# Reproducible Research: Activity Monitoring

#### Daniel Escasa

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#### {r setup, include=FALSE} knitr::opts\_cnunk\$set(echo = IRU

### Introduction

Excerpts from the original repo from which this was forked

It is now possible to collect a large amount of data about personal movement using activity monitoring devices such as a Fitbit, Nike Fuelband, or Jawbone Up. These type of devices are part of the "quantified self" movement – a group of enthusiasts who take measurements about themselves regularly to improve their health, to find patterns in their behavior, or because they are tech geeks. But these data remain under-utilized both because the raw data are hard to obtain and there is a lack of statistical methods and software for processing and interpreting the data.

This project makes use of data from a personal activity monitoring device which collects data at five-minute intervals through out the day.

The data consists of two months of data from an anonymous individual collected during the months of October and November, 2012 and include the number of steps taken in five-minute intervals each day.

Variable name	Description
steps	Number of steps taking in a five-minute interval (missing values are coded as
	NA)
date	The date on which the measurement was taken, in YYYY-MM-DD format
interval	Identifier for the five-minute interval in which measurement was taken

The following are the questions to be addressed:

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

- What is the average daily activity pattern?
- Are there differences in activity patterns between weekdays and weekends?

## 1 Boring admin stuff

#### 1.1 Install the needed libraries

```
if (!require("ggplot2")) {
   message("Installing ggplot2")
    install.packages("ggplot2")
}
## Loading required package: ggplot2
if (!require("knitr")) {
  message("Installing knitr")
  install.packages("knitr")
}
## Loading required package: knitr
if (!require("scales")) {
  message("Installing scales")
  install.packages("scales")
}
## Loading required package: scales
if (!require("numform")) {
  message("Installing numform")
  install.packages("numform")
}
## Loading required package: numform
if (!require("timeDate")) {
  message("Installing timeDate")
  install.packages("timeDate")
}
```

### ## Loading required package: timeDate

#### 1.2 Download the zipped dataset if not yet present

### 1.3 Unzip the dataset if not yet present

```
if (!file.exists("activity.csv")) {
    message("Extracting dataset")
```

#### 1.4 Read in the data

```
# Shouldn't need error checking here
activity <- read.csv("activity.csv")</pre>
```

#### 1.5 Load the libraries

```
library(ggplot2)
library(knitr)
library(scales)
library(numform)
library(timeDate)
```

## 2 Now let's get cooking

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

Compute the total number, mean, and median steps per day. As an aside, take note of the use of the superassignment (<<-) instead of the usual assignment operator (<-). This is because the with(){} creates its own scope. And yes, the scope applies to assigning to colnames().

```
with(data = activity,{
    totalDaySteps <<- aggregate(steps, by = list(date), FUN = sum)
    meanDaySteps <<- aggregate(steps, by = list(date), FUN = mean)
    medianDaySteps <<- aggregate(steps, by = list(date), FUN = median)
})</pre>
```

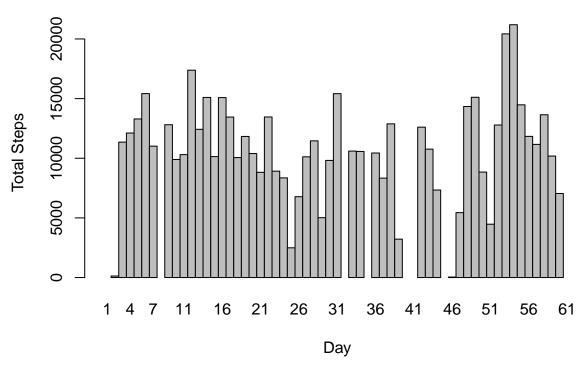
For convenience, merge the three into one dataframe and assign meaningful colnames().

```
dayStats <- merge(totalDaySteps, meanDaySteps, by = "Group.1")
dayStats <- merge(dayStats, medianDaySteps, by = "Group.1")
colnames(dayStats) <- c("date", "totalSteps", "meanSteps", "medianSteps")</pre>
```

Plot total number of steps taken each day, telling barplot() to remove spaces between bars. Also, using (as.Date(date) - as.Date(totalDaySteps[1, 1]) + 1) as names.arg so labels on x axis are day numbers, not the dates. This is so the plot is neater.

```
with(dayStats,
    barplot(space = 0, totalSteps,
        main = "Total Number of Steps per Day",
        xlab = "Day", ylab = "Total Steps",
        names.arg = (as.Date(date) - as.Date(totalDaySteps[1, 1]) + 1)))
```

# **Total Number of Steps per Day**



```
dev.copy(png, "plots/01-totalSteps.png")

## png
## 3

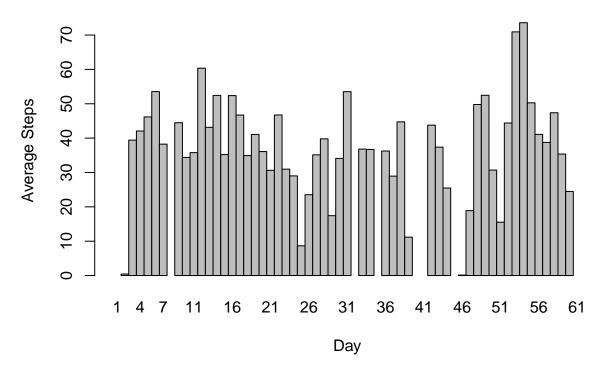
dev.off()

## pdf
## 2

Plot mean of number of steps per day.

with(dayStats,
    barplot(space = 0, meanSteps,
        main = "Average Number of Steps per Day",
        xlab = "Day", ylab = "Average Steps",
        names.arg = (as.Date(date) - as.Date(totalDaySteps[1, 1]) + 1)))
```

# **Average Number of Steps per Day**



```
dev.copy(png, "plots/02-meanSteps.png")
## png
## 3
dev.off()
## pdf
## 2
```

And here's the dataframe of average number of steps per day:

```
kable(dayStats[, c(1, 3)],
     col.names = c("date", "mean"))
```

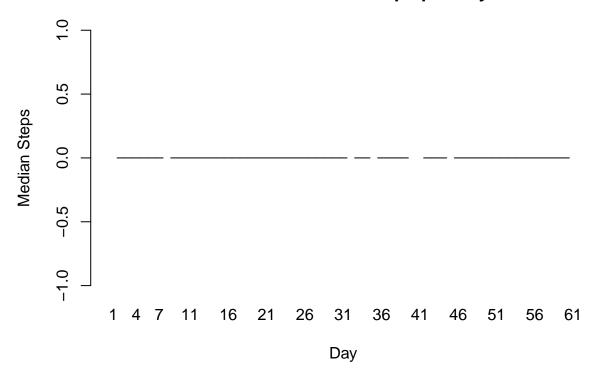
date	mean
2012-10-01	NA
2012-10-02	0.4375000
2012-10-03	39.4166667
2012-10-04	42.0694444
2012-10-05	46.1597222
2012-10-06	53.5416667
2012-10-07	38.2465278
2012-10-08	NA
2012-10-09	44.4826389
2012-10-10	34.3750000
2012-10-11	35.7777778
2012-10-12	60.3541667

date	mean
2012-10-13	43.1458333
2012-10-14	52.4236111
2012-10-15	35.2048611
2012-10-16	52.3750000
2012-10-17	46.7083333
2012-10-18	34.9166667
2012-10-19	41.0729167
2012-10-20	36.0937500
2012-10-21	30.6284722
2012-10-22	46.7361111
2012-10-23	30.9652778
2012-10-23	29.0104167
2012-10-24	8.6527778
2012-10-26	23.5347222
2012-10-20	35.1354167
2012-10-27	39.7847222
2012-10-28	17.4236111
2012-10-29	34.0937500
2012-10-30	53.5208333
2012-10-31	
2012-11-01 2012-11-02	NA
	36.8055556
2012-11-03	36.7048611
2012-11-04	NA
2012-11-05	36.2465278
2012-11-06	28.9375000
2012-11-07	44.7326389
2012-11-08	11.1770833
2012-11-09	NA
2012-11-10	NA
2012-11-11	43.7777778
2012-11-12	37.3784722
2012-11-13	25.4722222
2012-11-14	NA
2012-11-15	0.1423611
2012-11-16	18.8923611
2012-11-17	49.7881944
2012-11-18	52.4652778
2012-11-19	30.6979167
2012-11-20	15.5277778
2012-11-21	44.3993056
2012 - 11 - 22	70.9270833
2012-11-23	73.5902778
2012-11-24	50.2708333
2012-11-25	41.0902778
2012-11-26	38.7569444
2012-11-27	47.3819444
2012-11-28	35.3576389
2012-11-29	24.4687500
2012-11-30	NA

Lastly, plot the median number of steps per day.

```
with(dayStats,
    barplot(space = 0, medianSteps,
        main = "Median Number of Steps per Day",
        xlab = "Day", ylab = "Median Steps",
        names.arg = (as.Date(date) - as.Date(totalDaySteps[1, 1]) + 1)))
```

## **Median Number of Steps per Day**



```
dev.copy(png, "plots/03-medianSteps.png")

## png
## 3
dev.off()

## pdf
## 2
```

Median is zero because of the large number of zero steps — after all, steps is mostly zero, save for an outlier or two, for interval from 0 to 530. NAs may also account for that.

Since the median is zero for all days, there's no point in presenting the dataframe.

The summary() function will also give us similar info:

#### kable(summary(dayStats))

_	date	totalSteps	meanSteps	medianSteps
	2012-10-01: 1	Min.: 41	Min.: 0.1424	Min. :0
	2012-10-02: 1	1st Qu.: 8841	1st Qu.:30.6979	1st Qu.:0
	2012-10-03: 1	Median :10765	Median: 37.3785	Median :0

date	totalSteps	meanSteps	medianSteps
2012-10-04: 1	Mean :10766	Mean :37.3826	Mean :0
2012-10-05: 1	3rd Qu.:13294	3rd Qu.:46.1597	3rd Qu.:0
2012-10-06: 1	Max. :21194	Max. :73.5903	Max. :0
(Other):55	NA's :8	NA's :8	NA's :8

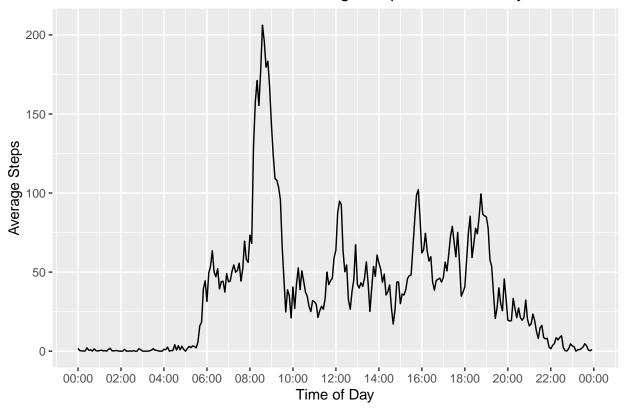
#### 2.2 What is the average daily activity pattern?

Create a dataframe with steps per interval, and plot the line.

Define a function that will take an integer – in this case, the interval – and convert it to a time string.

Average the number of steps over each five-minute interval and plot the time series.





```
dev.copy(png, "plots/04-intervalSteps.png")
## png
## 3
dev.off()
## pdf
```

So, from the plot, on average there is negligible activity from midnight (see below) to around 5:30 in the morning, slowly climbing at around 7:00, and a spike of over 200 steps at around 8:30.

##

From then on, there is a minimum of 25 to a maximum of 100 steps, and activity dies down starting close to 11:30 PM.

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

The starting time and corresponding average number of steps are in intervalSteps[which.max(intervalSteps\$steps), 1] and intervalSteps[which.max(intervalSteps\$steps), 2] respectively. I thought I'd include the end time of that interval by adding five minutes.

```
intervalStart <- intervalSteps[which.max(intervalSteps$steps), 1]
intervalEnd <- intervalSteps[which.max(intervalSteps$steps), 1] + 5
averageSteps <- intervalSteps[which.max(intervalSteps$steps), 2]</pre>
```

Since the steps from interval 0 to interval 600 are mostly zero, I'm betting that the measurements start midnight. If they don't, I'm screwed. Or not, maybe just have to make adjustments in the conversions.

What I'm fairly sure of is that the intervals are of the form hhmm, where hh and mm are the hour and minute components, respectively, of the interval. As evidence, interval on the 24th row is 155, and on the 25th is 200. Examine the 36th and 37th rows, and interval goes from 255 to 300. Where are 160, 165,..., between 15 and 200, or 260, 265,..., between 255 and 300? First, 160 is equivalent to 200, and 260 to 300. That's further evidence that the intervals are time stamps, of the form hhmm.

Output will be in 24-hour format.

## [1] "Maximum number of steps on average taken from 08:35 to 08:40, number of steps = 206.169811"

#### 2.3 Imputing missing values

First question: how many missing values are there?

```
missing <- is.na(activity$steps)
sprintf("Number of missing steps: %s.", f_comma(sum(missing), mark = ","))
## [1] "Number of missing steps: 2,304."
sprintf("Mean: %s.", mean(missing))</pre>
```

## [1] "Mean: 0.131147540983607."

So, do those 2,304 NAs make a difference? Let's find out.

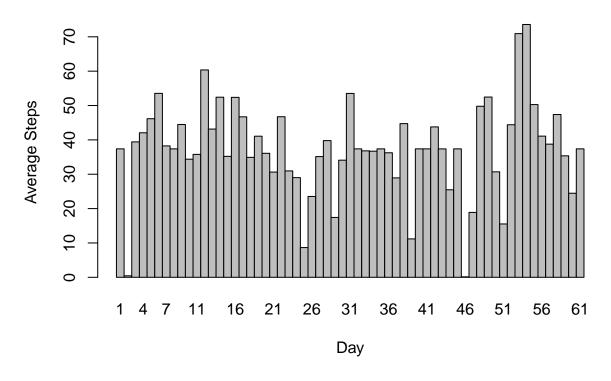
Define a helper function fillNA.

Parameter	Description
steps	Cell to be converted, if NA, to the average number of steps for interval (second parameter)
interval	Identifier for the five-minute interval from which to get the mean number of steps. The function will match intervalSteps\$interval with the parameter interval, and return the average steps if the steps parameter is NA

Make a copy of activity, then apply fillNA to that copy.

Set up the plot, then plot it.

# Average Number of Steps per Day, NA imputed



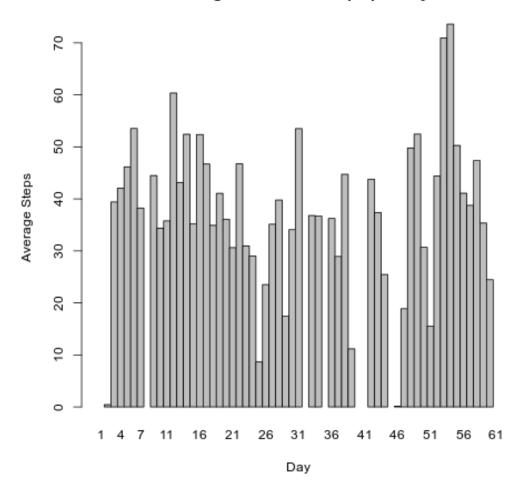
dev.copy(png, "plots/05-imputedSteps.png")

## png ## 3

Compare that with the plot of Average Number of Steps per Day, without imputing NAs:

knitr::include\_graphics("plots/02-meanSteps.png")

### Average Number of Steps per Day



Without the imputed data, average steps for days 1, 8, 32, 35, 40, 41, 45, and 61 are zero because of the NA steps for those days. Imputing creates data points that look reasonable, consistent in height with the other data points.

Meanwhile, the heights for the other days don't show any appreciable change.

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

Create a new Boolean column that determines whether the day is a weekday or weekend.

```
adjActivity$is.Weekday <- timeDate::isWeekday(adjActivity$date)
```

Take the mean number of steps per interval broken down into is.Weekday and !is.Weekday stepsPerInterval <- aggregate(steps ~ interval + is.Weekday, data = adjActivity, mean)

Use the interval column to create a time column.

```
stepsPerInterval$time <- as.POSIXct(int2Time(intervalSteps$interval), format = "%H:%M", tz = "MST")
```

Map the is. Weekday column to the more readable "Weekend" and "Weekday".

## Time Series Plot Comparison of Steps



```
dev.copy(png, "plots/06-WkDayVsWkEnd.png")
```

## png ## 4