### Reproducible Research: Peer Assessment 1

• load all the libraries

```
is.installed <- function(mypkg) is.element(mypkg, installed.packages()[,1])
if (is.installed('dplyr') == 'FALSE') {install.packages("dplyr")} else{library(dplyr)}
## Warning: package 'dplyr' was built under R version 3.2.2
## 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
if (is.installed('ggthemes') == 'FALSE') {install.packages("ggthemes")} else{library(ggthemes)}
## Warning: package 'ggthemes' was built under R version 3.2.2
## Loading required package: ggplot2
## Warning: package 'ggplot2' was built under R version 3.2.2
if (is.installed('scales') == 'FALSE') {install.packages("scales")} else{library(scales)}
if (is.installed('RColorBrewer') == 'FALSE')
 {install.packages("RColorBrewer")} else{library(RColorBrewer)}
if (is.installed('lubridate') == 'FALSE') {install.packages("lubridate")} else{library(lubridate)}
if (is.installed('ggplot2') == 'FALSE') {install.packages("ggplot2")} else{library(ggplot2)}
if (is.installed('plyr') == 'FALSE') {install.packages("plyr")} else{library(plyr)}
## -----
## You have loaded plyr after dplyr - this is likely to cause problems.
## If you need functions from both plyr and dplyr, please load plyr first, then dplyr:
## library(plyr); library(dplyr)
##
## Attaching package: 'plyr'
## The following object is masked from 'package:lubridate':
##
##
      here
##
## The following objects are masked from 'package:dplyr':
##
      arrange, count, desc, failwith, id, mutate, rename, summarise,
##
##
      summarize
```

```
if (is.installed('knitr') == 'FALSE') {install.packages("knitr")} else{library(knitr)}

## Warning: package 'knitr' was built under R version 3.2.2

if (is.installed('lattice') == 'FALSE') {install.packages("lattice")} else{library(lattice)}

if (is.installed('RCurl') == 'FALSE') {install.packages("RCurl")} else{library(RCurl)}

## Loading required package: bitops
```

- Load the data (i.e. read.csv())
- Process/transform the data (if necessary) into a format suitable for your analysis
- Set working directory

• Read the CSV

```
data <- read.csv(paste(curdir,'/activity.csv',sep=""))</pre>
```

• Ignore missing value

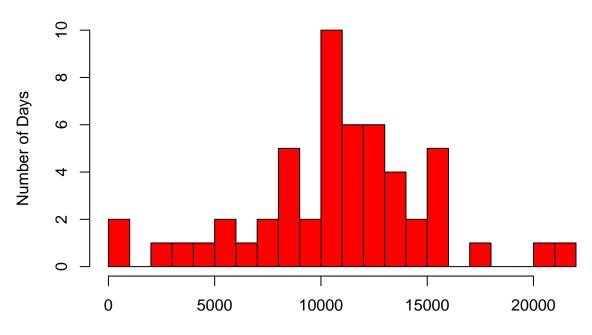
```
dataClean <- subset(data, is.na(data$steps) == F)
totalPerDay <- ddply(dataClean, .(date), summarise, steps=sum(steps))</pre>
```

### What is mean total number of steps taken per day?

- If you do not understand the difference between a histogram and a barplot, research the difference between them. Make a histogram of the total number of steps taken each day
- Plot / Make a histogram of the total number of steps taken each day

```
hist(totalPerDay$steps , breaks = 20, main="Number of Steps", xlab="Total number of steps taken each day", ylab = "Number of Days",col="red")
```

# **Number of Steps**



Total number of steps taken each day

- Calculate and report the mean and median of the total number of steps taken per day
- Mean

### mean(totalPerDay\$steps)

#### ## [1] 10766.19

• Median

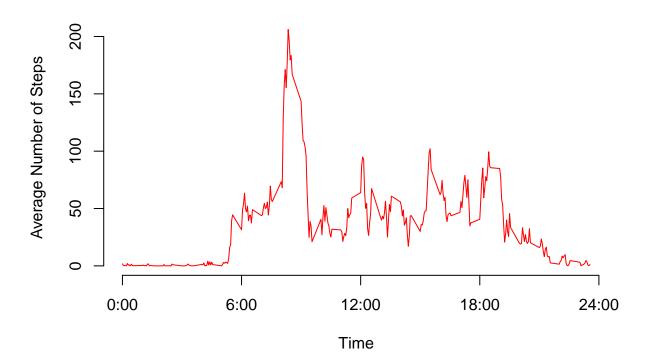
#### median(totalPerDay\$steps)

## [1] 10765

### What is the average daily activity pattern?

- Make a time series plot (i.e. type = "l") of the 5-minute interval (x-axis) and the average number of steps taken, averaged across all days (y-axis)
- Which 5-minute interval, on average across all the days in the dataset, contains the maximum number of steps?

## **Average Daily Activity Pattern**



maxSteps <- averagePerInterval[which.max(averagePerInterval\$steps),] # 8.35 + 5-minute = (8.35-8.40)

### Imputing missing values

• Calculate and report the total number of missing values in the dataset (i.e. the total number of rows with NAs)

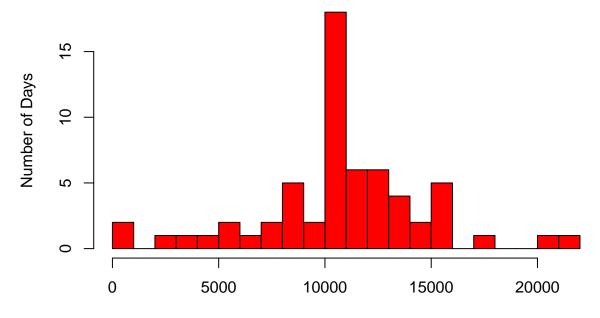
```
missingvalCount <- sum(is.na(data$steps)) # 2304
```

- Devise a strategy for filling in all of the missing values in the dataset. The strategy does not need to be sophisticated. For example, you could use the mean/median for that day, or the mean for that 5-minute interval, etc.
- Fill NA with the average value for each 5 minutes interval
- Create a new dataset that is equal to the original dataset but with the missing data filled in.

- Make a histogram of the total number of steps taken each day and Calculate and report the mean and median total number of steps taken per day.
- Do these values differ from the estimates from the first part of the assignment?
- What is the impact of imputing missing data on the estimates of the total daily number of steps?
- $\bullet \quad Total Steps Missing Value Fill in$

• Trying plot the data to investigate

## **Number of Steps**



Total number of steps taken each day

```
mean(totalPerDayStepsMissingvalueFillin$steps) # 10766.19
## [1] 10766.19
median(totalPerDayStepsMissingvalueFillin$steps) # 10766.19
## [1] 10766.19
abs(mean(totalPerDay$steps)-
      mean(totalPerDayStepsMissingvalueFillin$steps)) # 0
## [1] 0
abs(median(totalPerDay$steps) - median(totalPerDayStepsMissingvalueFillin$steps))/
  median(totalPerDay$steps)
## [1] 0.0001104207
#0.0001104207
totalDifference <- sum(totalPerDayStepsMissingvalueFillin$steps)
- sum(dataClean$steps) # 86129.51
## [1] -570608
totalDifference
## [1] 656737.5
```

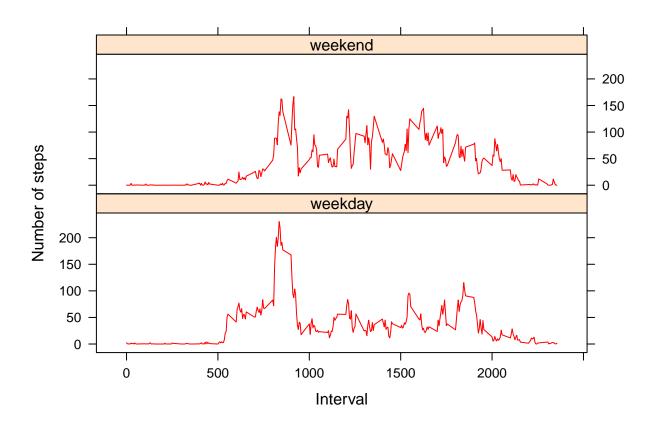
### Are there differences in activity patterns between weekdays and weekends?

• Create a new factor variable in the dataset with two levels - "weekday" and "weekend" indicating whether a given date is a weekday or weekend day.

```
## [1] "English_United States.1252"

missingValFillin$weekdays <- weekdays(as.Date(missingValFillin$date))
missingValFillin$weekdays <- ifelse(missingValFillin$weekdays
%in% c("Saturday", "Sunday"),"weekend", "weekday")
average <- ddply(missingValFillin, .(interval, weekdays), summarise, steps=mean(steps))</pre>
```

- Make a panel plot containing a time series plot (i.e. type = "l") of the 5-minute interval (x-axis) and the average number of steps taken, averaged across all weekday days or weekend days (y-axis).
- See the README file in the GitHub repository to see an example of what this plot should look like using simulated data.



End of reporting