Reproducible Research: Peer Assessment 1

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```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
## filter, lag

## The following objects are masked from 'package:base':
##
intersect, setdiff, setequal, union
```

Load the data and do basic exploration

We directly read in the zip file be use of *unzip*

```
data <- read.csv(unzip("./activity.zip"))</pre>
```

Next we have a look at the data and how it is formatted

head(data)

```
steps
                  date interval
## 1
        NA 2012-10-01
## 2
        NA 2012-10-01
                              5
        NA 2012-10-01
                             10
        NA 2012-10-01
                             15
        NA 2012-10-01
                             20
## 5
        NA 2012-10-01
## 6
                             25
```

tail(data)

```
date interval
##
         steps
## 17563
            NA 2012-11-30
                               2330
## 17564
            NA 2012-11-30
                               2335
## 17565
            NA 2012-11-30
                               2340
            NA 2012-11-30
## 17566
                               2345
            NA 2012-11-30
## 17567
                               2350
            NA 2012-11-30
## 17568
                               2355
```

Obviously, missing values in the steps variable are encoded as NA. I personally prefer NaN for missing values so lets change this first:

```
data$steps[is.na(data$steps)] <- NaN
```

Analysis of the daily activity

We use the dplyr package to analyse our data per day. First, we group the date by the date variable and afterwards we write a summary and look at the first values:

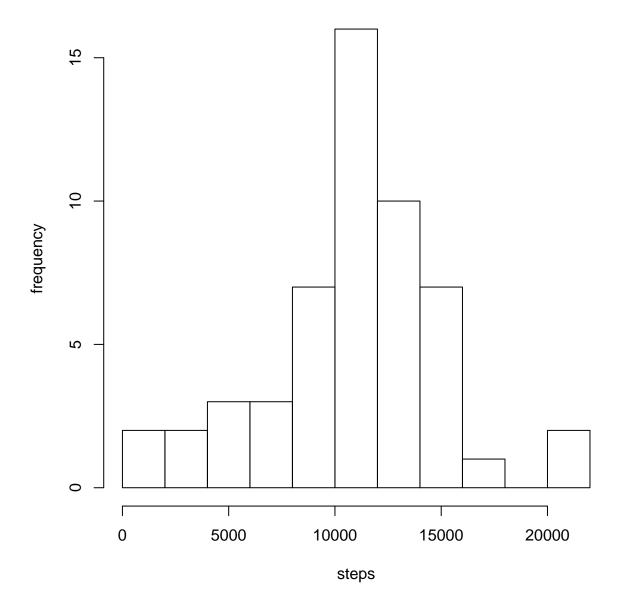
```
dated <- group_by(data,date)
summary <- summarize(dated,means=(mean(steps,na.rm=TRUE)),medians=median(steps,na.rm=TRUE),sums=sum(steps,na.rm=TRUE)
head(summary)</pre>
```

```
## Source: local data frame [6 x 4]
##
##
                   means medians sums
           date
##
         (fctr)
                   (dbl)
                            (dbl) (dbl)
## 1 2012-10-01
                               NA
                      NA
                                    {\tt NaN}
## 2 2012-10-02 0.43750
                                0
                                    126
## 3 2012-10-03 39.41667
                                0 11352
## 4 2012-10-04 42.06944
                                0 12116
## 5 2012-10-05 46.15972
                                0 13294
## 6 2012-10-06 53.54167
                                0 15420
```

Lets see how active the subject is by plotting a simple histogramm of the total number of steps taken per day, as stored in summary\$sums

```
hist(summary$sums,breaks=10,xlab = "steps",ylab="frequency")
```

Histogram of summary\$sums



And now we calculate the mean and median of the total number of steps taken per day. Note, that we ignore NaN for the calculation. Otherwise, both, mean and median would be NaN!

```
mean_totals <- mean(summary$sums,na.rm=TRUE)
mean_totals
## [1] 10766.19</pre>
```

```
median_totals <- median(summary$sums,na.rm=TRUE)
median_totals</pre>
```

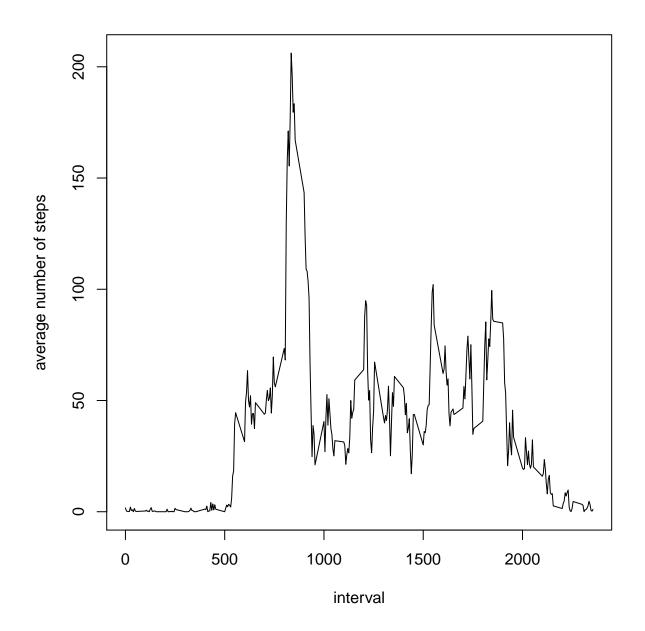
[1] 10765

As can be seen, mean and median are pretty close to each other. The interval with the maximum number of steps is

What is the average daily activity pattern?

Next we would like to see how active the subject is in different phases of a day. So we again group our data, this time by the variable interval and we summarize the average number of steps and plot the result

```
intervaled <- group_by(data, interval)
summary2 <- summarize(intervaled,meanPerInterval=(mean(steps,na.rm=TRUE)),medianPerInterval=(median(steps))
plot(summary2$interval,summary2$meanPerInterval,type="l",xlab = "interval",ylab="average number of steps")</pre>
```



What is the interval with the maximum number of steps on average?

The interval can be determined with the following code:

summary2\$interval[summary2\$meanPerInterval==max(summary2\$meanPerInterval)]

[1] 835

Treating missing values

How many missing values do we have?

```
sum(is.na(data$steps))
```

```
## [1] 2304
```

As we have seen, there are a lot of missing values in our data set. In total we have 2304 of them. Since missing values may bias our analysis we may replace them with some resonable value. We choose the following approach: - find a missing value - replace it with the mean value of its 5-minute interval

```
na_logi <- is.na(data$steps)
rep_summary2 <- rep_len(summary2$meanPerInterval,length.out = length(data$steps))
data <- mutate(data,steps_na_rm=steps)
data$steps_na_rm[na_logi] <- rep_summary2[na_logi]
sum(is.na(data$steps_na_rm))</pre>
```

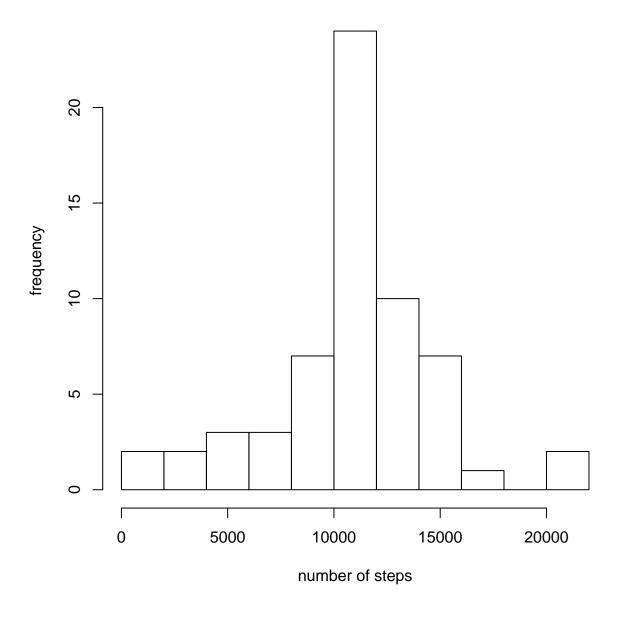
```
## [1] 0
```

```
dated <- group_by(data,date)
summary <- summarize(dated,means=(mean(steps_na_rm)),medians=median(steps_na_rm))
sums=sum(steps_na_rm))
head(summary)</pre>
```

```
## Source: local data frame [6 x 4]
##
##
           date
                   means medians
                                      sums
##
         (fctr)
                            (dbl)
                                     (dbl)
                   (dbl)
## 1 2012-10-01 37.38260 34.11321 10766.19
## 2 2012-10-02 0.43750 0.00000
                                    126.00
## 3 2012-10-03 39.41667 0.00000 11352.00
## 4 2012-10-04 42.06944 0.00000 12116.00
## 5 2012-10-05 46.15972 0.00000 13294.00
## 6 2012-10-06 53.54167 0.00000 15420.00
```

```
hist(summary$sums,breaks=10,xlab = "number of steps",ylab="frequency")
```

Histogram of summary\$sums



```
mean_totals_na_rm <- mean(summary$sums)
mean_totals_na_rm

## [1] 10766.19

median_totals_na_rm <- median(summary$sums)
median_totals_na_rm</pre>
```

[1] 10766.19

Differences between workdays and weekend

To analyze the activity patterns and find differences between workdays and weekends we first of to add a new variable to our data set which stores the weekday abbreviation:

```
data <- mutate(data,wkd=weekdays(as.POSIXct(data$date),abbreviate = TRUE))</pre>
```

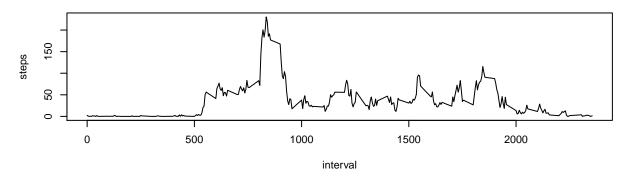
Now we have to reshape our a data a little bit, such that we can compare weekends and workdays directly

```
weed <- filter(data, wkd=="Sa" | wkd =="So")
wrkd <- filter(data,wkd!="Sa" & wkd!="So")
weed <- group_by(weed,interval)
wrkd <- group_by(wrkd,interval)
weed <- summarise(weed,weed_meansteps=mean(steps_na_rm))
wrkd <- summarise(wrkd,wrkd_meansteps=mean(steps_na_rm))
alldays <- mutate(wrkd,weed_meansteps=wrkd_meansteps)
alldays$weed_meansteps <- weed$weed_meansteps
alldays <- mutate(alldays, difference=wrkd_meansteps-weed_meansteps)</pre>
```

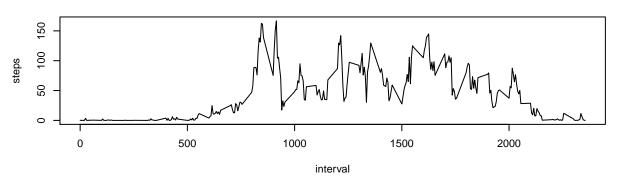
Lets plot the result as well as the difference between work- and weekenddays (workdays-weekends):

```
par(mfrow=c(3,1))
plot(alldays$interval,alldays$wrkd_meansteps,type="l",xlab = "interval",ylab="steps",main="Average step
plot(alldays$interval,alldays$weed_meansteps,type="l",xlab = "interval",ylab="steps",main="Average step
plot(alldays$interval,alldays$difference,type="l",xlab = "interval",ylab="workdays-weekends",main="Diff
```

Average steps taken on workdays



Average steps taken on weekends



Difference between workdays and weekends

