# Reproducible Research: Peer Assessment 1

### Loading and preprocessing the data

First of all, the following code downloads and pre-prosesses the Personal Activity data:

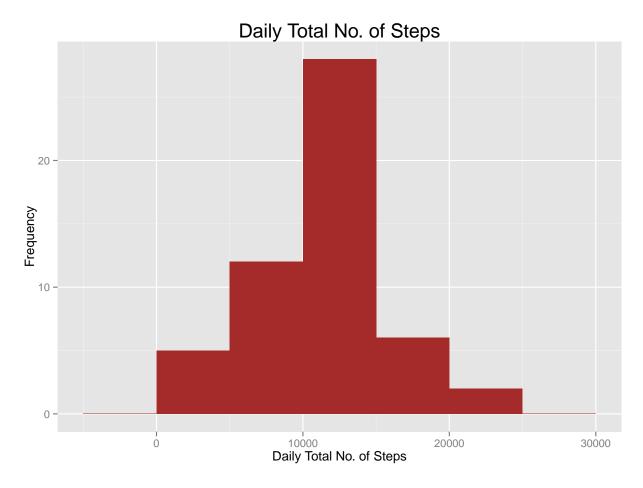
```
# Set data source URL and file paths
url <- 'https://github.com/luongthevinh/RepData_PeerAssessment1/raw/master/activity.zip'
fileName <- 'activity.csv'

# Download data
setInternet2(TRUE)
tempFile <- tempfile()
download.file(url, tempFile)
dat <- read.csv(unzip(tempFile, fileName))
dat$date <- as.Date(dat$date)

# Load the "reshape2" and "ggplot2" packages for reshaping and plotting data
library(reshape2)
library(ggplot2)</pre>
```

#### What are the mean and median total numbers of steps taken per day?

Let's now look at a histogram of the subject's total number of steps taken each day during the months of October-November 2012:

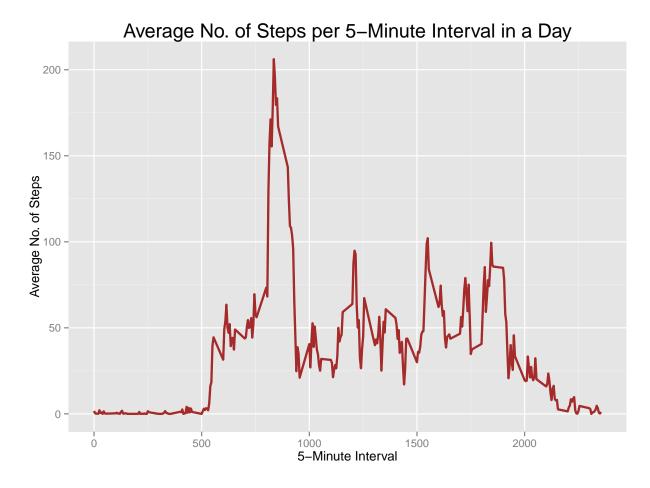


```
# Compute mean and median of Daily Total No. of Steps
dailyNumSteps_mean <- mean(dailyNumSteps$steps)
dailyNumSteps_med <- median(dailyNumSteps$steps)
```

During the concerned time period, the subject's average and median numbers of steps taken daily were 10,766 and 10,765, respectively.

#### What is the average daily activity pattern?

Next, let's chart out the subject's average activity pattern during a day, broken into 5-minute time intervals:



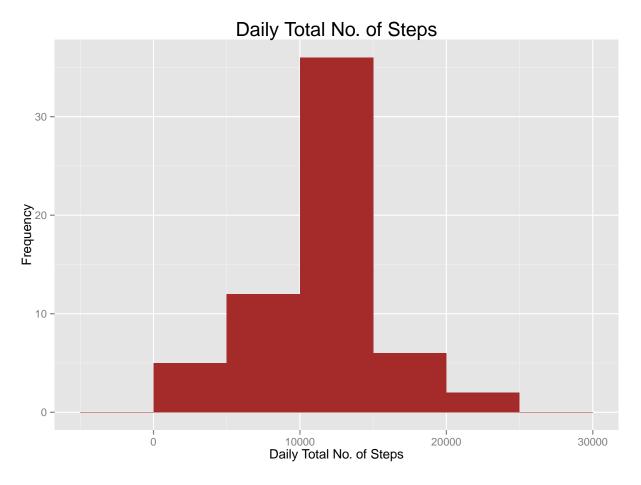
During such a typical day, the peak activity level occurs at around **08:35**.

## Imputing missing values

```
# Identify rows with missing data
datWithNAs_melt <- melt(dat, id = c('date', 'interval'))
naRowNums <- is.na(datWithNAs_melt$value)</pre>
```

The original data source contains 2,304 missing values. In order to reduce the potential bias caused by such missing values, we impute each missing data point with the average number of steps in the same time interval across days in the observed time period.

Let's now re-look at the histogram of the subject's total number of steps taken each day, with the missing data points imputed by the above process:



```
# Compute mean and median of Daily Total No. of Steps
dailyNumSteps_mean <- mean(dailyNumSteps$steps)
dailyNumSteps_med <- median(dailyNumSteps$steps)</pre>
```

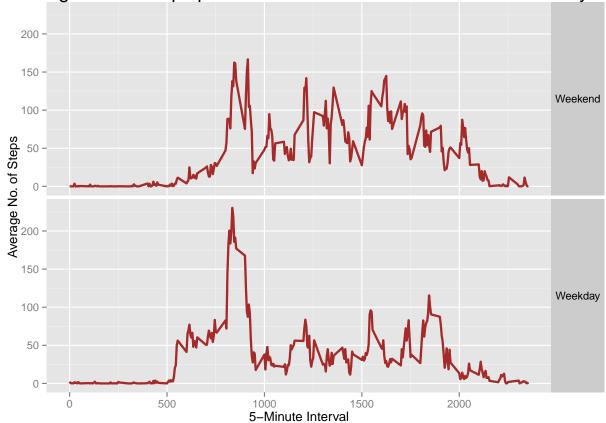
With the imputed data set, the subject's average and median number of steps taken daily were 10,766 and 10,766, respectively. These values are very close to those computed for case with missing data. This is likely to be because of our choosing the "averaging" imputing method.

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

Let's now assess if the daily activity pattern differs between working weekdays and weekends.

```
# Plot daily activity patterns during weekends vs. weekdays
datImputed_melt$wkday_or_wkend <-
        as.factor((as.POSIXlt(datImputed_melt$date)$wday %in% 1:5) + 1)
levels(datImputed_melt$wkday_or_wkend) <- c("Weekend", "Weekday")
dailyNumSteps_avgByInterval <- dcast(datImputed_melt,
        wkday_or_wkend + interval ~ variable, mean)
ggplot(dailyNumSteps_avgByInterval, aes(interval, steps)) +
        geom_line(size = 1, colour = 'brown') +
        facet_grid(wkday_or_wkend ~ .) +
        theme(plot.title = element_text(size = rel(1.5)), strip.text.y = element_text(angle=0)) +
        ggtitle("Average No. of Steps per 5-Minute Interval: Weekend vs. Weekday") +
        xlab("5-Minute Interval") + ylab("Average No. of Steps")</pre>
```

Average No. of Steps per 5-Minute Interval: Weekend vs. Weekday



This comparison hows that the subject was much more active during weekends than he/she was during weekdays. This is perhaps because the subject is a white-collar worker and spent most of his/her late-morning and afternoon time sitting down at a stationary workstation.