# Learn you some R

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April 9, 2015

## 1 Introduction

This guide is meant for people with little or no programming experience. That being said, this guide moves quickly and is aimed towards people who have spent a lot of time learning. There are many guides for programming which are aimed at computer scientists and use lots of jargon computer scientists have heard for years. This can be difficult to understand. On the other hand, many guides are too simplified such that you can't really do anything interesting with the material taught. This guide attempts to provide meaningful examples and explain the concepts which cannot be easily Googled. This guide comes with a supplementary R file which contains all of the example code used in this guide. If you have any feedback please mail me at sam.d.pollard@gmail.com

R is what is called an *open* project. That means that anyone can contribute to R. This results in a large number of what are called *packages*, which consist of a bunch of R files which "work together" to provide a service. This also results in R being really really huge in terms of the number of features it provides.

To install R, you must download it. I found the easiest way to do this is to go to http://cran.cs.wwu.edu/, which is hosted by WWU. You then install it like any normal program.

There are two basic ways to interact with R. One is in *interactive* mode, the other is by reading in a source file. The latter can be done in windows by clicking *File* then *Source R Code...*. Alternatively, you can edit an R source file by clicking *File* then *Open Script*. Interactive mode is useful for temporary or scratch work, while saving the commands you want to run into a text file and running them is more useful for automation and running experiments more than once.

Here is a sample run using interactive mode. The [1] symbol is a way of saying "here is the first element of the output which you're not storing anywhere." As we will see, it is common to store the output (or return value) of a function call in a variable. The symbol > is called a prompt, which is the interpreter's (we'll define what that is later) way of saying "I'm ready to process another command." Anything following a # is a comment. This is to make the commands more human-readable, but are ignored by R entirely.

```
> data(trees) # Load a sample dataset
> nrow(trees) # Count the sample size
```

<sup>&</sup>lt;sup>1</sup>Technically, it refers to the first element of a one-element vector. These will be explained later.

```
[1] 31
> colnames(trees)
[1] "Girth" "Height" "Volume"
> max(trees["Volume"]) # Find the largest in the column
[1] 77
> max(trees[3]) # The same as above because we access the third column
[1] 77
> apply(trees, 2, mean)
    Girth Height Volume
13.24839 76.00000 30.17097
```

The last command is a bit complicated but fits with intuition. The apply function takes as an argument some data frame (in this case trees), a 1 (for row), 2 (for columns), and a function (mean here), and applies that function to each row or column in the data set. It is generally considered good practice to use built-in functions or "one-liners" such as apply because they make code easier to understand and maintain.

# 2 Basic Syntax and Examples

R is what is called an *interpreted* language. This means that there is an aptly-named *interpreter* which keeps track of all the variables you have and how to execute the commands you run. The benefit to being interpreted is you have the flexibility of running one line at a time.

One potential drawback of an interpreted language is it can be difficult to find the errors in your code, because they may not show up until the source code is run. This can be mitigated by testing small chunks of your code as you write it.

In the first example, the object trees is what is called a *data frame*. In R, this is denoted data.frame. This can be thought of as your typical spreadsheet format: A bunch of columns together. The one restriction is that each column must have equal length. Here is a quick example:

```
# c can be thought of as "combine"
> name <- c("H","He","Li")</pre>
> mass < c(1.0079,4.0026,6.941) # i.e. make a vector from the arguments
> atomic_number <- seq(1,length(name))</pre>
> mydf <- data.frame(atomic_number, name, mass)</pre>
> mydf
                                  # Print out the data.frame
  atomic_number name
                        mass
1
               1
                    H 1.0079
2
               2
                   He 4.0026
3
                   Li 6.9410
```

A few remarks are in order. First, the <- symbol is used. This performs an assignment of the right-hand value to the left-hand value. The <- symbol is read as "gets." This is the official way to do things, but you will often see code such as name = c("H","He","Li") instead, which is also assignment (this is common syntax in other programming languages).

I recommend using <- because = is used in another context in R, and it makes the distinction clearer. The fourth line calls the data.frame function which takes as input some number of vectors and returns a data frame. This return value is stored in the variable mydf. This is a general from of function calling, which follows the notation from mathematics: the function name followed by the arguments in parentheses. The name, mass, and atomic\_number form the header of the data frame. Sometimes data sets have headers, sometimes they don't. When loading in a csv you may specify whether you want R to try to put the first row as the header.

Unlike most other programming languages, the period can be used in variable names. It is typically used to denote a more specific way to go about things. For example, the function read is used in a lot of contexts: It can be used to ask the user for input, it can read files, and many other things. A very useful "version" of read is read.csv, which takes in a csv file and turns it into a data frame.

Consider the following example, which will allow us to read in a csv and allow us to use all of R's features to analyze the data:

```
> mydf <- read.csv("report.csv", header = TRUE, sep = ",")
```

The last argument is the *separator*. This may be whatever you want. Cryptically, a csv (comma separated value) file may not be separated by commas. Oftentimes, it is tabseparated. To accomodate this, one can say sep = "\t". The backslash here *escapes* the succeeding character t and instead interprets this as a tab.

# 3 Some Subtleties

### 3.1 Vectors

The previous example used the c function. The reason this is used is to put data in the correct format. This is an important part of R and programming in general. The distinction here is between a *vector* and a *scalar*. This fits with mathematical intuition but not so much with the real world. Ask a mathematician what a vector is and she will respond, "It is something that behaves like a vector." While this is sort of a joke, there really isn't any better definition. In R, vectors are all over the place. Here are some examples and a non-example (can you spot which?)

```
n <- nrow(trees)  # 1. The number of trees
tag <- seq(1,n)  # 2. A sequence from 1 to n, inclusive
lifespan <- 70*runif(n)  # 3. Random values between 0 and 70, inclusive
volume <- trees["Volume"]  # 4. The Volume of each tree</pre>
```

The last example is *not* a vector. It is a data frame. This means things that require vectors won't work on the data frame. For example,

```
> mean(volume)
[1] NA
Warning message:
In mean.default(volume) : argument is not numeric or logical: returning NA
```

However, if we make a small modification:

```
> volumevec <- trees[["Volume"]] # 5.</pre>
```

This is a vector.

If you're stuck, the class function may describe what sort of data you're dealing with. Another useful method is ls() which lists all the variables in your current workspace.

But what about the first example? Is it a vector? In some sense, yes. A vector can have one element. It is easiest to think of a scalar as the special vector which has one element. In general, most things you can do with scalars you can do with vectors. For example,

```
lifespan <- lifespan + 15
```

would increase the lifespan variable by 15 (add 15 to each element in lifespan). Notice that this statement contains lifespan on the right-hand side and the left-hand side. In English we would say, "take each element in lifespan and add 15 to it, storing the result in the variable lifespan."

## 3.2 Building Up

A common task when working with spreadsheets is combining data from multiple sources into a single structure. Here, we will look at the cbind, and paste functions. We will keep using our previous examples and create a data frame which represents all of the vectors we have created. We begin this somewhat contrived example by creating "names" for each of the trees.

```
> treenames <- paste("T", tag, sep = "")</pre>
```

This creates a sequence of strings. Notice that R is smart enough to determine that while you only specified a single string "T" (instead of a vector of strings of the same length as tag), it concatenates "T" with each element of tag. For example, tree 15 would be named T15.

Now, to create a new data frame from the existing data we created, try

```
> mydf <- cbind(treenames, trees, lifespan)
```

This is close to what we want. The function cbind takes as input data frames or vectors, and combines them *by column* into a new data frame. There is an analogous function called rbind which combines the data by row. Here is what the first few entries look like:

#### > head(mydf)

	treenames	${\tt Girth}$	Height	Volume	lifespan
1	T1	8.3	70	10.3	15.18891
2	T2	8.6	65	10.3	15.85794
3	Т3	8.8	63	10.2	15.41223
4	T4	10.5	72	16.4	15.30183
5	T5	10.7	81	18.8	15.86208
6	Т6	10.8	83	19.7	15.87640

By the way, head is a nice way to just see the beginning of a data frame (it is common to deal with thousands of rows of data). But this doesn't look quite right: the row names are just 1, 2, 3, ... when we want them to represent our cleverly-named tree names. R allows us to change row names and column names using the aptly-named functions rownames and columns. So,

### > rownames(mydf) <- treenames

We can also retrieve the row or column names by putting this function call on the right-hand side of an assignment like so: names <- rownames(mydf). To anyone with programming experience, this is a bit strange. The function can be used both as an *l*-value, a.k.a. the left-hand side of an assignment statement a.k.a. the location of the variable getting changed) and an *r*-value, a.k.a. the right-hand side of an assignment statement a.k.a. the return value of a function call. This is where the <- notation keeps us sane: it directs us where the data we computed is being stored.

However, this data frame looks a bit strange:

### > head(mydf)

	${\tt treenames}$	${\tt Girth}$	Height	Volume	lifespan
T1	T1	8.3	70	10.3	15.18891
T2	T2	8.6	65	10.3	15.85794
ТЗ	T3	8.8	63	10.2	15.41223
T4	T4	10.5	72	16.4	15.30183
T5	T5	10.7	81	18.8	15.86208
Т6	Т6	10.8	83	19.7	15.87640

So let's delete that first row:

That is some cryptic R code. There are other ways to accomplish the same task, but this one easily generalizes. In general, we access things by [row, column]. So by default, we are accessing every row. That is, there is nothing preceding the comma. Now, we are accessing every column but the first one (think of – as removing the first column). Alternatively, we could *include* every other column. So

$$> mydf <- mydf[,c(2,3,4,5)]$$

would accomplish the same task. If we only cared about a few rows of mydf we could write

$$>$$
 sample  $<$ - mydf[seq(4,9),]

This takes the 4th through 9th rows of mydf.

Learn you some R 5 RESOURCES

### 3.3 Default Parameters

Returning to the first example of creating the tree names, we had to specify that sep = "", or the *empty string*. If this was left out, i.e. paste("T",tag), then by default the strings are concatenated using a space as separator. We would get T 15 instead of T15. Why is this? The motivation behind this is that most of the time a user will want the values separated by spaces. Without anything, there is the *default parameter* for sep, which is a space. Parameter specification takes the general form tag = value. This is common notation, and = is the character to use when specifying these parameters.

# 4 Plotting

R has very powerful plotting features. They are described as "beautiful" by data visualization geeks. One reason for this is that R can output in vector formats, which is to say they scale to any size. There may be many more supported filetypes, but arguably the most important are png and pdf which this guide will explain.

The Resources section contains a link to a web page describing the par function, which can be thought of as *parameters* for plotting. Because of R's default function arguments, you will almost never have to worry about all of the dozens of parameters.

Without further ado, we provide an extended example which plots a bunch of random points and fits a curve to it. Which if you think about it, is the solution to about half of all scientific problems.

## 5 Resources

Arguably the most useful resource you can make use of in R is the help function. You pass help any function as an argument, which will direct you to an online help source. For example, help(rbind). This is the same as ?rbind.

Here are some other resources.

- http://learnxinyminutes.com/docs/r/ This is a pretty basic guide but is the easiest to follow.
- https://github.com/sampollard/pcrystal. This is a bit complicated but I've found looking at other people's code can help.
- http://cran.r-project.org/doc/contrib/Short-refcard.pdf This is a useful cheat sheet. Much of it won't make sense right away, but will help you with commonly-used functions.
- https://stat.ethz.ch/R-manual/R-patched/library/graphics/html/par.html This is a documentation file on the par function which is used for plots. In general, the stat.ethz.ch website is a good resource (if you come across it in Google searches).