# Practice Assignment

The goal of this assignment is to provide a "bridge" between the first two weeks of lectures and assignment 1 for those either new to R or struggling with how to approach the assignment.

This guided example, will **not** provide a solution for programming assignment 1. However, it will guide you through some core concepts and give you some practical experience to hopefully make assignment 1 seems less daunting.

To begin, download this file and unzip it into your R working directory. http://s3.amazonaws.com/practice\_assignment/diet\_data.zip

You can do this in R with the following code:

```
dataset_url <- "http://s3.amazonaws.com/practice_assignment/diet_data.zip"
download.file(dataset_url, "diet_data.zip")
unzip("diet_data.zip", exdir = "diet_data")</pre>
```

If you're not sure where your working directory is, you can find out with the getwd() command. Alternatively, you can view/change it through the Tools > Global Options menu in R Studio.

So assuming you've unzipped the file into your R directory, you should have a folder called diet\_data. In that folder there are five files. Let's get a list of those files:

```
list.files("diet_data")
## [1] "Andy.csv" "David.csv" "John.csv" "Mike.csv" "Steve.csv"
```

Okay, so we have 5 files. Let's take a look at one to see what's inside:

```
andy <- read.csv("diet_data/Andy.csv")
head(andy)</pre>
```

```
##
    Patient.Name Age Weight Day
## 1
             Andy 30
                          140
                                1
## 2
             Andy
                   30
                          140
                                2
             Andy
                   30
## 3
                          140
                                3
                   30
             Andy
                          139
## 5
             Andy
                   30
                          138
                                5
## 6
             Andy
                   30
```

It appears that the file has 4 columns, Patient.Name, Age, Weight, and Day. Let's figure out how many rows there are by looking at the length of the 'Day' column:

```
length(andy$Day)
```

```
## [1] 30
```

30 rows. OK.

Alternatively, you could look at the dimensions of the data.frame:

## dim(andy)

```
## [1] 30 4
```

This tells us that we 30 rows of data in 4 columns. There are some other commands we might want to run to get a feel for a new data file, str(), summary(), and names().

```
str(andy)
```

```
## 'data.frame': 30 obs. of 4 variables:
## $ Patient.Name: Factor w/ 1 level "Andy": 1 1 1 1 1 1 1 1 1 1 1 1 ...
## $ Age : int 30 30 30 30 30 30 30 30 30 ...
## $ Weight : int 140 140 140 139 138 138 138 138 138 138 ...
## $ Day : int 1 2 3 4 5 6 7 8 9 10 ...
```

#### summary(andy)

```
Patient.Name
                                     Weight
                        Age
                                                      Day
##
    Andy:30
                                                        : 1.00
                  \mathtt{Min}.
                          :30
                                 Min.
                                         :135
                                                Min.
                                 1st Qu.:137
##
                   1st Qu.:30
                                                 1st Qu.: 8.25
##
                  Median:30
                                 Median:138
                                                Median :15.50
##
                  Mean
                           :30
                                 Mean
                                         :137
                                                Mean
                                                        :15.50
                                                 3rd Qu.:22.75
##
                   3rd Qu.:30
                                 3rd Qu.:138
##
                  Max.
                           :30
                                 Max.
                                         :140
                                                 Max.
                                                        :30.00
```

# names (andy)

```
## [1] "Patient.Name" "Age" "Weight" "Day"
```

So we have 30 days of data. To save you time, all of the other files match this format and length. I've made up 30 days worth of weight data for 5 subjects of an imaginary diet study.

Let's play around with a couple of concepts. First, how would we see Andy's starting weight? We want to subset the data. Specifically, the first row of the 'Weight' column:

```
andy[1, "Weight"]
```

#### ## [1] 140

We can do the same thing to find his final weight on Day 30:

```
andy[30, "Weight"]
```

# ## [1] 135

Alternatively, you could create a subset of the 'Weight' column where the data where 'Day' is equal to 30.

```
andy[which(andy$Day == 30), "Weight"]
```

## [1] 135

```
andy[which(andy[,"Day"] == 30), "Weight"]
```

## [1] 135

Or, we could use the subset() function to do the same thing:

```
subset(andy$Weight, andy$Day==30)
```

## [1] 135

There are lots of ways to get from A to B when using R. However it's important to understand some of the various approaches to subsetting data.

Let's assign Andy's starting and ending weight to vectors:

```
andy_start <- andy[1, "Weight"]
andy_end <- andy[30, "Weight"]</pre>
```

We can find out how much weight he lost by subtracting the vectors:

```
andy_loss <- andy_start - andy_end
andy_loss</pre>
```

## [1] 5

Andy lost 5 pounds over the 30 days. Not bad. What if we want to look at other subjects or maybe even everybody at once?

Let's look back to the list.files() command. It returns the contents of a directory in alphabetical order. You can type ?list.files at the R prompt to learn more about the function.

Let's take the output of list.files() and store it:

```
files <- list.files("diet_data")
files</pre>
```

```
## [1] "Andy.csv" "David.csv" "John.csv" "Mike.csv" "Steve.csv"
```

Knowing that 'files' is now a list of the contents of 'diet\_data' in alphabetical order, we can call a specific file by subsetting it:

```
files[1]
```

## [1] "Andy.csv"

#### files[2]

```
## [1] "David.csv"
```

```
files[3:5]
```

```
## [1] "John.csv" "Mike.csv" "Steve.csv"
```

Let's take a quick look at John.csv:

```
head(read.csv(files[3]))
```

```
## Warning: cannot open file 'John.csv': No such file or directory
```

```
## Error: cannot open the connection
```

Woah, what happened? Well, John.csv is sitting inside the diet\_data folder. We just tried to run the equivalent of read.csv("John.csv") and R correctly told us that there isn't a file called John.csv in our working directory. To fix this, we need to append the directory to the beginning of the file name.

One approach would be to use paste() or sprintf(). However, if you go back to the help file for list.files(), you'll see that there is an argument called full.names that will append (technically prepend) the path to the file name for us.

```
files_full <- list.files("diet_data", full.names=TRUE)
files_full</pre>
```

```
## [1] "diet_data/Andy.csv" "diet_data/David.csv" "diet_data/John.csv"
## [4] "diet_data/Mike.csv" "diet_data/Steve.csv"
```

Pretty cool. Now let's try taking a look at John.csv again:

```
head(read.csv(files_full[3]))
```

```
##
     Patient.Name Age Weight Day
## 1
             John 22
                         175
## 2
             John 22
                         175
             John 22
## 3
                         175
                                3
## 4
             John
                   22
                         175
                                4
## 5
             John
                   22
                         175
                                5
## 6
             John 22
                         175
                                6
```

Success! So what if we wanted to create one big data frame with everybody's data in it? We'd do that with rbind and a loop. Let's start with rbind:

```
andy_david <- rbind(andy, read.csv(files_full[2]))</pre>
```

This line of code took our existing data frame, Andy, and added the rows from David.csv to the end of it. We can check this with:

# head(andy\_david)

```
##
     Patient.Name Age Weight Day
## 1
              Andy
                    30
                           140
                                  1
## 2
                                  2
              Andy
                    30
                           140
## 3
                                  3
              Andy
                    30
                           140
## 4
              Andy
                    30
                           139
                                  4
## 5
              Andy
                    30
                           138
                                  5
## 6
              Andy
                    30
                           138
                                  6
```

# tail(andy\_david)

```
##
      Patient.Name Age Weight Day
## 55
             David 35
                           203
                                25
             David
## 56
                    35
                           203
                                26
## 57
             David
                    35
                           202
                                27
                           202
                                28
## 58
             David
                    35
## 59
             David
                    35
                           202
                                29
## 60
                           201
                                30
             David
                    35
```

One thing to note, rbind needs 2 arguments. The first is an existing data frame and the second is what you want to append to it. This means that there are occassions when you might want to create an empty data frame just so there's *something* to use as the existing data frame in the rbind argument.

Don't worry if you can't imagine when that would be useful because you'll see an example in just a little while.

Now, let's create a subset of the data frame that shows us just the 25th day for Andy and David.

```
day_25 <- andy_david[which(andy_david$Day == 25), ]
day_25</pre>
```

```
## Patient.Name Age Weight Day
## 25 Andy 30 135 25
## 55 David 35 203 25
```

Now you could manually go through and append everybody's data to the same data frame using rbind, but that's not a practical solution if you've got lots and lots of files. So let's try using a loop.

To understand what's happening in a loop, let's try something:

```
for (i in 1:5) {print(i)}
```

```
## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
```

As you can see, for each pass through the loop, i increases by 1 from 1 through 5. Let's apply that concept to our list of files.

```
for (i in 1:5) {
         dat <- rbind(dat, read.csv(files_full[i]))
}</pre>
```

## Error: object 'dat' not found

Whoops. Object 'dat' not found. This is because you can't rbind something into a file that doesn't exist yet. So let's create an empty data frame called 'dat' before running the loop.

```
dat <- data.frame()</pre>
for (i in 1:5) {
        dat <- rbind(dat, read.csv(files_full[i]))</pre>
}
str(dat)
## 'data.frame':
                    150 obs. of 4 variables:
## $ Patient.Name: Factor w/ 5 levels "Andy", "David", ...: 1 1 1 1 1 1 1 1 1 1 ...
   $ Age
                  : int 30 30 30 30 30 30 30 30 30 ...
   $ Weight
                  : int 140 140 140 139 138 138 138 138 138 ...
##
                        1 2 3 4 5 6 7 8 9 10 ...
   $ Day
                  : int
```

Cool. We now have a data frame called 'dat' with all of our data in it. Out of curiousity, what would happen if we had put dat <- data.frame() inside of the loop? Let's see:

```
for (i in 1:5) {
          dat2 <- data.frame()
           dat2 <- rbind(dat2, read.csv(files_full[i]))
}
str(dat2)</pre>
```

## head(dat2)

```
##
    Patient.Name Age Weight Day
## 1
            Steve 55
                         225
                               1
## 2
            Steve 55
                         225
                               2
## 3
            Steve 55
                         225
                               3
## 4
            Steve 55
                         224
                               4
## 5
            Steve 55
                         224
                               5
## 6
            Steve 55
                         224
```

Because we put dat2 <- data.frame() inside of the loop, dat2 is being rewritten with each pass of the loop. So we only end up with the data from the last file in our list.

Back to dat... So what if we wanted to know the median weight for all the data? Let's use the median() function.

#### median(dat\$Weight)

#### ## [1] NA

NA? Why did that happen? Type 'dat' into the console and you'll see a print out of all 150 obversations. Scroll back up to row 77, and you'll see that we have some missing data from John, which is recorded as NA by R.

We need to get rid of those NA's for the purposes of calculating the median. There are several approaches. For instance, we could subset the data using complete.cases() or is.na(). But if you look at ?median, you'll see there is an argument called na.rm that will strip the NA values out for us.

```
median(dat$Weight, na.rm=TRUE)
```

```
## [1] 190
```

So 190 is the median weight. We can find the median weight of day 30 by taking the median of a subset of the data where Day=30.

```
dat_30 <- dat[which(dat[, "Day"] == 30),]
dat_30</pre>
```

```
##
       Patient.Name Age Weight Day
## 30
                Andy
                       30
                             135
                                   30
## 60
               David
                             201
                                   30
                       35
## 90
                John
                       22
                             177
                                   30
## 120
                       40
                             192
                                   30
                Mike
## 150
               Steve
                       55
                             214
                                   30
```

```
median(dat_30$Weight)
```

```
## [1] 192
```

We've done a lot of manual data manipulation so far. Let's build a function that will return the median weight of a given day.

Let's start out by defining what the arguments of the function should be. These are the parameters that the user will define. The first parameter the user will need to define is the directory that is holding the data. The second parameter they need to define is the day for which they want to calculate the median.

So our function is going to start out something like this:

```
weightmedian <- function(directory, day) { # content of the function }</pre>
```

So what goes in the content? Let's think through it logically. We need a data frame with all of the data from the CSV's. We'll then subset that data frame using the argument day and take the median of that subset.

In order to get all of the data into a single data frame, we can use the method we worked through earlier using list.files() and rbind().

Essentially, these are all things that we've done in this example. Now we just need to combine them into a single function.

So what does the function look like?

You can test this function by running it a few different times:

```
weightmedian(directory = "diet_data", day = 20)

## [1] 197.5

weightmedian("diet_data", 4)

## [1] 188

weightmedian("diet_data", 17)
```

## [1] 198

Hopefully, this has given you some practice applying the basic concepts from weeks 1 and 2. If you can work your way through this example, you should have all of the tools needed to complete part 1 of assignment 1. Parts 2 and 3 are really just expanding on the same basic concepts, potentially adding in some ideas like cbinds and if-else.

One last quick note: The approach I'm showing above for building the data frame is submoptimal. It works, but generally speaking, you don't want to build data frames or vectors by copying and re-copying them inside of a loop. If you've got a lot of data it can become very, very slow. However, this tutorial is meant to provide an introduction to these concepts, and you can use this approach successfully for programming assignments 1 and 3.

If you're interested in learning the better approach, check out Hadley Wickam's excellent material on functionals within R: <a href="http://adv-r.had.co.nz/Functionals.html">http://adv-r.had.co.nz/Functionals.html</a>. But if you're new to both programming and R, I would skip it for now as it will just confuse you. Come back and revisit it once you are able to write working functions using the approach above.