

# Reproducible Research: Peer Assessment 1

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This report analyzes the number of steps taken by an anonymous individual user of a personal fitness armband device similar to the Nike “Fit” armband. The number of steps were measured over five minute intervals, 24 hours a day during the months of October and November 2012.

The 2012 steps per five minute interval data is for the months of October (30 days) and November (31 days) for a total of 61 days. In each hour there are 12 “five minute” intervals ( $60/5 = 12$  intervals per hour). Thus, there are 288 five minute intervals in a 24 hour day ( $288 = 24 \text{ hours} * 12 \text{ intervals per hour}$ ). Therefore, with 288 measurements per day for 61 days (24 hour days) one would **expect 17,568 observations** ( $17,568 = 61 \text{ days} * 288 \text{ per 24 hour day}$ ).

## Loading and preprocessing the data

The dataset is stored in a comma-separated-value (CSV) file in the main directory of a **GitHub** repository. So, we can load the data with an R “read.csv()” function. In this case name the **R** dataframe “**activity**”, the same name as the input filename. Finally, the initial struture of the **R** dataframe is shown with the **R** `str()` function:

```
#### Set directory to the local GitHub project of this assignment.
setwd("~/GitHub\\RepData_PeerAssessment1")

activity <- read.csv("activity.csv",
                     na.strings = "NA", stringsAsFactors = FALSE )

str(activity)
```

```
## 'data.frame':   17568 obs. of  3 variables:
## $ steps      : int   NA NA NA NA NA NA NA NA NA NA NA ...
## $ date       : chr   "2012-10-01" "2012-10-01" "2012-10-01" "2012-10-01" ...
## $ interval   : int    0  5 10 15 20 25 30 35 40 45 ...
```

As expected, for 61 days, the “**activity**” data frame has **17,568 observations**. The “**activity**” data frame has three variables: “**steps**”, “**date**” and “**interval**”. The initial values for the “**steps**” variable are missing and are marked as “NA”. The “**date**” variable is a character string we will want to convert to an **R** date type using the **R** “**as.Date()**” function. The “**interval**” variable is an integer that initially appears to be incremented by 5 for each observation, this first impression will be modified on closer observation.

So, let’s convert “**date**” to an R date type and take a closer look at “**steps**” and “**interval**”. We want to know if all of the “**steps**” are missing? or if not all are missing, how many are missing? and what percentage of the dataset is that? For the “**interval**” variable, we want to know how often the pattern restarts at zero (or does it increase all the way through the data set?). We will print 289 observations of “**interval**”, one more than the 288 observations per day.

```
activity$date <- as.Date(activity$date) # coerce "date" to date data type
sum(is.na(activity$steps))             # How many missing values?
```

```
## [1] 2304
```

```
mean(is.na(activity$steps)) # What percent are missing values?
```

```
## [1] 0.1311475
```

```
head(activity$interval, 289) # What does a full daily cycle look like?
```

```
## [1] 0 5 10 15 20 25 30 35 40 45 50 55 100 105
## [15] 110 115 120 125 130 135 140 145 150 155 200 205 210 215
## [29] 220 225 230 235 240 245 250 255 300 305 310 315 320 325
## [43] 330 335 340 345 350 355 400 405 410 415 420 425 430 435
## [57] 440 445 450 455 500 505 510 515 520 525 530 535 540 545
## [71] 550 555 600 605 610 615 620 625 630 635 640 645 650 655
## [85] 700 705 710 715 720 725 730 735 740 745 750 755 800 805
## [99] 810 815 820 825 830 835 840 845 850 855 900 905 910 915
## [113] 920 925 930 935 940 945 950 955 1000 1005 1010 1015 1020 1025
## [127] 1030 1035 1040 1045 1050 1055 1100 1105 1110 1115 1120 1125 1130 1135
## [141] 1140 1145 1150 1155 1200 1205 1210 1215 1220 1225 1230 1235 1240 1245
## [155] 1250 1255 1300 1305 1310 1315 1320 1325 1330 1335 1340 1345 1350 1355
## [169] 1400 1405 1410 1415 1420 1425 1430 1435 1440 1445 1450 1455 1500 1505
## [183] 1510 1515 1520 1525 1530 1535 1540 1545 1550 1555 1600 1605 1610 1615
## [197] 1620 1625 1630 1635 1640 1645 1650 1655 1700 1705 1710 1715 1720 1725
## [211] 1730 1735 1740 1745 1750 1755 1800 1805 1810 1815 1820 1825 1830 1835
## [225] 1840 1845 1850 1855 1900 1905 1910 1915 1920 1925 1930 1935 1940 1945
## [239] 1950 1955 2000 2005 2010 2015 2020 2025 2030 2035 2040 2045 2050 2055
## [253] 2100 2105 2110 2115 2120 2125 2130 2135 2140 2145 2150 2155 2200 2205
## [267] 2210 2215 2220 2225 2230 2235 2240 2245 2250 2255 2300 2305 2310 2315
## [281] 2320 2325 2330 2335 2340 2345 2350 2355 0
```

Over 2,300 observations of the “**steps**” variable are missing, while this is a lot; it is still only 13.1% of the 17,568 observations. So, for our initial analysis, we can simply ignore the missing values, by removing them and only work with complete cases. Later, we can try to guess (technically, “impute”) the missing values in a process called “imputation” and see whether that changes the analysis. Not shown, but there are no missing values for “**date**” and “**interval**”.

The “**interval**” variable does not make sense as an integer.

Although, the “**interval**” variable does reset to zero at observation 289, as expected (recall there are 288 observations per 24 hour day); the 288th observation is 2,355 rather than the 1,435 one would expect if one multiplied the intervals 0 (zero) through 287 by 5 ( $1,435 = 287 \times 5$ ). That is a big gap between 2,355 and 1,435 so something different is going on.

If we examine the first dozen observations we see the “**interval**” variable jumps from 55 to 100. It is 100 when it should be 60. But, wait, if the “1” in “100” represents “one hour” and the “23” in “2355” represents “23 hours” then it is clear that the “**interval**” variable is actually hours and minutes with the leading zeros removed. That is, “100” should be understood as “01:00” and “2355” should be understood as “23:55” and so on.

We can fix the integer representation by using the R “**sprintf()**” function to restore the leading zeros to the time and store the result in a variable named “**HHMM**” which in turn, can be combined with the date to build a POSIX standard date time string in a variable named, “**datetime**”.

```
# Convert the interval to HHMM by formatting with leading zero
activity$HHMM <- sprintf("%04d", as.integer(activity$interval))
# Now we can combine date and time as a string
```

```

# and format the resulting string as a POSIX datetime string
datetimestring    <- paste(activity$date, activity$HHMM)
activity$datetime <- strptime(datetimestring,
                             "%Y-%m-%d %H%M", tz = "")

# Use the weekday() function to interrogate the POSIX datetime string
# to obtain a day of week abbreviation ("Sun", "Mon", "Tue", etc) for each date
activity$dayofweek <- weekdays(activity$datetime, abbreviate=TRUE)

# Create a weekday/weekend factor by first assigning all data to "weekday"
# and then check for "Sat" or "Sun" and reassign to "weekend".
activity$daytype = "weekday"
activity$daytype[activity$dayofweek == "Sat" | activity$dayofweek == "Sun"] <- "weekend"

# Convert to R "factor" data type
activity$dayofweek = factor(activity$dayofweek)
activity$daytype    = factor(activity$daytype)

```

With these changes, the “**activity**” data set is ready for the first stage of our analysis where we simply ignore (remove) the missing values. But, for cosmetic reasons, we might want to reorder the “**activity**” data frame variables in a more logical order:

```

activity <- subset( activity, select = c(datetime, date, dayofweek, daytype, HHMM, interval, steps))
str(activity)

```

```

## 'data.frame':   17568 obs. of  7 variables:
## $ datetime : POSIXlt, format: "2012-10-01 00:00:00" "2012-10-01 00:05:00" ...
## $ date      : Date, format: "2012-10-01" "2012-10-01" ...
## $ dayofweek: Factor w/ 7 levels "Fri","Mon","Sat",...: 2 2 2 2 2 2 2 2 2 ...
## $ daytype   : Factor w/ 2 levels "weekday","weekend": 1 1 1 1 1 1 1 1 1 ...
## $ HHMM      : chr  "0000" "0005" "0010" "0015" ...
## $ interval  : int   0 5 10 15 20 25 30 35 40 45 ...
## $ steps     : int  NA NA NA NA NA NA NA NA NA NA ...

```

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

If we sum the steps for all of the intervals for a given date, we have a daily total. This use of the **R** `aggregate()` function follows an example in Jared Lander’s book “*R for Everyone*” page 121 where he uses the “**diamonds**” data frame which comes with the **ggplot2** package.

Once we have a daily total for each of the days; we can calculate a value for the average (mean) and median by removing the NAs, that would otherwise cause an NA result. We round the steps to zero because users expect a whole number of steps and we already have five significant digits of precision.

```

PerDay <- aggregate(steps ~ date, data=activity, sum)
meanstepsperday <- round(mean(PerDay$steps, na.rm = TRUE), digits = 0)
meanstepsperday

```

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

```
medianstepsperday <- median(PerDay$steps, na.rm = TRUE)
medianstepsperday
```

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

The **average (mean) number of steps per day** is **10,766** ; and the **median number of steps per day** is very close at **10,765**.

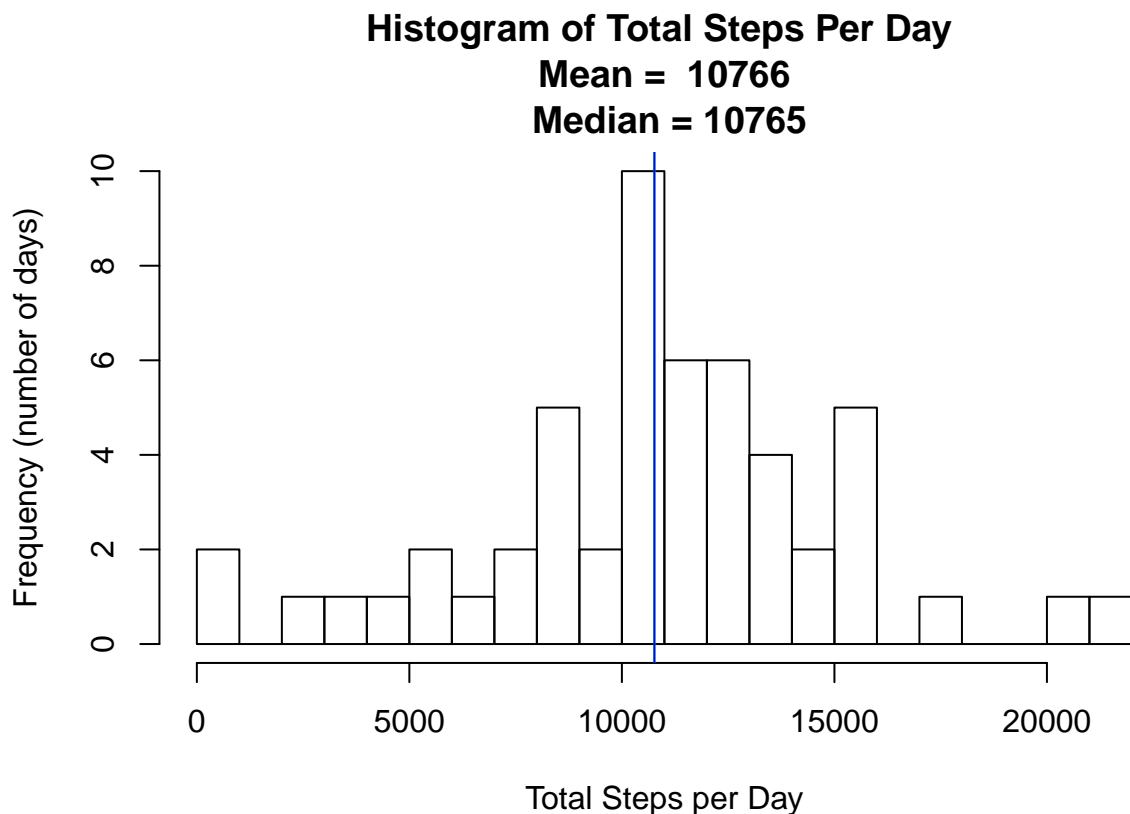
### What is the average daily activity pattern?

These 10,000+ values seem plausible for an active person with a 10,000 steps a day goal, which seems to be popular goal:

"The origins of the 10,000-steps recommendation aren't exactly scientific. Pedometers sold in Japan in the 1960s were marketed under the name "manpo-kei," which translates to "10,000 steps meter" ... studies conducted since then suggest that people who increased their walking to 10,000 steps daily experience health benefits."

Rachael Rettner, "*The Truth About '10,000 Steps' a Day*" LiveScience.org, March 2014  
retrieved from **LiveScience.org** during September 2015

<http://www.livescience.com/43956-walking-10000-steps-healthy.html>



While a central value (mean or median) near 10,000 seems plausible for an active person with a 10,000 steps per day goal; the extremes of near zero steps per day and a maximum over 20,000 steps per day may require further inquiry. For example, did the person spend a sick day in bed (with near zero steps)? and did the

person participate in a 10,000 step walk in addition to their normal 10,000 steps (resulting in over 20,000 steps per day)?

```
# Calculate steps per five minute interval
# (in 24 hour cycle)
# PerIntervalSum      <- aggregate(steps ~ factor(HHMM), activity, sum)
PerIntervalMean      <- aggregate(steps ~ factor(HHMM), activity, mean)
PerIntervalMedian    <- aggregate(steps ~ factor(HHMM), activity, median)
ColumnNames <- c("HHMM", "steps")
# colnames(PerIntervalSum)      <- ColumnNames
colnames(PerIntervalMean)      <- ColumnNames
colnames(PerIntervalMedian)    <- ColumnNames

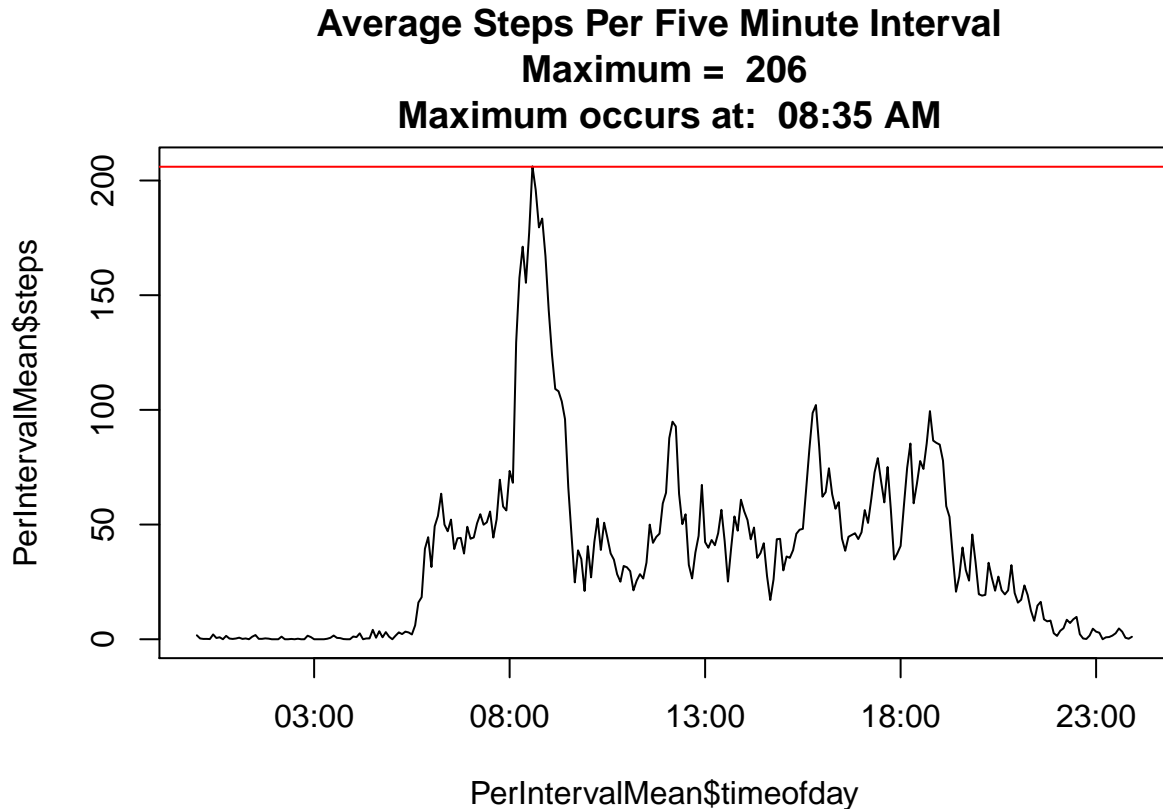
PerIntervalMean$timeofday      <- strptime(PerIntervalMean$HHMM, "%H%M", tz = "")
PerIntervalMedian$timeofday    <- strptime(PerIntervalMedian$HHMM, "%H%M", tz = "")

Max5MinuteSteps <- PerIntervalMean[PerIntervalMean$steps == max(PerIntervalMean$steps), ]
# ColumnNames <- c("timeofday", "steps")
# colnames(Max5MinuteSteps) <- ColumnNames
Max5MinuteSteps
```

```
##      HHMM      steps      timeofday
## 104 0835 206.1698 2015-09-19 08:35:00
```

```
plot(PerIntervalMean$timeofday, PerIntervalMean$steps, type = "l",
     main = paste("Average Steps Per Five Minute Interval",
                  "\n Maximum = " , round(max(PerIntervalMean$steps), digits=0),
                  "\n Maximum occurs at: ", format(Max5MinuteSteps$timeofday, "%H:%M AM")
     )

abline(h = round(max(PerIntervalMean$steps), digits=0), col = "red")
```



```
# abline(v = Max5MinuteSteps$timeofday, col = "red")
```

### Imputing missing values

As noted earlier, over 2,300 observations of the “steps” variable are missing, while this is a lot; it is still only 13.1% of the 17,568 observations.

```
sum(is.na(activity$steps))      # How many missing values?
```

```
## [1] 2304
```

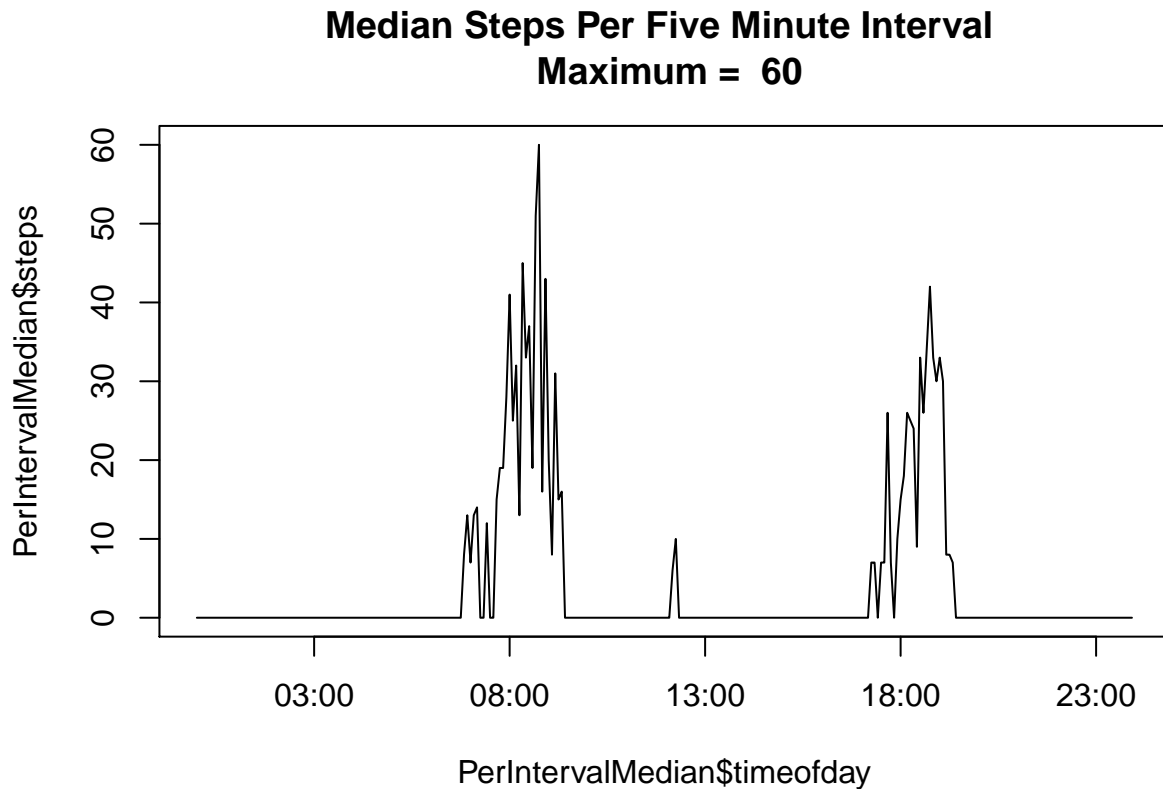
```
mean(is.na(activity$steps))    # What percent missing values?
```

```
## [1] 0.1311475
```

There seems to be a regular pattern for time of day, but we still have to decided whether we should impute with 5 minute interval **averages** or **medians**.

We have already computed a graph of the **five minute means (averages)**, for comparison here is a graph of **median steps per five minutes**, it peaks at 60 steps and often takes on a zero value.

```
plot(PerIntervalMedian$timeofday, PerIntervalMedian$steps, type = "l",
     main = paste("Median Steps Per Five Minute Interval",
                  "\n Maximum = " , round(max(PerIntervalMedian$steps), digits=0)
     )
)
```



So, let's use the “**steps**” variable in the “**PerIntervalMedian**” data frame to create a variable we can use to **impute (fill-in)** the missing values of the “**steps**” variable in the “**activity**” data frame. We need to expand the “**PerIntervalMedian**” version of “**steps**” from just one day to all **61 days** (17,568 observations). We can do this in **R** by repeating the variable **61 times** using the **R rep()** function:

```
# Expand PerIntervalMedian from one day to 61 days by repeating the daily values
activity$fill <- rep(PerIntervalMedian$steps, 61)
str(activity$fill)
```

```
## int [1:17568] 0 0 0 0 0 0 0 0 0 0 ...
```

We now have a prediction variable “**fill**” that we can use to “fill-in” (replace) the missing values (NA) of the “**steps**” variable.

Now we can define two variables **maskNA** and **maskValue**. These zero-one variables are like dummy variables in econometrics or bit-masks in computer science.

```
maskNA    <- is.na(activity$steps)  # 1 where steps is missing; 0 everywhere else
maskValue <- ~is.na(activity$steps) # 1 where steps has a value; 0 everywhere else
```

In order to fill in the missing values; we want to multiply “steps” by “maskValue” and multiply “fill” by “maskNA” and then add the resulting two variables together. What happens is “maskValue” preserves the values in “steps” while zeroing out the NAs; while “fill” times maskNA preserves the values where there are NAs and converts all of the other values to zero.

Unfortunately, just multiplying “steps” by maskValue won’t replace the NAs with zeros because NA times any value is still NA; so we have to use another method to replace the NAs in “steps” with zeros.

```
# What we would have liked to have done:
# activity$stepsNoMissing <- (activity$steps*maskValue) + (activity$fill*maskNA)
#
#Instead
# Stackoverflow
# http://stackoverflow.com/questions/10139284/set-na-to-0-in-r
stepsNA2Zero <- activity$steps
stepsNA2Zero[is.na(stepsNA2Zero)] <- 0

# Now we can add the values
activity$stepsNoNA <- stepsNA2Zero + (activity$fill*maskNA)
sum(is.na(activity$stepsNoNA))      # How many missing values in "NoNA" version?
```

```
## [1] 0
```

```
mean(is.na(activity$stepsNoNA))      # What percent missing values in "NoNA" version?
```

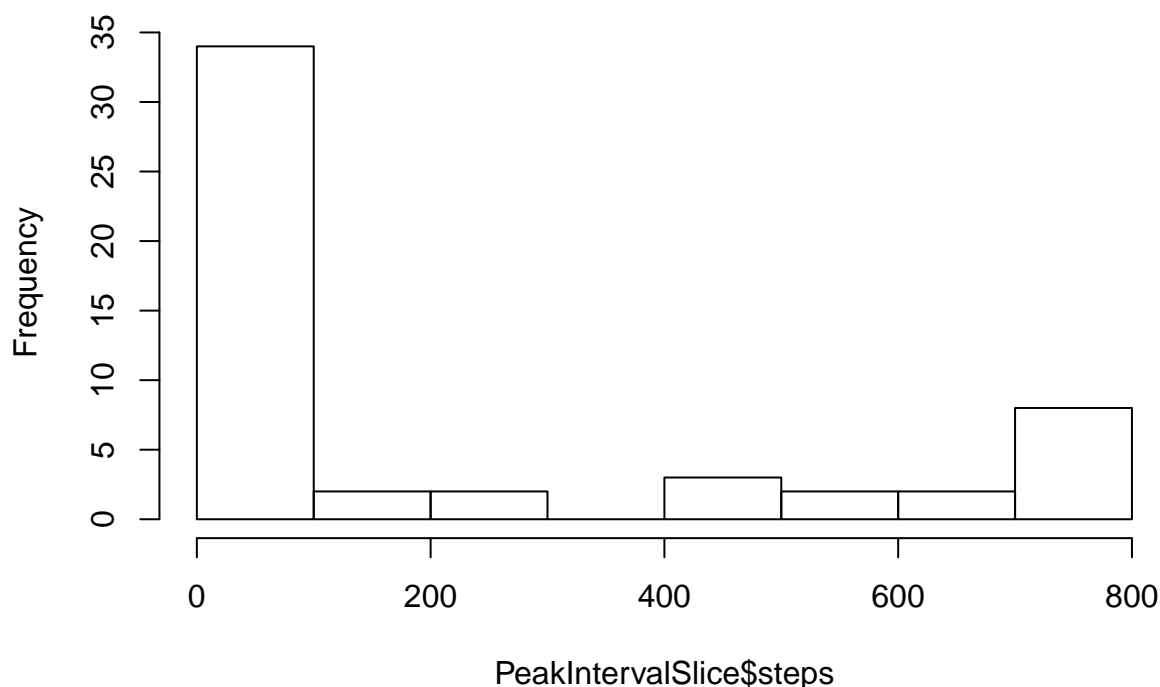
```
## [1] 0
```

Is the average being skewed by one extreme value at 8:35 AM?

```
PeakIntervalSlice <- activity[activity$HHMM == "0835", ]
# PeakIntervalSlice # display of data omitted for space reasons.
hist(PeakIntervalSlice$steps,
     main = ("Histogram of raw steps @ 8:35 AM"))
)
```



## Histogram of raw steps @ 8:35 AM



The histogram shows the problem; the distribution of raw (not summarized) steps at 8:35 AM is bi-modal with the peaks at extreme values of zero and 700-800 steps per hour; neither a mean, nor a median is likely to summarize this well!

Visual inspection of the data shows that zeros are not limited to weekends and the 700+ values occur during weekdays and are not the result of weekend special events or treadmill time.

My guess is that the five minute intervals are too narrow to capture the commute of this individual. The peak walking of the morning commute may occur during this window or may occur a little later or earlier depending on whether the commuter is running early or late. The five minute interval being too narrow is analogous to the bin on a histogram being too narrow – the histogram is noisy and the overall shape is lost.

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

```
DayTypePerIntervalMean <- aggregate(steps ~ factor(HHMM)+daytype, activity, mean)
DayTypePerIntervalMean
```

```
##      factor(HHMM) daytype      steps
## 1           0000 weekday  2.3333333
## 2           0005 weekday  0.4615385
## 3           0010 weekday  0.1794872
## 4           0015 weekday  0.2051282
## 5           0020 weekday  0.1025641
## 6           0025 weekday  1.5128205
## 7           0030 weekday  0.7179487
```

## 8	0035 weekday	1.1794872
## 9	0040 weekday	0.0000000
## 10	0045 weekday	1.8461538
## 11	0050 weekday	0.4102564
## 12	0055 weekday	0.0000000
## 13	0100 weekday	0.4358974
## 14	0105 weekday	0.0000000
## 15	0110 weekday	0.2051282
## 16	0115 weekday	0.4615385
## 17	0120 weekday	0.0000000
## 18	0125 weekday	1.5128205
## 19	0130 weekday	2.2820513
## 20	0135 weekday	0.0000000
## 21	0140 weekday	0.2307692
## 22	0145 weekday	0.2307692
## 23	0150 weekday	0.3589744
## 24	0155 weekday	0.0000000
## 25	0200 weekday	0.0000000
## 26	0205 weekday	0.0000000
## 27	0210 weekday	1.4358974
## 28	0215 weekday	0.0000000
## 29	0220 weekday	0.0000000
## 30	0225 weekday	0.1794872
## 31	0230 weekday	0.0000000
## 32	0235 weekday	0.3076923
## 33	0240 weekday	0.0000000
## 34	0245 weekday	0.0000000
## 35	0250 weekday	2.1025641
## 36	0255 weekday	1.2820513
## 37	0300 weekday	0.0000000
## 38	0305 weekday	0.0000000
## 39	0310 weekday	0.0000000
## 40	0315 weekday	0.0000000
## 41	0320 weekday	0.0000000
## 42	0325 weekday	0.8461538
## 43	0330 weekday	1.1794872
## 44	0335 weekday	0.5128205
## 45	0340 weekday	0.4102564
## 46	0345 weekday	0.1025641
## 47	0350 weekday	0.0000000
## 48	0355 weekday	0.0000000
## 49	0400 weekday	0.1282051
## 50	0405 weekday	1.2820513
## 51	0410 weekday	2.1794872
## 52	0415 weekday	0.0000000
## 53	0420 weekday	0.4615385
## 54	0425 weekday	0.0000000
## 55	0430 weekday	3.2564103
## 56	0435 weekday	0.1538462
## 57	0440 weekday	3.8205128
## 58	0445 weekday	0.8974359
## 59	0450 weekday	2.2307692
## 60	0455 weekday	0.6666667
## 61	0500 weekday	0.0000000

## 62	0505 weekday	2.1282051
## 63	0510 weekday	4.0769231
## 64	0515 weekday	2.1794872
## 65	0520 weekday	4.3589744
## 66	0525 weekday	2.6666667
## 67	0530 weekday	2.8461538
## 68	0535 weekday	8.2307692
## 69	0540 weekday	21.0769231
## 70	0545 weekday	24.4615385
## 71	0550 weekday	52.0256410
## 72	0555 weekday	58.0769231
## 73	0600 weekday	42.7948718
## 74	0605 weekday	66.9487179
## 75	0610 weekday	72.5897436
## 76	0615 weekday	79.2564103
## 77	0620 weekday	66.0769231
## 78	0625 weekday	62.0256410
## 79	0630 weekday	68.6410256
## 80	0635 weekday	49.3076923
## 81	0640 weekday	57.4615385
## 82	0645 weekday	56.5128205
## 83	0650 weekday	48.5384615
## 84	0655 weekday	62.1794872
## 85	0700 weekday	51.6410256
## 86	0705 weekday	51.8205128
## 87	0710 weekday	63.7948718
## 88	0715 weekday	71.6153846
## 89	0720 weekday	65.1282051
## 90	0725 weekday	60.3589744
## 91	0730 weekday	67.8461538
## 92	0735 weekday	55.8974359
## 93	0740 weekday	64.3333333
## 94	0745 weekday	85.5128205
## 95	0750 weekday	69.2564103
## 96	0755 weekday	68.1794872
## 97	0800 weekday	84.1538462
## 98	0805 weekday	72.5384615
## 99	0810 weekday	146.2564103
## 100	0815 weekday	185.7435897
## 101	0820 weekday	205.1025641
## 102	0825 weekday	187.9487179
## 103	0830 weekday	202.2051282
## 104	0835 weekday	234.1025641
## 105	0840 weekday	222.4358974
## 106	0845 weekday	186.5897436
## 107	0850 weekday	192.4358974
## 108	0855 weekday	178.6410256
## 109	0900 weekday	171.3846154
## 110	0905 weekday	126.0512821
## 111	0910 weekday	91.6153846
## 112	0915 weekday	84.1025641
## 113	0920 weekday	103.5128205
## 114	0925 weekday	91.9230769
## 115	0930 weekday	57.3333333

## 116	0935 weekday	34.4102564
## 117	0940 weekday	27.8717949
## 118	0945 weekday	41.1794872
## 119	0950 weekday	39.7692308
## 120	0955 weekday	17.1025641
## 121	1000 weekday	37.4615385
## 122	1005 weekday	16.8717949
## 123	1010 weekday	38.5641026
## 124	1015 weekday	47.0769231
## 125	1020 weekday	29.0256410
## 126	1025 weekday	32.7435897
## 127	1030 weekday	31.4102564
## 128	1035 weekday	22.2307692
## 129	1040 weekday	21.7948718
## 130	1045 weekday	25.5384615
## 131	1050 weekday	21.5641026
## 132	1055 weekday	21.9230769
## 133	1100 weekday	20.2051282
## 134	1105 weekday	24.3846154
## 135	1110 weekday	10.2051282
## 136	1115 weekday	14.8461538
## 137	1120 weekday	23.5384615
## 138	1125 weekday	23.3076923
## 139	1130 weekday	32.6666667
## 140	1135 weekday	50.2307692
## 141	1140 weekday	44.9487179
## 142	1145 weekday	48.4358974
## 143	1150 weekday	50.7435897
## 144	1155 weekday	55.6666667
## 145	1200 weekday	54.4615385
## 146	1205 weekday	70.5641026
## 147	1210 weekday	81.9230769
## 148	1215 weekday	72.5897436
## 149	1220 weekday	46.4615385
## 150	1225 weekday	46.3076923
## 151	1230 weekday	63.8205128
## 152	1235 weekday	30.4871795
## 153	1240 weekday	21.2820513
## 154	1245 weekday	28.0256410
## 155	1250 weekday	30.8974359
## 156	1255 weekday	54.9487179
## 157	1300 weekday	21.8717949
## 158	1305 weekday	23.5641026
## 159	1310 weekday	21.6923077
## 160	1315 weekday	11.7435897
## 161	1320 weekday	34.0000000
## 162	1325 weekday	43.0769231
## 163	1330 weekday	30.0769231
## 164	1335 weekday	23.0256410
## 165	1340 weekday	22.9743590
## 166	1345 weekday	38.1282051
## 167	1350 weekday	22.2307692
## 168	1355 weekday	32.5641026
## 169	1400 weekday	45.5641026

## 170	1405 weekday	37.6410256
## 171	1410 weekday	30.3589744
## 172	1415 weekday	44.4871795
## 173	1420 weekday	26.2564103
## 174	1425 weekday	29.7179487
## 175	1430 weekday	29.8974359
## 176	1435 weekday	12.5128205
## 177	1440 weekday	10.6923077
## 178	1445 weekday	21.3589744
## 179	1450 weekday	41.5897436
## 180	1455 weekday	37.4358974
## 181	1500 weekday	31.0000000
## 182	1505 weekday	34.8974359
## 183	1510 weekday	29.1025641
## 184	1515 weekday	30.8461538
## 185	1520 weekday	38.9230769
## 186	1525 weekday	35.7435897
## 187	1530 weekday	41.2051282
## 188	1535 weekday	48.7179487
## 189	1540 weekday	91.7435897
## 190	1545 weekday	95.4358974
## 191	1550 weekday	92.6923077
## 192	1555 weekday	68.2051282
## 193	1600 weekday	44.5384615
## 194	1605 weekday	42.2820513
## 195	1610 weekday	53.8461538
## 196	1615 weekday	31.9743590
## 197	1620 weekday	22.1794872
## 198	1625 weekday	24.8717949
## 199	1630 weekday	19.2307692
## 200	1635 weekday	19.2564103
## 201	1640 weekday	22.9743590
## 202	1645 weekday	29.9230769
## 203	1650 weekday	24.7692308
## 204	1655 weekday	30.6923077
## 205	1700 weekday	20.0256410
## 206	1705 weekday	43.2051282
## 207	1710 weekday	31.6410256
## 208	1715 weekday	46.0512821
## 209	1720 weekday	58.1794872
## 210	1725 weekday	71.3589744
## 211	1730 weekday	54.1794872
## 212	1735 weekday	66.7692308
## 213	1740 weekday	84.0769231
## 214	1745 weekday	59.7692308
## 215	1750 weekday	34.4615385
## 216	1755 weekday	37.6153846
## 217	1800 weekday	24.4871795
## 218	1805 weekday	44.8717949
## 219	1810 weekday	66.0769231
## 220	1815 weekday	82.2307692
## 221	1820 weekday	61.7179487
## 222	1825 weekday	74.3333333
## 223	1830 weekday	79.4615385

## 224	1835	weekday	82.6153846
## 225	1840	weekday	92.6923077
## 226	1845	weekday	117.9230769
## 227	1850	weekday	103.5641026
## 228	1855	weekday	91.3589744
## 229	1900	weekday	87.9743590
## 230	1905	weekday	77.1282051
## 231	1910	weekday	63.0512821
## 232	1915	weekday	54.5384615
## 233	1920	weekday	38.1282051
## 234	1925	weekday	20.5384615
## 235	1930	weekday	29.3589744
## 236	1935	weekday	46.8974359
## 237	1940	weekday	30.0256410
## 238	1945	weekday	17.5128205
## 239	1950	weekday	44.0512821
## 240	1955	weekday	26.3333333
## 241	2000	weekday	12.4358974
## 242	2005	weekday	3.4871795
## 243	2010	weekday	4.8974359
## 244	2015	weekday	11.1538462
## 245	2020	weekday	5.9230769
## 246	2025	weekday	3.3333333
## 247	2030	weekday	7.0769231
## 248	2035	weekday	4.9743590
## 249	2040	weekday	7.3333333
## 250	2045	weekday	11.8461538
## 251	2050	weekday	25.0000000
## 252	2055	weekday	16.8717949
## 253	2100	weekday	10.6666667
## 254	2105	weekday	19.1538462
## 255	2110	weekday	29.2820513
## 256	2115	weekday	18.8974359
## 257	2120	weekday	14.5641026
## 258	2125	weekday	8.0512821
## 259	2130	weekday	12.5128205
## 260	2135	weekday	16.5384615
## 261	2140	weekday	6.8974359
## 262	2145	weekday	7.5641026
## 263	2150	weekday	8.2820513
## 264	2155	weekday	3.5641026
## 265	2200	weekday	1.5384615
## 266	2205	weekday	4.5384615
## 267	2210	weekday	6.5384615
## 268	2215	weekday	11.5641026
## 269	2220	weekday	9.6153846
## 270	2225	weekday	11.1794872
## 271	2230	weekday	13.2564103
## 272	2235	weekday	3.0000000
## 273	2240	weekday	0.0000000
## 274	2245	weekday	0.1538462
## 275	2250	weekday	1.9487179
## 276	2255	weekday	1.6153846
## 277	2300	weekday	3.5897436

## 278	2305 weekday	3.8717949
## 279	2310 weekday	0.0000000
## 280	2315 weekday	1.1282051
## 281	2320 weekday	1.3076923
## 282	2325 weekday	1.9230769
## 283	2330 weekday	3.1025641
## 284	2335 weekday	1.8717949
## 285	2340 weekday	2.0769231
## 286	2345 weekday	0.2051282
## 287	2350 weekday	0.3076923
## 288	2355 weekday	1.4615385
## 289	0000 weekend	0.0000000
## 290	0005 weekend	0.0000000
## 291	0010 weekend	0.0000000
## 292	0015 weekend	0.0000000
## 293	0020 weekend	0.0000000
## 294	0025 weekend	3.7142857
## 295	0030 weekend	0.0000000
## 296	0035 weekend	0.0000000
## 297	0040 weekend	0.0000000
## 298	0045 weekend	0.4285714
## 299	0050 weekend	0.0000000
## 300	0055 weekend	0.5000000
## 301	0100 weekend	0.0000000
## 302	0105 weekend	2.5714286
## 303	0110 weekend	0.0000000
## 304	0115 weekend	0.0000000
## 305	0120 weekend	0.0000000
## 306	0125 weekend	0.0000000
## 307	0130 weekend	0.5714286
## 308	0135 weekend	0.6428571
## 309	0140 weekend	0.0000000
## 310	0145 weekend	0.7857143
## 311	0150 weekend	0.0000000
## 312	0155 weekend	0.0000000
## 313	0200 weekend	0.0000000
## 314	0205 weekend	0.0000000
## 315	0210 weekend	0.2857143
## 316	0215 weekend	0.0000000
## 317	0220 weekend	0.0000000
## 318	0225 weekend	0.0000000
## 319	0230 weekend	0.0000000
## 320	0235 weekend	0.0000000
## 321	0240 weekend	0.0000000
## 322	0245 weekend	0.0000000
## 323	0250 weekend	0.0000000
## 324	0255 weekend	0.0000000
## 325	0300 weekend	0.0000000
## 326	0305 weekend	0.0000000
## 327	0310 weekend	0.0000000
## 328	0315 weekend	0.0000000
## 329	0320 weekend	0.7857143
## 330	0325 weekend	0.0000000
## 331	0330 weekend	2.8571429

## 332	0335 weekend	0.7857143
## 333	0340 weekend	0.7142857
## 334	0345 weekend	0.0000000
## 335	0350 weekend	0.0000000
## 336	0355 weekend	0.0000000
## 337	0400 weekend	4.1428571
## 338	0405 weekend	0.0000000
## 339	0410 weekend	3.6428571
## 340	0415 weekend	0.0000000
## 341	0420 weekend	0.0000000
## 342	0425 weekend	1.3571429
## 343	0430 weekend	6.5000000
## 344	0435 weekend	2.0714286
## 345	0440 weekend	2.5714286
## 346	0445 weekend	0.6428571
## 347	0450 weekend	5.5714286
## 348	0455 weekend	2.3571429
## 349	0500 weekend	0.0000000
## 350	0505 weekend	0.0000000
## 351	0510 weekend	0.0000000
## 352	0515 weekend	2.4285714
## 353	0520 weekend	0.4285714
## 354	0525 weekend	3.7857143
## 355	0530 weekend	0.0000000
## 356	0535 weekend	0.0000000
## 357	0540 weekend	1.9285714
## 358	0545 weekend	1.2857143
## 359	0550 weekend	4.4285714
## 360	0555 weekend	6.6428571
## 361	0600 weekend	0.0000000
## 362	0605 weekend	0.0000000
## 363	0610 weekend	1.3571429
## 364	0615 weekend	19.4285714
## 365	0620 weekend	5.0714286
## 366	0625 weekend	5.4285714
## 367	0630 weekend	6.2142857
## 368	0635 weekend	11.5714286
## 369	0640 weekend	6.5714286
## 370	0645 weekend	9.7857143
## 371	0650 weekend	6.2142857
## 372	0655 weekend	12.4285714
## 373	0700 weekend	22.0000000
## 374	0705 weekend	23.6428571
## 375	0710 weekend	13.5000000
## 376	0715 weekend	6.8571429
## 377	0720 weekend	7.5714286
## 378	0725 weekend	24.8571429
## 379	0730 weekend	21.7857143
## 380	0735 weekend	12.0714286
## 381	0740 weekend	18.6428571
## 382	0745 weekend	25.0714286
## 383	0750 weekend	26.0714286
## 384	0755 weekend	22.6428571
## 385	0800 weekend	43.3571429



## 386	0805 weekend	56.1428571
## 387	0810 weekend	82.5714286
## 388	0815 weekend	78.9285714
## 389	0820 weekend	76.5714286
## 390	0825 weekend	64.7142857
## 391	0830 weekend	107.9285714
## 392	0835 weekend	128.3571429
## 393	0840 weekend	122.0714286
## 394	0845 weekend	160.0000000
## 395	0850 weekend	158.2142857
## 396	0855 weekend	134.6428571
## 397	0900 weekend	65.6428571
## 398	0905 weekend	118.4285714
## 399	0910 weekend	157.8571429
## 400	0915 weekend	175.0000000
## 401	0920 weekend	104.2857143
## 402	0925 weekend	107.2142857
## 403	0930 weekend	90.9285714
## 404	0935 weekend	75.3571429
## 405	0940 weekend	16.2142857
## 406	0945 weekend	32.0000000
## 407	0950 weekend	21.6428571
## 408	0955 weekend	32.0714286
## 409	1000 weekend	49.2142857
## 410	1005 weekend	55.1428571
## 411	1010 weekend	53.1428571
## 412	1015 weekend	68.2142857
## 413	1020 weekend	66.5000000
## 414	1025 weekend	101.0714286
## 415	1030 weekend	80.1428571
## 416	1035 weekend	79.7142857
## 417	1040 weekend	70.6428571
## 418	1045 weekend	36.1428571
## 419	1050 weekend	34.9285714
## 420	1055 weekend	59.8571429
## 421	1100 weekend	62.4285714
## 422	1105 weekend	44.4285714
## 423	1110 weekend	52.2857143
## 424	1115 weekend	55.3571429
## 425	1120 weekend	41.8571429
## 426	1125 weekend	35.2857143
## 427	1130 weekend	35.5714286
## 428	1135 weekend	49.2857143
## 429	1140 weekend	33.9285714
## 430	1145 weekend	33.9285714
## 431	1150 weekend	32.9285714
## 432	1155 weekend	69.0000000
## 433	1200 weekend	90.0714286
## 434	1205 weekend	135.4285714
## 435	1210 weekend	130.8571429
## 436	1215 weekend	149.0000000
## 437	1220 weekend	110.5714286
## 438	1225 weekend	60.9285714
## 439	1230 weekend	28.4285714

## 440	1235 weekend	37.7857143
## 441	1240 weekend	41.1428571
## 442	1245 weekend	64.7857143
## 443	1250 weekend	84.5000000
## 444	1255 weekend	101.6428571
## 445	1300 weekend	99.3571429
## 446	1305 weekend	85.3571429
## 447	1310 weekend	103.3571429
## 448	1315 weekend	122.4285714
## 449	1320 weekend	80.3571429
## 450	1325 weekend	93.6428571
## 451	1330 weekend	78.0714286
## 452	1335 weekend	31.0000000
## 453	1340 weekend	87.2857143
## 454	1345 weekend	96.5000000
## 455	1350 weekend	117.2142857
## 456	1355 weekend	139.5000000
## 457	1400 weekend	84.1428571
## 458	1405 weekend	91.8571429
## 459	1410 weekend	80.4285714
## 460	1415 weekend	60.4285714
## 461	1420 weekend	61.1428571
## 462	1425 weekend	59.3571429
## 463	1430 weekend	75.1428571
## 464	1435 weekend	69.2857143
## 465	1440 weekend	35.0000000
## 466	1445 weekend	39.2142857
## 467	1450 weekend	49.2857143
## 468	1455 weekend	61.4285714
## 469	1500 weekend	27.2857143
## 470	1505 weekend	39.3571429
## 471	1510 weekend	53.2857143
## 472	1515 weekend	61.1428571
## 473	1520 weekend	65.5714286
## 474	1525 weekend	81.2142857
## 475	1530 weekend	67.4285714
## 476	1535 weekend	111.5714286
## 477	1540 weekend	58.2857143
## 478	1545 weekend	107.6428571
## 479	1550 weekend	128.3571429
## 480	1555 weekend	127.8571429
## 481	1600 weekend	111.1428571
## 482	1605 weekend	125.0000000
## 483	1610 weekend	132.2142857
## 484	1615 weekend	150.0714286
## 485	1620 weekend	153.6428571
## 486	1625 weekend	157.0000000
## 487	1630 weekend	112.5000000
## 488	1635 weekend	92.3571429
## 489	1640 weekend	105.0714286
## 490	1645 weekend	88.7142857
## 491	1650 weekend	105.9285714
## 492	1655 weekend	79.8571429
## 493	1700 weekend	120.7142857

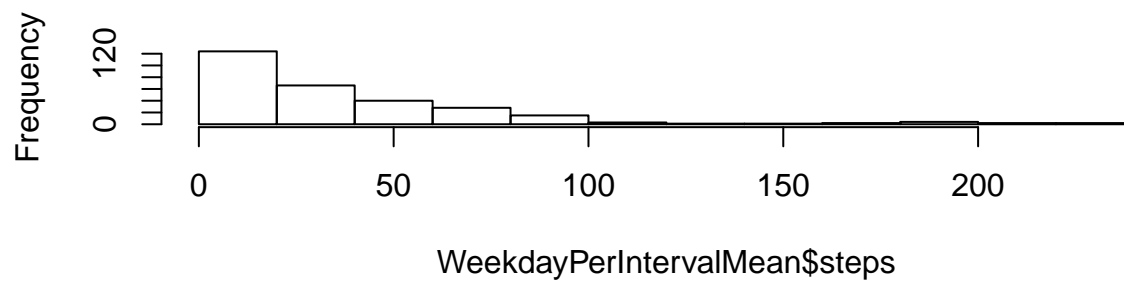
## 494	1705 weekend	92.7857143
## 495	1710 weekend	103.8571429
## 496	1715 weekend	103.5000000
## 497	1720 weekend	113.2142857
## 498	1725 weekend	100.0714286
## 499	1730 weekend	110.0714286
## 500	1735 weekend	39.8571429
## 501	1740 weekend	50.0714286
## 502	1745 weekend	47.4285714
## 503	1750 weekend	35.6428571
## 504	1755 weekend	37.0000000
## 505	1800 weekend	85.7857143
## 506	1805 weekend	94.6428571
## 507	1810 weekend	98.7142857
## 508	1815 weekend	93.9285714
## 509	1820 weekend	52.4285714
## 510	1825 weekend	49.5000000
## 511	1830 weekend	72.7857143
## 512	1835 weekend	50.9285714
## 513	1840 weekend	64.8571429
## 514	1845 weekend	48.0000000
## 515	1850 weekend	39.2857143
## 516	1855 weekend	69.5714286
## 517	1900 weekend	76.2142857
## 518	1905 weekend	79.7857143
## 519	1910 weekend	44.0714286
## 520	1915 weekend	50.0714286
## 521	1920 weekend	31.2857143
## 522	1925 weekend	21.2142857
## 523	1930 weekend	21.9285714
## 524	1935 weekend	20.8571429
## 525	1940 weekend	30.7142857
## 526	1945 weekend	47.9285714
## 527	1950 weekend	50.1428571
## 528	1955 weekend	53.5714286
## 529	2000 weekend	39.6428571
## 530	2005 weekend	62.2857143
## 531	2010 weekend	59.5714286
## 532	2015 weekend	95.1428571
## 533	2020 weekend	85.0000000
## 534	2025 weekend	70.8571429
## 535	2030 weekend	83.6428571
## 536	2035 weekend	66.9285714
## 537	2040 weekend	53.5714286
## 538	2045 weekend	47.7142857
## 539	2050 weekend	52.6428571
## 540	2055 weekend	29.2857143
## 541	2100 weekend	30.6428571
## 542	2105 weekend	11.8571429
## 543	2110 weekend	7.2142857
## 544	2115 weekend	20.2142857
## 545	2120 weekend	6.5714286
## 546	2125 weekend	7.9285714
## 547	2130 weekend	20.6428571

```
## 548      2135 weekend  15.6428571
## 549      2140 weekend  13.6428571
## 550      2145 weekend   8.4285714
## 551      2150 weekend   7.7142857
## 552      2155 weekend   0.0000000
## 553      2200 weekend   1.2142857
## 554      2205 weekend   1.2857143
## 555      2210 weekend   0.0000000
## 556      2215 weekend   0.0000000
## 557      2220 weekend   0.0000000
## 558      2225 weekend   1.7857143
## 559      2230 weekend   0.0000000
## 560      2235 weekend   0.0000000
## 561      2240 weekend   1.2142857
## 562      2245 weekend   0.0000000
## 563      2250 weekend   0.6428571
## 564      2255 weekend  12.9285714
## 565      2300 weekend   2.5000000
## 566      2305 weekend   0.0000000
## 567      2310 weekend   0.0000000
## 568      2315 weekend   0.0000000
## 569      2320 weekend   0.0000000
## 570      2325 weekend   0.6428571
## 571      2330 weekend   1.2142857
## 572      2335 weekend  12.5714286
## 573      2340 weekend   6.7142857
## 574      2345 weekend   1.8571429
## 575      2350 weekend   0.0000000
## 576      2355 weekend   0.0000000
```

```
WeekdayPerIntervalMean <- DayTypePerIntervalMean[DayTypePerIntervalMean$daytype == "weekday", ]
WeekendPerIntervalMean <- DayTypePerIntervalMean[DayTypePerIntervalMean$daytype == "weekend", ]

par(mfrow = c(2, 1) )
hist(WeekdayPerIntervalMean$steps)
hist(WeekendPerIntervalMean$steps)
```

**Histogram of WeekdayPerIntervalMean\$steps**



**Histogram of WeekendPerIntervalMean\$steps**

