## An Introduction to R

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These materials introduce R, as originally developed as S, by John *Chambers* and others at Bell Laboratories in 1976, and implemented and made into an Open Source program by Robert *Gentleman* and Ross *Ihaka* in 1995.

As you learn R, there is nothing wrong with making errors when learning a programming language like R.

You learn from your mistakes, and there is no harm done.

Try out the code embedded into the accompanying text and experiment with new variations to discover how the system will respond.

<sup>1</sup> https://www.r-project.org/Licenses/GPL-2

### Downloading and installing R and RStudio



R can be downloaded for free from CRAN\*.

A *binary version* is usually simplest to use and can be installed in Windows and Mac fairly easily.

A binary version is available for Windows from the web page

http://cloud.r-project.org/bin/windows/base.

The "setup program" setup is usually a file with a name like

R-3.6.1-win.exe

Clicking on this file will start an almost automatic installation of the R system. Clicking "Next" several times is often all that is necessary in order to complete the installation.

<sup>\*</sup>http://cloud.r-project.org

### Downloading and installing R and RStudio



An R icon ( will appear on your computer's desktop upon completion.

RStudio is also very popular. You can download the "Open Source Edition" of "RStudio Desktop" from http://www.rstudio.com/, and follow the instructions to install it on your computer.

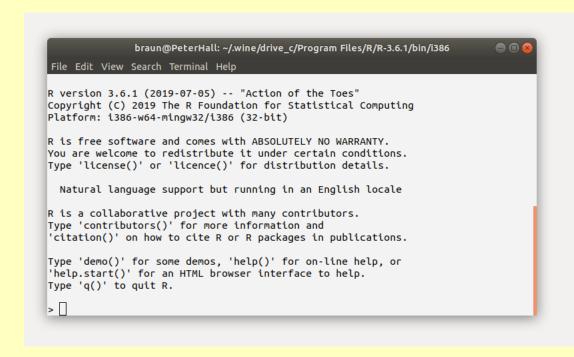
Although much or all of what is described here can be carried out in RStudio, there will be little further comment about that environment.

Thus, you might find that some of the instructions to be carried out at the command line can also be carried out with the menu system in RStudio.





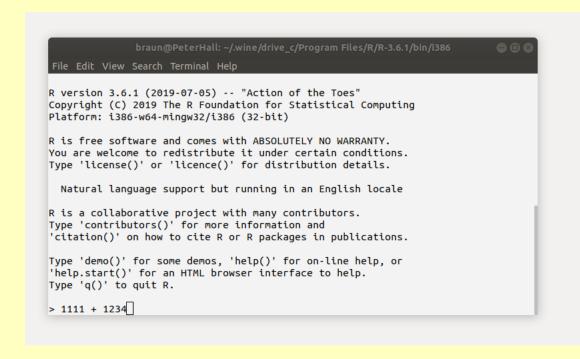
Clicking on the R icon, or opening RStudio similarly, should provide you with access to a window or pane, called the *R console* in which you can execute commands.



The > sign is the R prompt which indicates where you can type in the command to be executed.



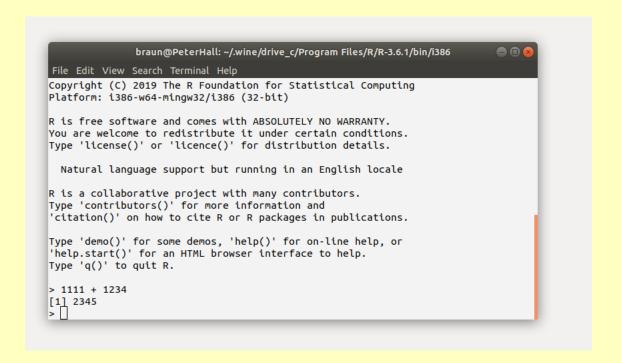
### You can do arithmetic of any type, including multiplication:



By hitting the "Enter" key, you are asking R to execute this calculation.



### The answer appears on the next line:



Often, you will type in commands such as this into a script window, as in RStudio, for later execution, through hitting "ctrl-R" or another related keystroke sequence.





Objects that are built in to R or saved in your workspace, i.e. the environment in which you are currently doing your calculations, can be displayed, simply by invoking their name.

For example,

the data set or data frame called women contains information on heights and weights of American women:

#### > women

##		height	weight	
##	1	58	115	
##	2	59	117	
##	3	60	120	
##	4	61	123	
##	5	62	126	
##	6	63	129	
##	7	64	132	
##	8	65	135	
##	9	66	139	
##	10	67	142	
##	11	68	146	
##	12	69	150	
##	13	70	154	
##	14	71	159	
##	15	72	164	



One of the major strengths of R is the availability of add-on *packages* that have been created by statisticians and computer scientists from around the world.

There are thousands of packages, e.g. graphics, ggplot2, and MPV.

A package contains functions and data which extend the abilities of R.

Every installation of R contains a number of packages by default (e.g. base, stats, and graphics) which are automatically loaded when you start R.



To load an additional package, for example, called *DAAG*, type

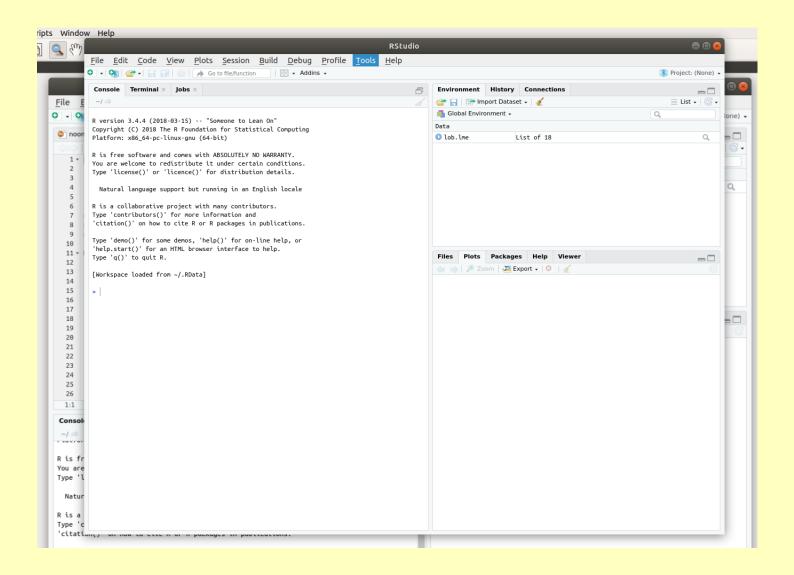
library (DAAG)

If you get a warning that the package is can't be found, then the package doesn't exist on your computer, but it can likely be installed. Try

install.packages("DAAG")

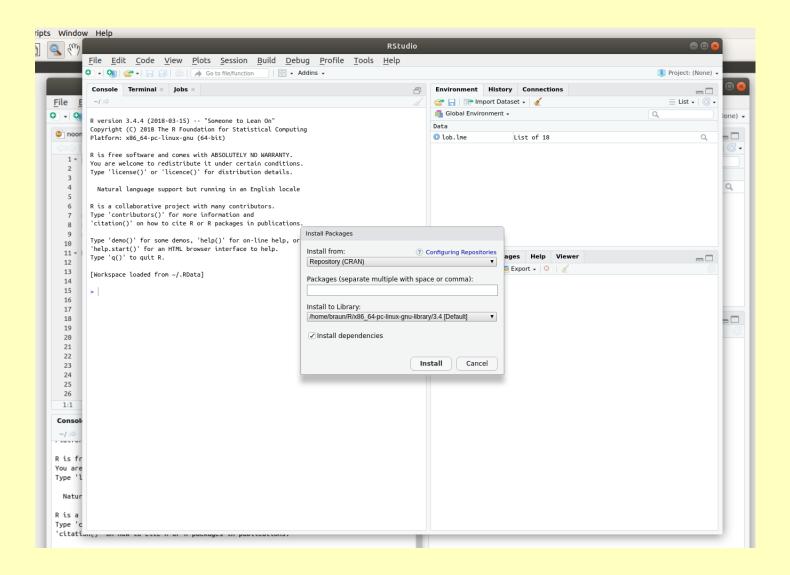


## In RStudio, it may be simpler to use the Tools menu.



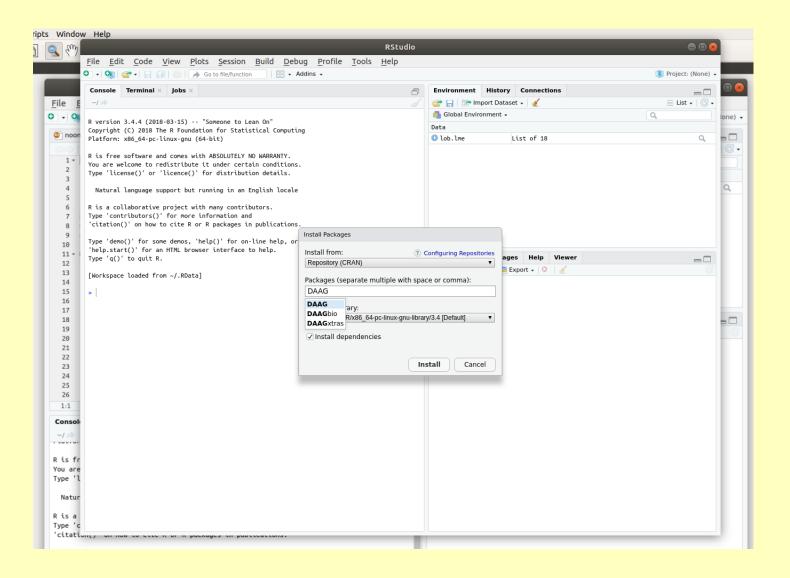


## **Choose "Install Packages":**





## Type in the name of the package you are requesting, and click "Install":







Once DAAG is installed, it can be loaded using the library() function, and you can access data frames and functions that were not available previously.

```
library (DAAG)
```

### For example, the seedrates data frame is now available:

```
## rate grain
## 1 50 21.2
## 2 75 19.9
## 3 100 19.2
## 4 125 18.4
## 5 150 17.9
```



## Using one object from a package at a time

The MPV package is installed on my system, but I have not loaded it. I only want to access the p2.12 data frame and nothing else.

To do this, just type the package name (MPV), followed by two colons (::) and the object name you seek.

```
MPV::p2.12
##
     temp usage
## 1
       21 185.79
## 2
       24 214.47
## 3
       32 288.03
## 4
       47 424 84
## 5
       50 454.68
## 6
        59 539 03
## 7
        68 621.55
## 8
       74 675.06
## 9
        62 562.03
       50 452.93
## 10
       41 369.95
##
## 12
       30 273.98
```

### **Calculations in R**



You can control the number of digits in the output with the options () function.

This is useful when reporting final results such as means and standard deviations, since including excessive numbers of digits can give a misleading impression of the accuracy in your results.

## **Compare**

```
583/31
## [1] 18.80645
```

with



## Observe the patterns in the following calculations.

```
options (digits = 18)
11111111111111
## [1] 1234567654321
111111111111111
## [1] 123456787654321
11111111111111111
## [1] 12345678987654320
```

The error in the final calculation is due to the way R stores information about numbers.

There are around 17 digits of numeric storage are avaiable.

#### **Data frames**



Most data sets are stored in R as data frames, such as the women object we encountered earlier.

Data frames are like matrices, but where the columns have their own names.

Columns can be of numeric type (numbers), character type (character strings in quotation marks), logical type (TRUE or FALSE), or factors (more on this later).

It is generally unwise to inspect data frames by printing their entire contents to your computer screen, as it is far better to use graphical procedures to display large amounts of data or to exploit numerical summaries.



The *summary()* function provides information about the main features of a data frame:

```
## height weight
## Min. :58.0 Min. :115
## 1st Qu.:61.5 1st Qu.:124
## Median :65.0 Median :135
## Mean :65.0 Mean :137
## 3rd Qu.:68.5 3rd Qu.:148
## Max. :72.0 Max. :164
```



## Columns can be of different types from each other. An example is the built-in chickwts data frame:

```
## weight feed

## Min. :108 casein :12

## 1st Qu.:204 horsebean:10

## Median :258 linseed :12

## Mean :261 meatmeal :11

## 3rd Qu.:324 soybean :14

## Max. :423 sunflower:12
```

One column is of factor type while the other is numeric.



If you want to see the first few rows of a data frame, you can use the *head()* function:

```
head (chickwts)

## weight feed

## 1 179 horsebean

## 2 160 horsebean

## 3 136 horsebean

## 4 227 horsebean

## 5 217 horsebean

## 6 168 horsebean
```

The tail() function displays the last few rows.

## **Data frames**



The number of rows can be determined using the *nrow()* function:

```
nrow(chickwts)
## [1] 71
```

Similarly, the *ncol()* function counts the number of columns.



## The str() function is another way to extract information about a data frame:

```
str(chickwts)

## 'data.frame': 71 obs. of 2 variables:

## $ weight: num 179 160 136 227 217 168 108 124 143 140 ...

## $ feed : Factor w/ 6 levels "casein", "horsebean", ..: 2 2 2 2 2 2 2 2
```

### My own preference is the summary () function:

```
summary(chickwts)
##
                    feed
  weight
##
  Min. :108 casein :12
##
  1st Qu.:204 horsebean:10
##
  Median :258 linseed :12
  Mean :261 meatmeal :11
##
##
  3rd Qu.:324 soybean :14
##
  Max. :423 sunflower:12
```





If you have prepared the data set yourself, you could simply type it into a text file, for example called mydata.txt, perhaps with a header indicating column names, and where you use blank spaces to separate the data entries.

The read.table() function will read in the data for you as follows:

```
mydata <- read.table("mydata.txt", header = TRUE)</pre>
```

The object mydata now contains the data read in from the external file.





You could use any name that you wish in place of mydata, as long as the first element of its name is an alphabetic character.

If the data entries are separated by commas and there is no header row, as in the file wx\_I3\_2006.txt, you would type:

```
wx1 <- read.table("wx_13_2006.txt", header=F, sep=",")</pre>
```





Often, your data will be in a spreadsheet.

If possible, export it as a .csv file and use something like the following to read it in.

```
wx2 <- read.table("wx_13_fwi_2006-2011.csv",
header=FALSE, sep=",")</pre>
```

If you cannot export to .csv, you can leave it as .xlsx and use the read.xslx() command in the xlsx package (Dragulescu and Arendt, 2018).





When reading in a file with columns separated by blanks with blank missing values, you can use code such as

```
dataset1 <- read.table("file1.txt", header=TRUE,
    sep=" ", na.string=" ")</pre>
```

This tells R that the blank spaces should be read in as missing values.



#### Observe the contents of dataset1:

```
dataset1
## x y z
## 1 3 4 NA
## 2 51 48 23
## 3 23 33 111
```

Note the appearance of NA.

This represents a *missing value*.

Functions such as is.na() are important for detecting missing values in vectors and data frames.

For more information about handling of missing values, check out the See Also section of help(is.na) and the *mice* package (van Buuren and Groothuis-Oudshoorn, 2011).





Sometimes, external software exports data files that are tab-separated. When reading in a file with columns separated by tabs with blank missing values, you could use code like

```
dataset2 <- read.table("file2.txt", header=TRUE,
    sep="\t", na.string=" ")</pre>
```

Again, observe the result:

```
dataset2
## x y z
## 1 33 223 NA
## 2 32 88 2
## 3 3 NA NA
```

If you need to skip the first 3 lines of a file to be read in, use the skip=3 argument.



## **Extracting information from data frames**

## To extract the height column from the women data frame, use the \$ operator:

```
women$height
## [1] 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72
```

### To extract the weight column from chickwts we use

```
## [1] 179 160 136 227 217 168 108 124 143 140 309

## [12] 229 181 141 260 203 148 169 213 257 244 271

## [23] 243 230 248 327 329 250 193 271 316 267 199

## [34] 171 158 248 423 340 392 339 341 226 320 295

## [45] 334 322 297 318 325 257 303 315 380 153 263

## [56] 242 206 344 258 368 390 379 260 404 318 352

## [67] 359 216 222 283 332
```





## If you want only the chicks who were fed horsebean, you can apply the subset() function to the chickwts data frame:

```
chickHorsebean <- subset(chickwts, feed == "horsebean")</pre>
chickHorsebean
##
      weight
             feed
## 1
         179 horsebean
## 2
         160 horsebean
## 3
         136 horsebean
## 4
         227 horsebean
## 5
         217 horsebean
## 6
         168 horsebean
## 7
         108 horsebean
## 8
         124 horsebean
## 9
         143 horsebean
## 10
         140 horsebean
```





You can now calculate the mean and standard deviation, and so on, of these weights:

```
mean(chickHorsebean$weight) # mean

## [1] 160.2

sd(chickHorsebean$weight) # standard deviation

## [1] 38.626
```





# In order to extract the 4th row from the chickHorsebean data frame, type

```
chickHorsebean[4, ]

## weight feed
## 4 227 horsebean
```

### To extract the element in the 2nd column of the 7th row of women, type

```
women[7, 2]
## [1] 132
```



# If we want the elements in the 4th through 7th row of the 2nd column of women, we can use

```
women[4:7, 2]
## [1] 123 126 129 132
```

## Note the use of the : operator:

```
4:7
## [1] 4 5 6 7
```





Another built-in data frame is airquality.

If we want to compute the mean for each of the first sapply (4 columns of this data ## Oz frame, we can use the ## sapply() function:

```
sapply(airquality[, 1:4], mean)
## Ozone Solar.R Wind Temp
## NA NA 9.9575 77.8824
```

The sapply() function applies the same function to all columns of the supplied data frame.

### **Factors**



Factors offer an alternative, often more efficient, way of storing character data.

For example, a factor with 6 elements and having the two levels, control and treatment can be created using:factor()

```
grp <- factor(grp)
grp

## [1] control treatment control treatment treatment
## [6] control
## Levels: control treatment</pre>
```



### Consider the built-in data frame *InsectSprays*

```
## count spray
## Min. : 0.0 A:12
## 1st Qu.: 3.0 B:12
## Median : 7.0 C:12
## Mean : 9.5 D:12
## 3rd Qu.:14.2 E:12
## Max. :26.0 F:12
```

The second column of this data frame is a *factor* representing the different types of spray used in the associated experiment.





The levels of a factor can be listed using the *levels()* function:

```
levels(InsectSprays$spray)
## [1] "A" "B" "C" "D" "E" "F"
```



Factors are a more efficient way of storing character data when there are repeats among the vector elements.

This is because the levels of a factor are internally coded as integers.

To see what the codes are for the spray factor, we can type

The labels for the levels are only stored once each, rather than being repeated.



We can change the labels for the factor using the *levels()* function as follows:

```
levels(InsectSprays$spray)[3] <- "Raid"</pre>
```

### Observe the effect of the change in

```
summary (InsectSprays$spray)
## A B Raid D E F
## 12 12 12 12 12 12
```

### **Factors**



The levels () function also offers a simple way to collapse categories.

Suppose we are interested in comparing the first three levels with the last three levels.

We can create a new factor for this purpose as follows:

### **Factors**



### **Check the result:**

```
## count spray newFactor
## Min. : 0.0 A :12 A:36
## 1st Qu.: 3.0 B :12 B:36
## Median : 7.0 Raid:12
## Mean : 9.5 D :12
## 3rd Qu.:14.2 E :12
## Max. :26.0 F :12
```

### References



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- 8. Stef van Buuren, Karin Groothuis-Oudshoorn (2011). mice: Multivariate Imputation by Chained Equations in R. Journal of Statistical Software, 45(3), 1-67. URL https://www.jstatsoft.org/v45/i03/.