Reproducible Research Project 1

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Loading and Preprocessing the Data

```
suppressMessages(library(tidyverse))

## Warning: package 'ggplot2' was built under R version 3.4.4

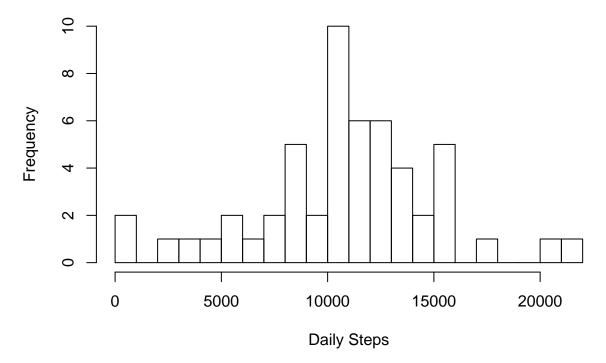
activity <- read.csv('/Users/brodyvogel/Desktop/activity.csv')

# there are a lot of NAs
no_nas <- activity[complete.cases(activity),]</pre>
```

What is the Mean Total Number of Steps Taken Per Day?

```
# aggregate the steps by date
sums <- aggregate(activity$steps ~ activity$date, FUN = sum)
# make the histogram
hist(sums$`activity$steps`, breaks = 25, xlab = 'Daily Steps', main = 'Total Steps by Day')</pre>
```

Total Steps by Day



The mean steps per day was 10766. The median number of steps per day was one less, 10765.

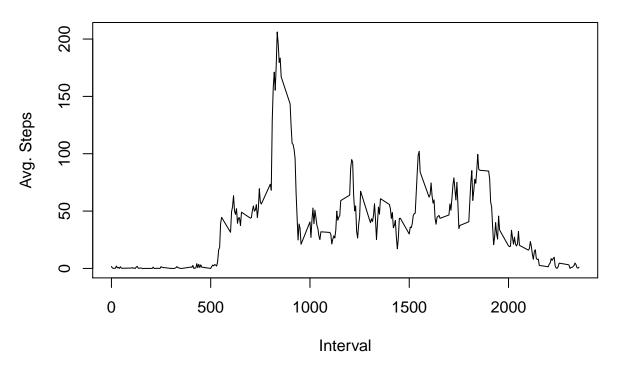
```
# calculate the mean and median
mean(sums[, 2])

## [1] 10766.19
median(sums[, 2])

## [1] 10765
```

What is the Average Daily Activity Pattern?

Average Steps by Interval



The largest number of average steps in a 5-minute interval was 206. The corresponding 5-minute interval was the 835 entry.

```
# calculate the max average steps and the corresponding interval
max_steps <- max(avg_by_interval[, 2])
max_steps

## [1] 206.1698
avg_by_interval[avg_by_interval[, 2] == max_steps, 1]

## [1] 835</pre>
```

Imputing Missing Values

The number of rows containing an NA is 2304.

```
# get the number of rows with an NA in any column
sum(is.na(activity$steps)) + sum(is.na(activity$date)) + sum(is.na(activity$interval))
```

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

My strategy is to fill in the missing values for steps with the average number of steps for that invterval from the rest of the data set.

```
# get just the NA entries
all_nas <- activity[is.na(activity$steps), ]
# throw out the steps observation
all_nas <- all_nas %>% select(c(date, interval))

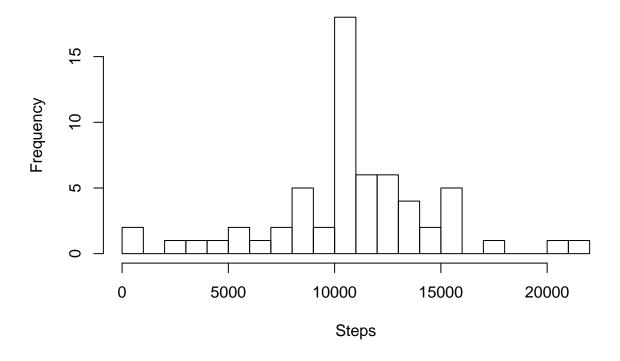
# so the merge works
names(avg_by_interval) <- c('interval', 'steps')

# create the interpolated data
interpolated_nas <- merge(all_nas, avg_by_interval, by = 'interval')

# merge the interpolated data with the clean data from before
interpolated_data <- rbind(no_nas, interpolated_nas)

# aggregate by date again
agg_interpolated_data <- aggregate(interpolated_data$steps ~ interpolated_data$date, FUN = sum)
hist(agg_interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAs Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAS Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAS Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps with NAS Interpolated_data[, 2], xlab = 'Steps', breaks = 25, main = 'Histogram of Steps', breaks = 25, main = 'Hi
```

Histogram of Steps with NAs Interpolated



The mean stays exactly the same, 10,766; with all the interpolated data, now, though, the median goes up one to 10,766 too.

So the data changes a little when observations are interpolated. The distribution is unchanged, however. The effect, then, is that interpolating the missing observations brings more uniformity to the data; whether this is justified or not would depend on the situation. I think it's fine, here.

```
# calculate the mean and median
mean(agg_interpolated_data[, 2])

## [1] 10766.19

median(agg_interpolated_data[, 2])

## [1] 10766.19
```

Are there Differences in Activity Patterns Between Weekdays and Weekends?

The trend lines are quite different depending on whether the activity happends on a weekend or weekday. It looks like activity is more spread out on the weekends, potentially because people are not working in a confined space.

```
# calculate the factor variable
no_nas$day <- weekdays(as.Date(no_nas$date, '%Y-%m-%d'))
no_nas$weekend_or_not <- ifelse(no_nas$day %in% c('Saturday', 'Sunday'), 'weekend', 'weekday')

# create a grouped data frame with averages
to_plot <- no_nas %>% group_by(weekend_or_not, interval) %>%
    summarize(steps = mean(steps))

# plot it
ggplot(to_plot, aes(interval, steps, color = weekend_or_not)) +
    theme_linedraw() +
    geom_line(size = 2) +
    labs(title = 'Average Steps by Day and Interval', x = 'Interval', y = 'Steps', color = "Weekend or We facet_grid(-weekend_or_not) +
    facet_wrap(-weekend_or_not, scales = 'free', nrow = 2)
```

Average Steps by Day and Interval

