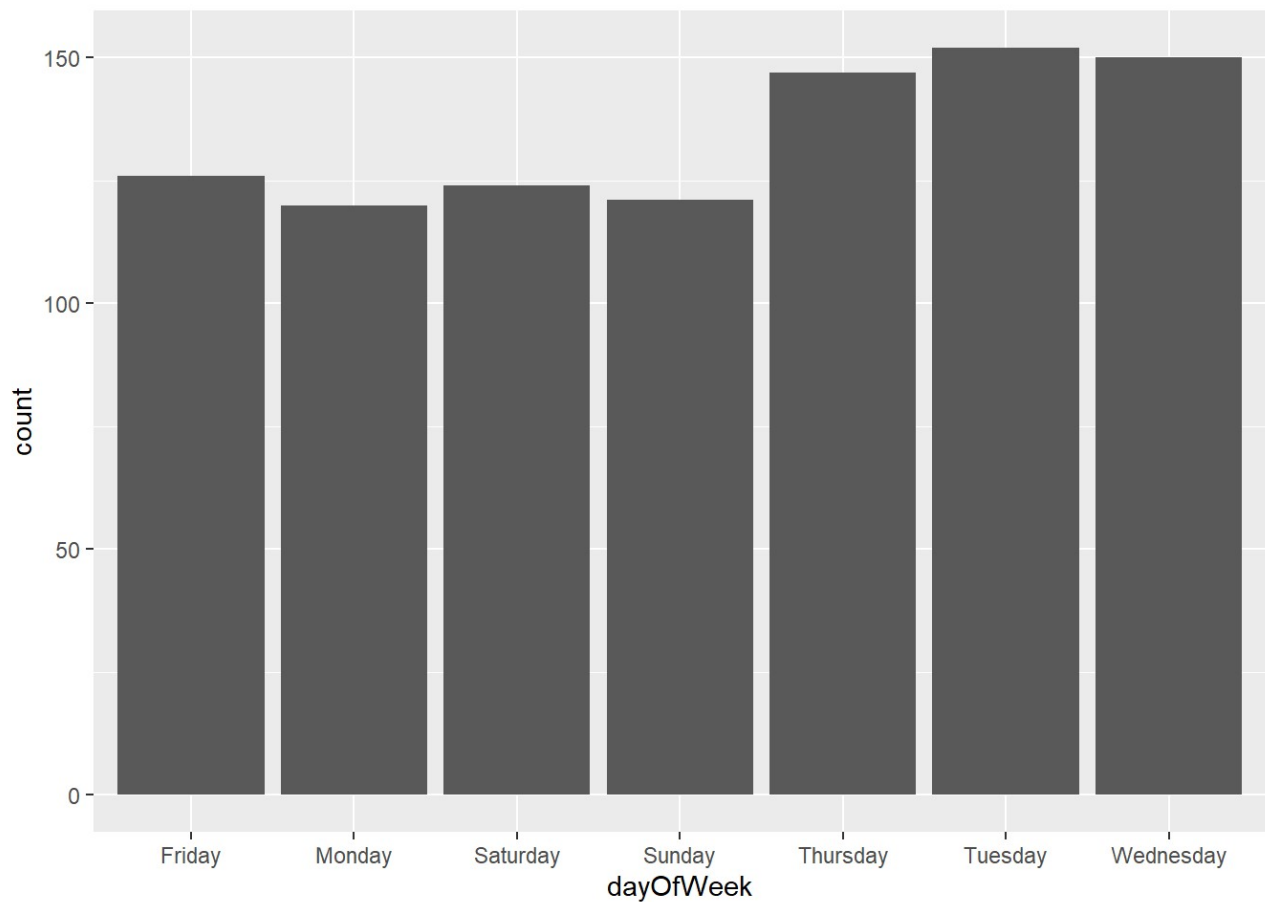


From the results we can clearly see that people prefer the use of devices to check their daily activities than to check their sleeping times as more users report their sleeping times than users report daily activities.

```
daily_activity$dayOfWeek<-weekdays(as.Date(daily_activity$ActivityDate))
head(daily_activity)
```

```
##           Id ActivityDate TotalSteps TotalDistance TrackerDistance
## 1 1503960366 2016-04-12      13162          8.50          8.50
## 2 1503960366 2016-04-13      10735          6.97          6.97
## 3 1503960366 2016-04-14      10460          6.74          6.74
## 4 1503960366 2016-04-15       9762          6.28          6.28
## 5 1503960366 2016-04-16      12669          8.16          8.16
## 6 1503960366 2016-04-17       9705          6.48          6.48
##   LoggedActivitiesDistance VeryActiveDistance ModeratelyActiveDistance
## 1                        0                1.88                0.55
## 2                        0                1.57                0.69
## 3                        0                2.44                0.40
## 4                        0                2.14                1.26
## 5                        0                2.71                0.41
## 6                        0                3.19                0.78
##   LightActiveDistance SedentaryActiveDistance VeryActiveMinutes
## 1                6.06                0                25
## 2                4.71                0                21
## 3                3.91                0                30
## 4                2.83                0                29
## 5                5.04                0                36
## 6                2.51                0                38
##   FairlyActiveMinutes LightlyActiveMinutes SedentaryMinutes Calories dayOfWeek
## 1                13                328                728        1985    Tuesd
ay
## 2                19                217                776        1797    Wednesd
ay
## 3                11                181                1218        1776    Thursd
ay
## 4                34                209                726        1745     Frid
ay
## 5                10                221                773        1863    Saturd
ay
## 6                20                164                539        1728     Sund
ay
```

```
ggplot(data=daily_activity)+geom_bar(mapping=aes(x=dayOfWeek))
```

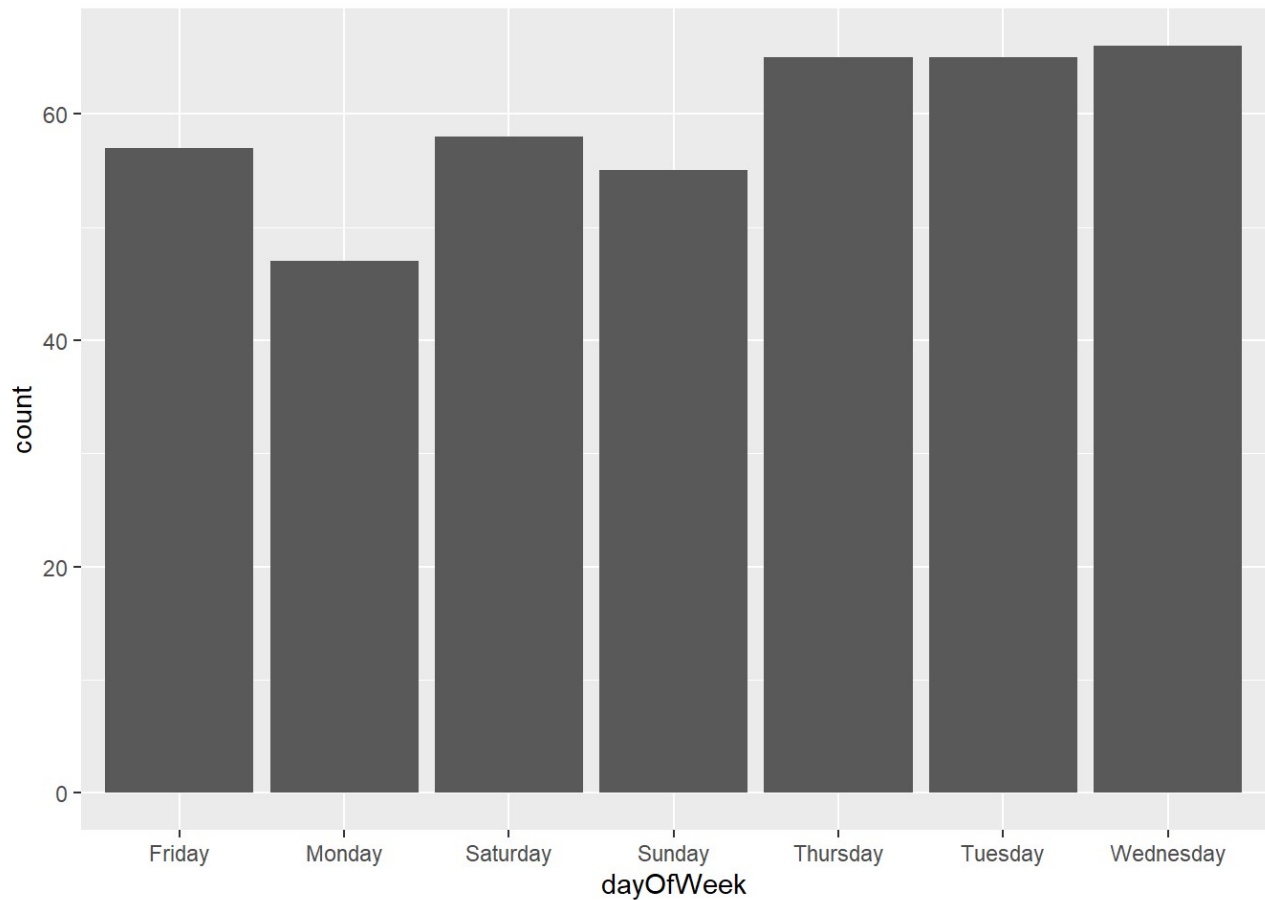


the graph shows that people are more likely to record exercise on Tuesday, Wednesday and Thursday.

```
sleep_day$dayOfWeek<-weekdays(as.Date(sleep_day$ActivityDate))  
head(sleep_day)
```

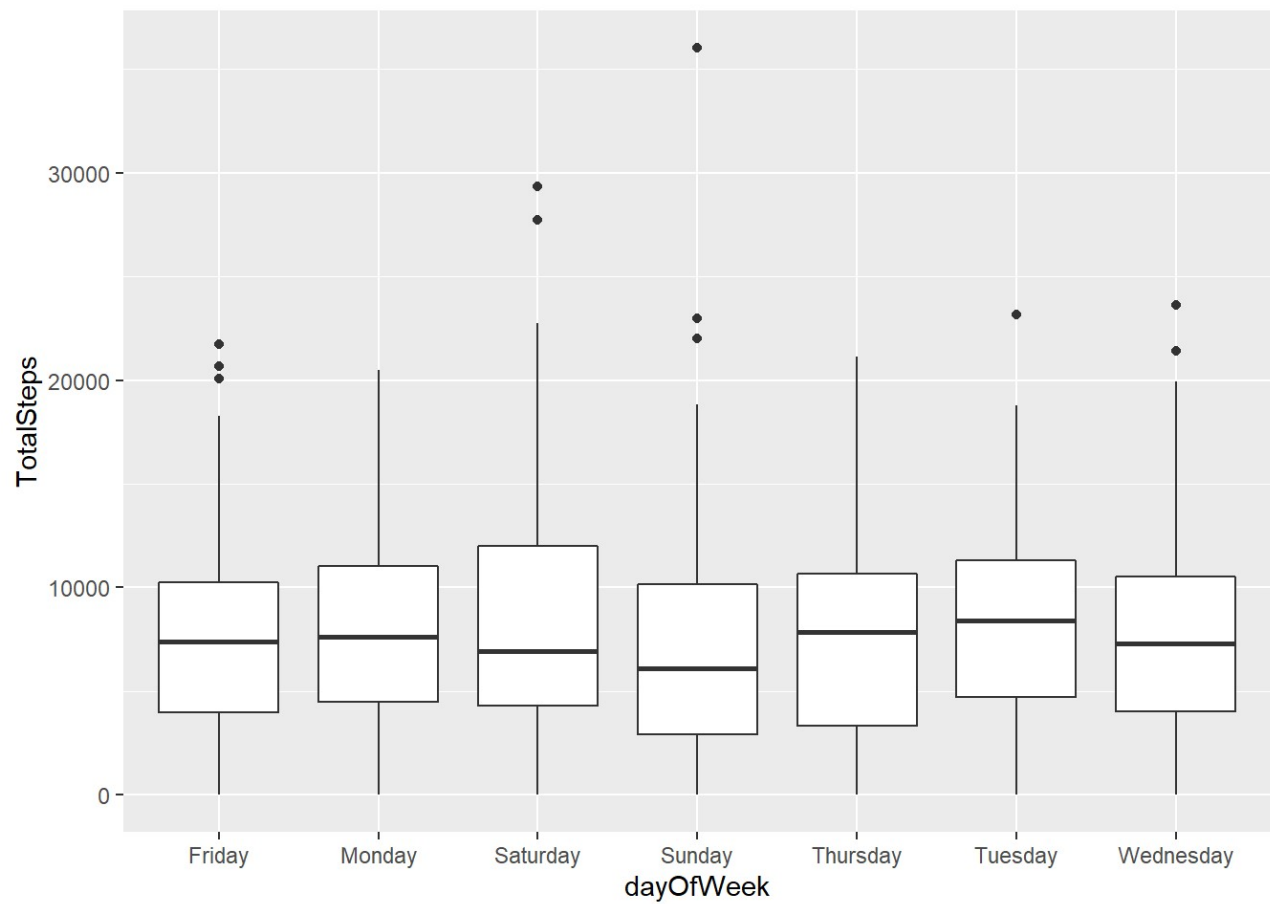
```
##           Id ActivityDate TotalSleepRecords TotalMinutesAsleep TotalTimeInBed
## 1 1503960366 2016-04-12           1           327           34
## 2 1503960366 2016-04-13           2           384           40
## 3 1503960366 2016-04-15           1           412           44
## 4 1503960366 2016-04-16           2           340           36
## 5 1503960366 2016-04-17           1           700           71
## 6 1503960366 2016-04-19           1           304           32
##    dayOfWeek
## 1    Tuesday
## 2 Wednesday
## 3    Friday
## 4   Saturday
## 5    Sunday
## 6    Tuesday
```

```
ggplot(data=sleep_day)+geom_bar(mapping=aes(x=dayOfWeek))
```

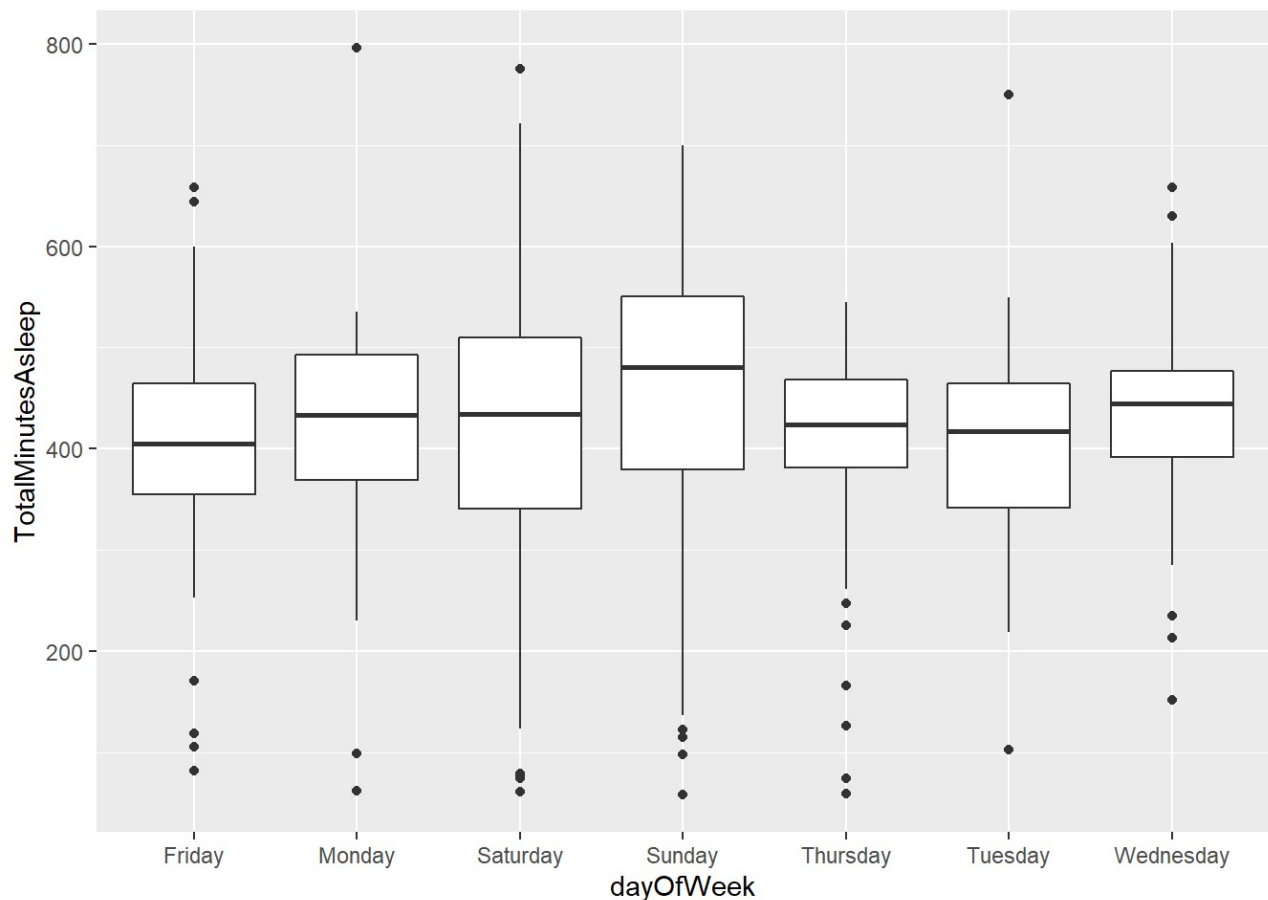


the graph shows that people are more likely to record sleeping times on Tuesday, Wednesday and Thursday.

```
ggplot(data=daily_activity)+geom_boxplot(mapping=aes(x=dayOfWeek,y=TotalSteps))
```



```
ggplot(data=sleep_day)+geom_boxplot(mapping=aes(x=dayOfWeek,y=TotalMinutesAsleep))
```



these graphs shows that people record less activity and longer sleeping times on weekend.

Share

from the results above we can see

- people are more likely to quite recording their sleeping times than activities.
- people are more likely to record excierse on Tuesday, Wednesday and Thursday.

From the analysis I make below suggestions.

- Emphasize more on activity record in advertise and products design.

Thanks google for provide me this oportunity to do this analysis. Thanks Mobius for share this dataset.