# Fitness\_tracker\_project1

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#### Project 1: Fitness tracker

#### Load the data and environment

```
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(ggplot2)
library(data.table)
##
## Attaching package: 'data.table'
## The following objects are masked from 'package:dplyr':
##
       between, first, last
library(stringr)
url <- 'https://github.com/ryeking2010/RepData_PeerAssessment1/raw/master/activity.csv'
data <- read.csv(url)</pre>
```

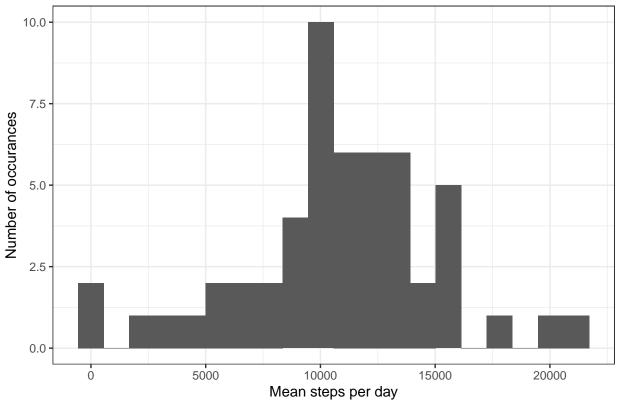
#### Analyzing the data

There are some NA's in here. They will be removed until imputed step. Why? They don't have the device on or working. We cannot say that means zero steps.

#### Let's look at the total number of steps taken each day

```
data_summary <- data[complete.cases(data), ] %>% group_by(date) %>%
  summarize(mean=sum(steps, na.rm = TRUE))
data_summary
## # A tibble: 53 x 2
##
      date
##
      <chr>
                 <int>
##
    1 2012-10-02
                   126
    2 2012-10-03 11352
##
   3 2012-10-04 12116
    4 2012-10-05 13294
##
    5 2012-10-06 15420
    6 2012-10-07 11015
    7 2012-10-09 12811
    8 2012-10-10 9900
   9 2012-10-11 10304
## 10 2012-10-12 17382
## # ... with 43 more rows
data_summary %>% ggplot(aes(x = mean)) + geom_histogram(bins = 20) +
  theme_bw() + labs(x='Mean steps per day', y='Number of occurances',
                    title = 'Frequency of mean steps per day')
```

## Frequency of mean steps per day



## let's look at the mean and steps per day.

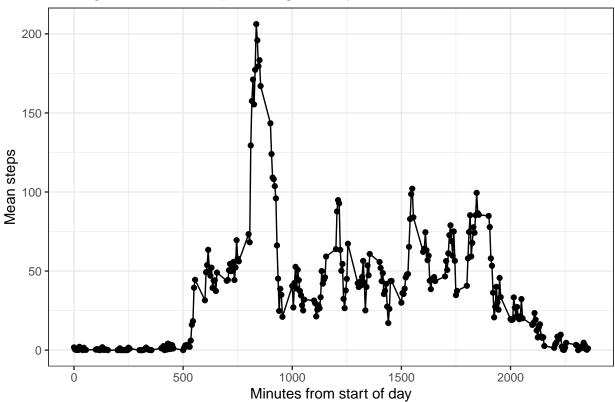
```
data_summary <- data[complete.cases(data), ] %>% group_by(date) %>%
  summarize(StepsSum=sum(steps, na.rm = TRUE))
mean(data_summary$StepsSum)
## [1] 10766.19
median(data_summary$StepsSum)
## [1] 10765
What's the max steps in a given interval?
data_summary.max <- data[complete.cases(data), ] %>% group_by(interval) %>%
  summarize(StepsMax=max(steps, na.rm = TRUE))
tophit <- data_summary.max %>% arrange(desc(StepsMax)) %>% head(n=1)
tophit
## # A tibble: 1 x 2
     interval StepsMax
##
        <int>
                 <int>
## 1
          615
                   806
The 615 interval has the max steps
But that's max... what's the max (highest) average steps taken
data_summary <- data[complete.cases(data), ] %>% group_by(interval) %>%
  summarize(mean=mean(steps, na.rm = TRUE))
data_summary %>% arrange(desc(mean))
## # A tibble: 288 x 2
##
      interval mean
##
         <int> <dbl>
##
           835 206.
  1
## 2
           840 196.
## 3
           850 183.
## 4
           845 180.
## 5
           830 177.
## 6
           820 171.
           855 167.
## 7
           815 158.
## 8
## 9
           825 155.
## 10
           900 143.
## # ... with 278 more rows
max.avg.results <- data_summary %% arrange(desc(mean)) %>% head(n=1)
print(paste("The largest interval is,", max.avg.results$interval, "with",
            max.avg.results$mean, "max steps on average"))
```

## [1] "The largest interval is, 835 with 206.169811320755 max steps on average"

The largest interval is, 835 with 206.169811320755 max steps on average

#### let's look at a time series

# Average number of steps during the day



#### Imputing data

How to impute data? There's many many theories out there 1. mean - on each interval 2. nearest neighbor - for example, steps on a Saturday vs. Monday and take mean 3. So much more, but #2 is really good.

Let's do the mean of the days of the week for the given interval If you really wanted a nice imputation, perhaps a lagging average?

```
data.avg.per.weekday <- data
data.avg.per.weekday$date <- as.Date(data$date) %>% weekdays()
data.avg.per.weekday <- data.avg.per.weekday[complete.cases(data.avg.per.weekday), ] %>%
group_by(date, interval) %>% summarize(mean=mean(steps, na.rm = TRUE))
```

## 'summarise()' has grouped output by 'date'. You can override using the '.groups' argument.

```
data.meanImpute <- data
for (i in 1:nrow(data.meanImpute)){
   if(is.na(data.meanImpute[i, 1])){
      data.meanImpute[i, 1] <- data.avg.per.weekday %>%
        filter(date == weekdays(as.Date(data[i, 2])) & interval == data[i, 3]) %>%
        .[[3]]
   }
}
print("Checking data:")

## [1] "Checking data:"

any(is.na(data.meanImpute))

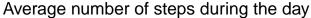
## [1] FALSE

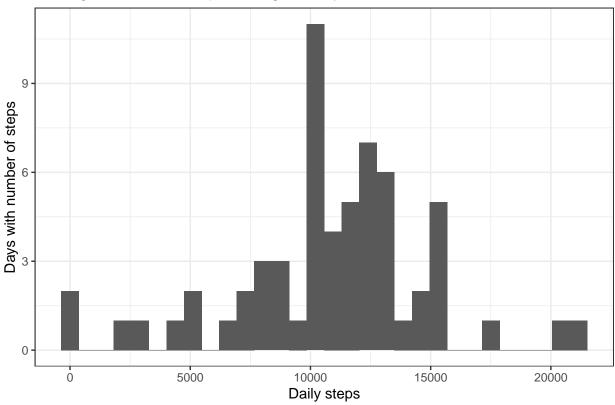
# cha-ching
```

#### Visualizing imputed data

Let's make a histogram of the total number of steps taken each day after missing values are imputed

## 'stat\_bin()' using 'bins = 30'. Pick better value with 'binwidth'.

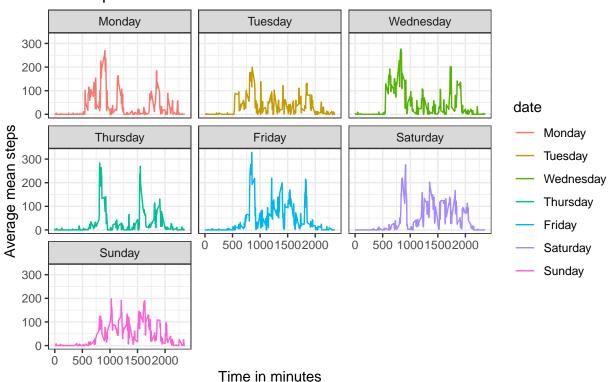




Let's use this for a days of the week plot

## 'summarise()' has grouped output by 'date'. You can override using the '.groups' argument.

# Average steps across days of the week with imputed data



Looks like people are more active saturday night and weekday mornings

They may want weekdays VS weekends, not all days.

```
data.meanImpute$date <- str_replace(data.meanImpute$date, "Saturday", "Weekend")
data.meanImpute$date <- str_replace(data.meanImpute$date, "Sunday", "Weekend")
data.meanImpute$date <- if_else(data.meanImpute$date == 'Weekend', 'Weekend', 'Weekday')
data.meanImpute.final <- data.meanImpute %>% group_by(date, interval) %>%
    summarise(meanSteps = mean(steps))
```

## 'summarise()' has grouped output by 'date'. You can override using the '.groups' argument.

# Average steps across days of the week with imputed data

