

Reproducible Research: Peer Assessment 1

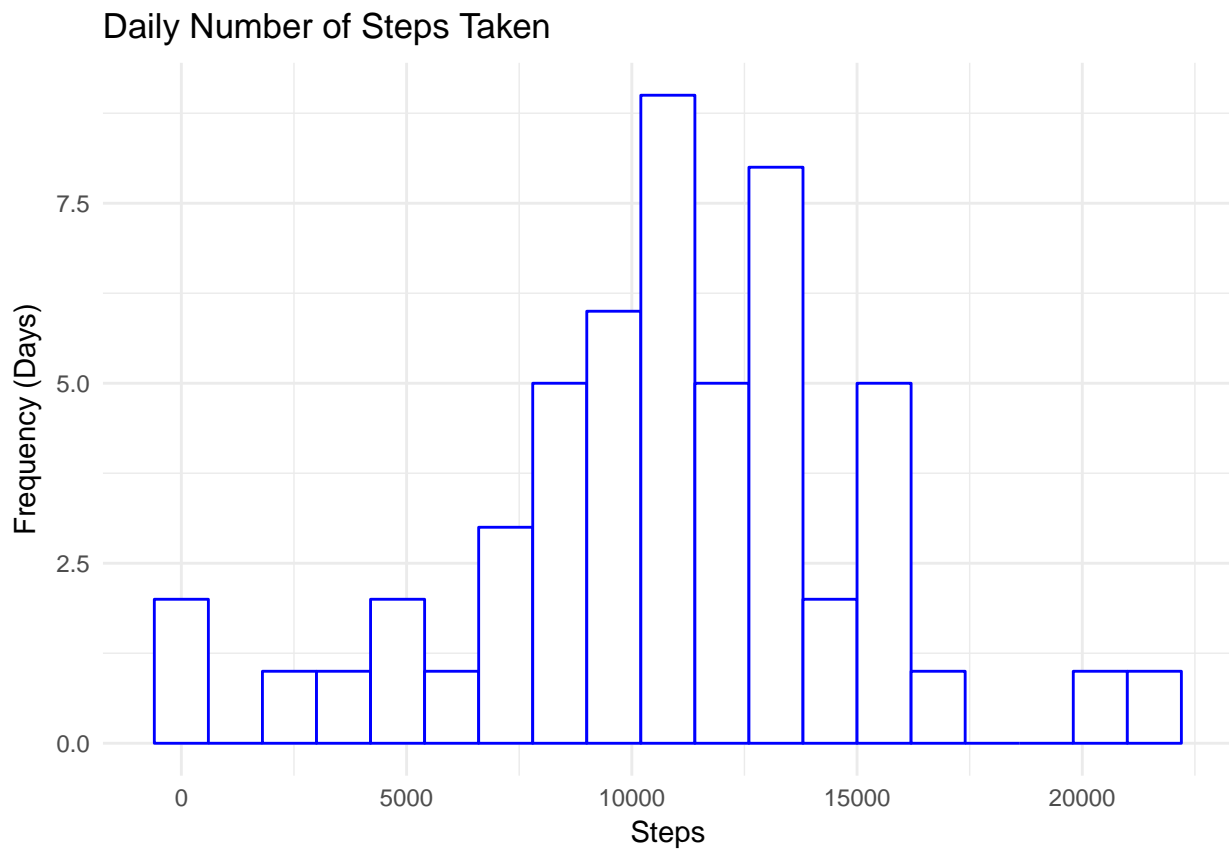
Loading and preprocessing the data

```
library(ggplot2)
data <- read.csv('activity.csv')
aggdata <- aggregate(steps~date, data, FUN=sum, na.rm=TRUE)
```

What is mean total number of steps taken per day?

First get a histogram of the data to give an idea of where the median is, and then compute explicitly.

```
myplot <- ggplot(aggdata, aes(x=steps))
myplot <- myplot + geom_histogram(binwidth=1200, col='blue', fill='white')
myplot <- myplot + labs(title="Daily Number of Steps Taken", x="Steps", y="Frequency (Days)")
myplot <- myplot + theme_minimal()
myplot
```



With my binning choice, the mean and median appear to be in the range [10,000, 11,000].

```
datmean <- mean(aggdata$steps, na.rm=TRUE)
datmed <- median(aggdata$steps, na.rm=TRUE)
datmean
```

```
## [1] 10766.19
```

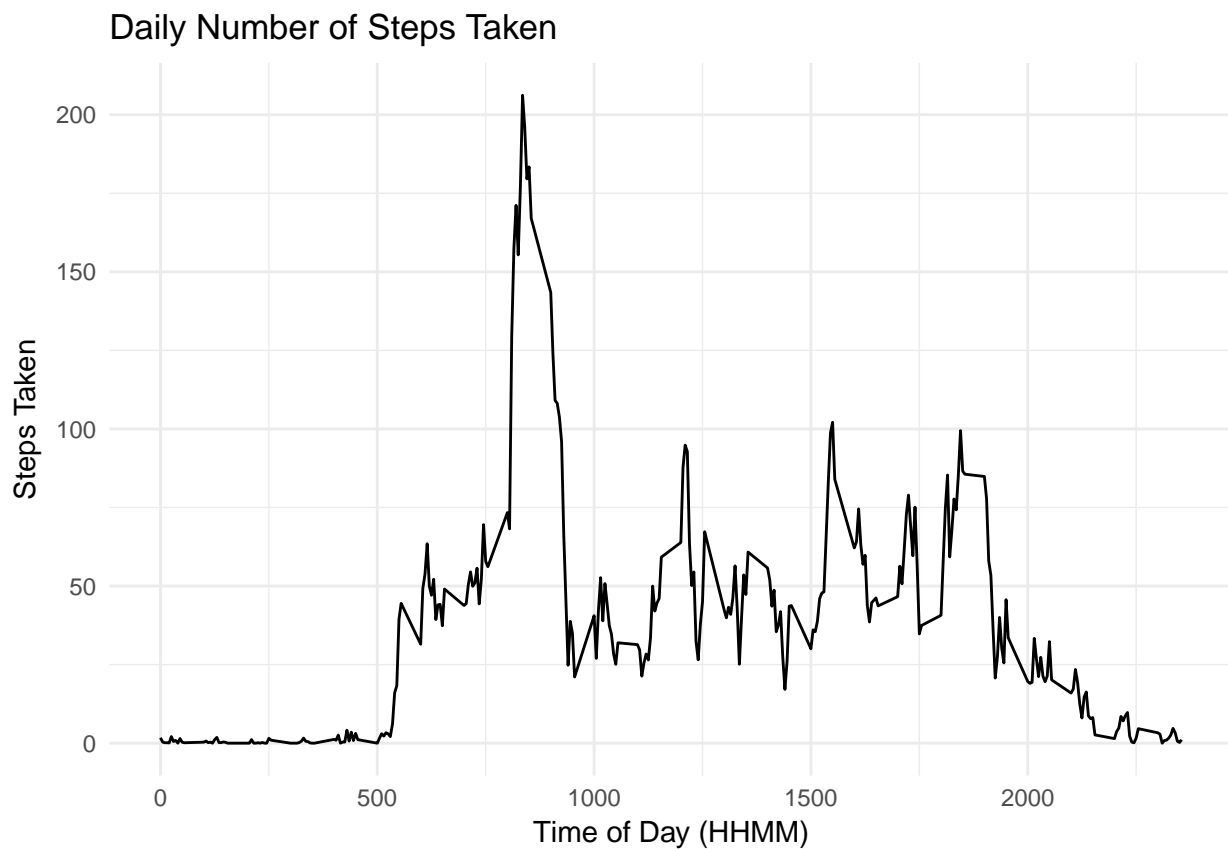
```
datmed
```

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

What is the average daily activity pattern?

Aggregate the data again, but use FUN=mean

```
aggdata <- aggregate(steps~interval, data, FUN=mean, na.rm=TRUE)
myplot<- ggplot(aggdata, aes(x=interval, y=steps)) +geom_line()
myplot <- myplot +theme_minimal()
myplot <- myplot + labs(title="Daily Number of Steps Taken", x="Time of Day (HHMM)", y="Steps Taken")
myplot
```



This person on average took the most steps sometime near 8AM.

```
mostactive <- aggdata$interval[which.max(aggdata$steps)]
mostactive
```

```
## [1] 835
```

Imputing missing values

Ignoring NA values introduces some bias into the dataset because the person was likely still somewhat active during those times that the measurement failed. A simple strategy for inferring where the missing values lie is to determine the median for that time interval.

Get the number of rows that need to have an NA replaced. There are 2304 NA values to replace.

```
summary(data)
```

```
##      steps      date      interval
## Min.   : 0.00 2012-10-01: 288 Min.   : 0.0
## 1st Qu.: 0.00 2012-10-02: 288 1st Qu.: 588.8
## Median : 0.00 2012-10-03: 288 Median :1177.5
## Mean   : 37.38 2012-10-04: 288 Mean   :1177.5
## 3rd Qu.: 12.00 2012-10-05: 288 3rd Qu.:1766.2
## Max.   :806.00 2012-10-06: 288 Max.   :2355.0
## NA's   :2304   (Other)   :15840
```

Use tapply to replace the NAs

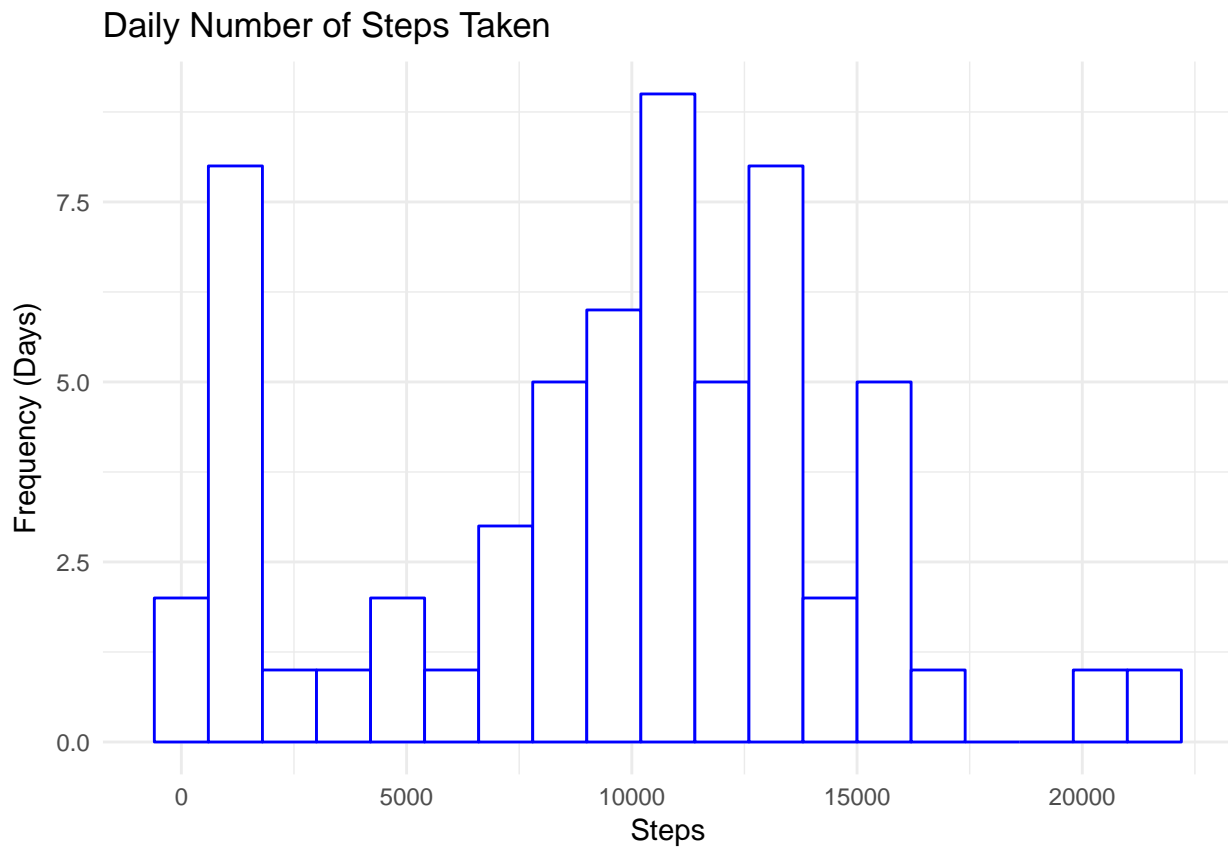
```
mynas <- is.na(data$steps)
avg_interval <- tapply(data$steps, data$interval, median, na.rm=TRUE, simplify=TRUE)
data$steps[mynas] <- avg_interval[as.character(data$interval[mynas])]
aggdata <- aggregate(steps~date, data, FUN=sum, na.rm=TRUE)
newmean <- mean(aggdata$steps)
newmean
```

```
## [1] 9503.869
```

```
newmedian <- median(aggdata$steps)
newmedian
```

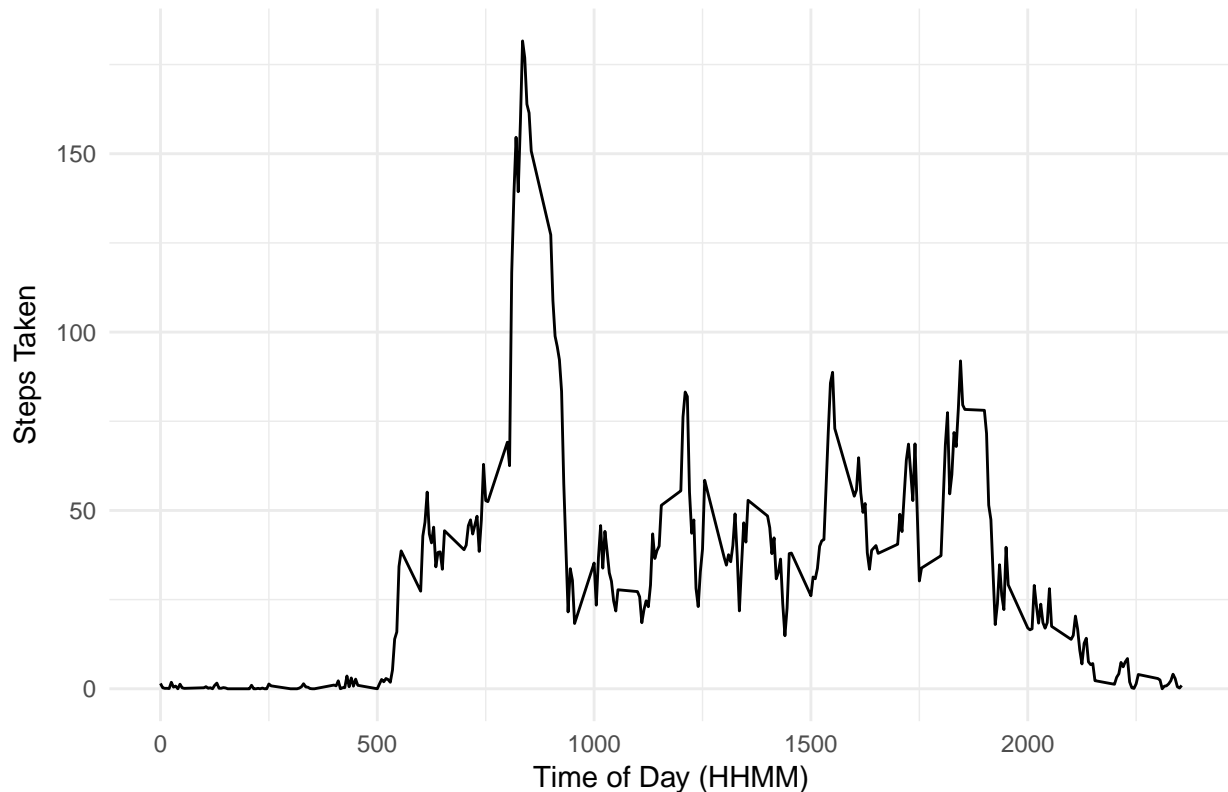
```
## [1] 10395
```

```
myplot <- ggplot(aggdata, aes(x=steps))
myplot <- myplot + geom_histogram(binwidth=1200, col='blue', fill='white')
myplot <- myplot + labs(title="Daily Number of Steps Taken", x="Steps", y="Frequency (Days)")
myplot <- myplot + theme_minimal()
myplot
```



```
aggdata <- aggregate(steps~interval, data, FUN=mean, na.rm=TRUE)
myplot<- ggplot(aggdata, aes(x=interval, y=steps)) +geom_line()
myplot <- myplot +theme_minimal()
myplot <- myplot + labs(title="Daily Number of Steps Taken", x="Time of Day (HHMM)", y="Steps Taken")
myplot
```

Daily Number of Steps Taken



```
mostactive <- aggdata$interval[which.max(aggdata$steps)]
mostactive
```

```
## [1] 835
```

Are there differences in activity patterns between weekdays and weekends?

```
data$day_of_week <- weekdays(as.Date(data$date))
weekend_days <- c("Saturday", "Sunday")
data$day_type <- ifelse(data$day_of_week %in% weekend_days, "Weekend", "Weekday")
weekend_data <- subset(data, data$day_type == "Weekend")
head(weekend_data)
```

```
##      steps      date interval day_of_week day_type
## 1441      0 2012-10-06         0   Saturday Weekend
## 1442      0 2012-10-06         5   Saturday Weekend
## 1443      0 2012-10-06        10   Saturday Weekend
## 1444      0 2012-10-06        15   Saturday Weekend
## 1445      0 2012-10-06        20   Saturday Weekend
## 1446      0 2012-10-06        25   Saturday Weekend
```

```
weekday_data <- subset(data, data$day_type == "Weekday")
we_agg <- aggregate(steps~date, weekend_data, FUN=sum, na.rm=TRUE)
```

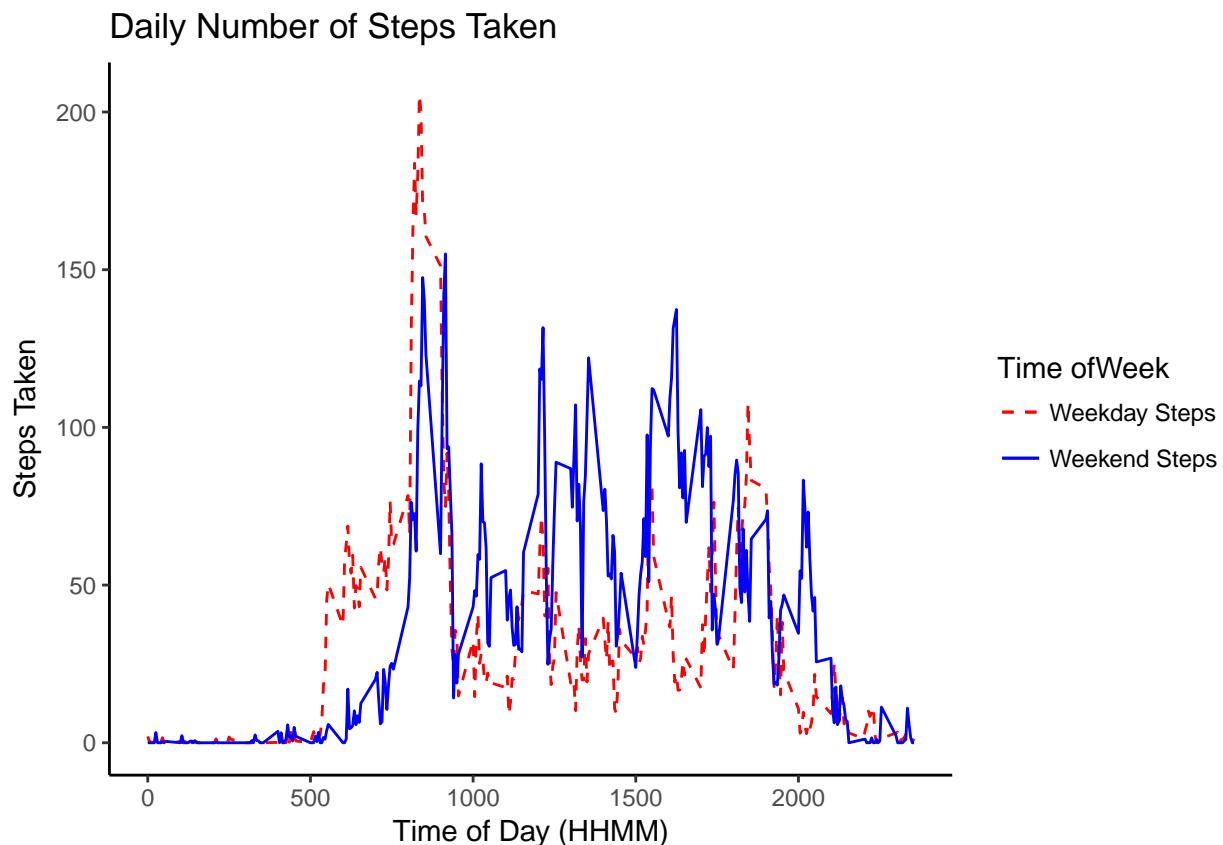
```
wd_agg <- aggregate(steps~date, weekday_data, FUN=sum, na.rm=TRUE)
mean(wd_agg$steps)
```

```
## [1] 10998.38
```

```
mean(wd_agg$steps)
```

```
## [1] 8972.489
```

```
we_agg <- aggregate(steps~interval, weekend_data, FUN=mean, na.rm=TRUE)
wd_agg <- aggregate(steps~interval, weekday_data, FUN=mean, na.rm=TRUE)
# Combine data frames
colnames(we_agg)[2] <- "Weekend_Steps"
colnames(wd_agg)[2] <- "Weekday_Steps"
plotframe <- merge(wd_agg, we_agg, by='interval')
#And plot the two time series on the same plot in order to compare
#Activity at different times of the week.
myplot<- ggplot(plotframe, aes(interval, y=value, color="Time of Week"))
myplot <- myplot +geom_line(aes(y=Weekday_Steps, col="Weekday Steps", linetype='dashed'))
myplot <- myplot +geom_line(aes(y = Weekend_Steps, col="Weekend Steps", linetype='solid'))
myplot <- myplot + scale_color_manual(values = c('red', 'blue'))
myplot <- myplot + scale_linetype_manual(values = c('dashed','solid'))
myplot <- myplot +theme_classic()
myplot <- myplot + guides(colour=guide_legend(override.aes=list(linetype=c(2,1))))
myplot <- myplot + labs(title="Daily Number of Steps Taken", x="Time of Day (HHMM)", y="Steps Taken", c
myplot
```



```
#or if you prefer a multi-panel plot
```

```
par(mfrow=c(1,2))
```

```
plot(plotframe$interval, plotframe$Weekend_Steps,lty=2, type='l', xlab="Time of Day", ylab="Weekend Steps")
```

```
plot(plotframe$interval, plotframe$Weekday_Steps, lty=1, type='l', xlab="Time of Day", ylab="Weekday Steps")
```

