Pre-processing and clean-up

This first step downloads the required libraries and data files.

```
## required libraries
library(ggplot2)
library(dplyr)
```

```
## Warning: package 'dplyr' was built under R version 3.1.2
```

```
##
## Attaching package: 'dplyr'
##
## The following object is masked from 'package:stats':
##
## filter
##
## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union
```

```
## download file
zipurl <- "http://d396qusza40orc.cloudfront.net/repdata%2Fdata%2Factivity.zip"
download.file(zipurl, "temp.zip", mode="wb")
unzip("temp.zip", "activity.csv")
dd <- read.table("activity.csv", sep=",", header=T)</pre>
```

Next, we remove NAs and take a quick look at the data using the R summary command.

```
## remove NAs
d <- dd[complete.cases(dd), ]

## quick look
summary(d)</pre>
```

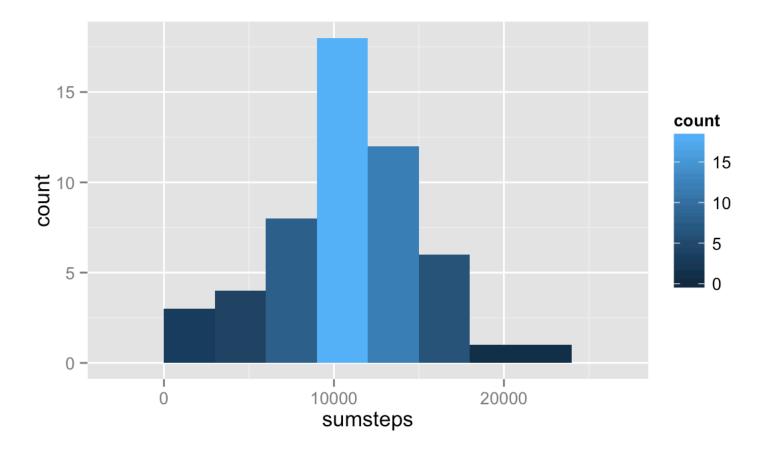
```
steps
                            date
                                           interval
           : 0.0
                    2012-10-02:
                                  288
##
                                        Min.
                                               :
    1st Qu.:
              0.0
                    2012-10-03:
                                  288
                                        1st Qu.: 589
    Median :
                    2012-10-04:
                                  288
                                        Median :1178
##
              0.0
##
    Mean
           : 37.4
                    2012-10-05:
                                  288
                                        Mean
                                               :1178
##
    3rd Qu.: 12.0
                    2012-10-06:
                                  288
                                        3rd Qu.:1766
##
    Max.
           :806.0
                    2012-10-07:
                                  288
                                        Max.
                                               :2355
##
                    (Other) :13536
```

What is mean total number of steps taken per day?

The code chunk below creates a summary of the steps taken per day, creates a histogram, and calculates both the **mean** and **median** steps per day.

```
## summarize data
ds <- d %>% group_by(date) %>% summarise(sumsteps=sum(steps))

## histogram
dsum <- ggplot(data=ds, aes(x=sumsteps))
dsum + geom_histogram(aes(fill=..count..), binwidth=3000)</pre>
```



```
## mean number of steps taken per day
dmean <- mean(ds$sumsteps) ## mean of total number of steps
dmean</pre>
```

```
## [1] 10766
```

```
dmedian <- median(ds$sumsteps) ## median of total number of steps
dmedian</pre>
```

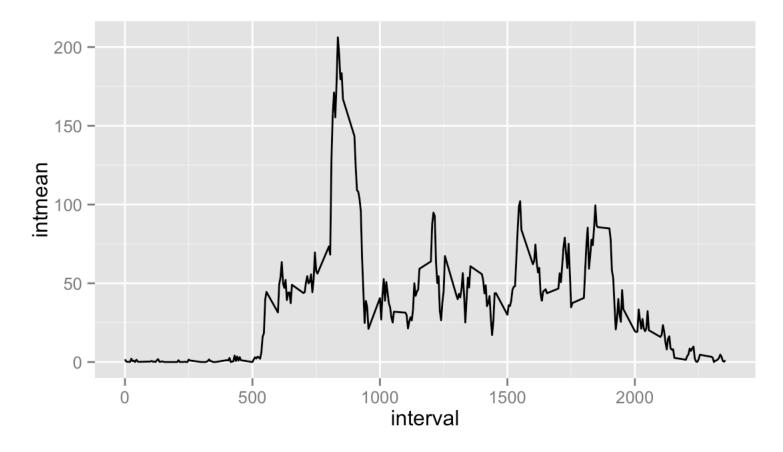
```
## [1] 10765
```

What is the average daily activity pattern?

Make a time series plot (i.e. type = "I") of the 5-minute interval (x-axis) and the average number of steps taken, averaged across all days (y-axis).

```
## summarize and plot
dint <- d %>% group_by(interval) %>% summarise(intmean=mean(steps))

dintline <- ggplot(data=dint, aes(x=interval, y=intmean))
  dintline + geom_line()</pre>
```



Which 5-minute interval, on average across all the days in the dataset, contains the maximum number of steps?

```
## max five minute interval
subset(dint, intmean == max(intmean), select=c(interval, intmean))
```

```
## Source: local data frame [1 x 2]
##
## interval intmean
## 1 835 206.2
```

Imputing missing values

Calculate and report the total number of missing values in the dataset (i.e. the total number of rows with NAs). Then create a new dataset that is equal to the original dataset but with the missing data filled in.

Processing

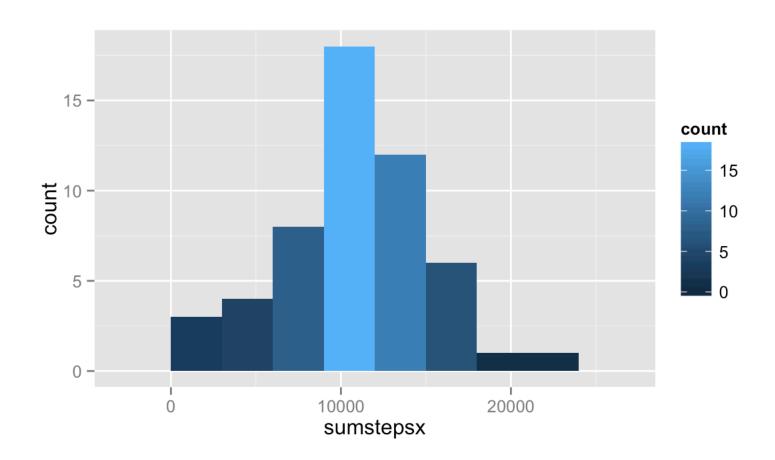
```
nas <- is.na(d) ## create NA index
dcln <- merge(d, dint, by="interval") ## merge in NA index
dcln$stepsx <- as.numeric(dcln$steps)
my.na <- is.na(dcln$steps)
dcln$stepsx[my.na] <- dcln$intmean[my.na] ## replace NAs with mean value</pre>
```

Histogram of total number of steps taken per day

```
dclns <- dcln %>% group_by(date) %>% summarise(sumstepsx=sum(stepsx)) ## summary

dclnshist <- ggplot(data=dclns, aes(x=sumstepsx))

dclnshist + geom_histogram(aes(fill=..count..), binwidth=3000)</pre>
```



```
dmeanx <- mean(dclns$sumstepsx) ## mean of total number of steps
dmeanx</pre>
```

```
## [1] 10766
```

```
dmedianx <- median(dclns$sumstepsx) ## median of total number of steps
dmedianx</pre>
```

```
## [1] 10765
```

Calculate difference between the mean and median of the original dataset vs. the revised dataset with inputted values

```
## difference in mean and median values with and without inputing
dmeandiff <- dmean - dmeanx
dmediandiff <- dmedian - dmedianx</pre>
```

The difference ends up being zero, i.e. there is no difference between the original mean and median values from the ones with inputed value. This is because the inputed values are the same as the mean and median from the original dataset.

```
original mean - inputted mean = 0
original median - inputted median = 0
```

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.

The differences are clear when looking at the two line plots below.

