# First steps in R

true

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This chapter is mostly aimed at people who are very new to R. However, people who do know R may still find useful insights from the sections where I emphasize how I use R.

### i Note

This is certainly not the best tutorial you'll ever find, so if you want other tutorials check out this curated list of R Tutorials here.

Furthermore, this tutorial teaches R the way I use it, which means you can do (and may have done) almost everything I do here with other code/functions/approaches. "The way I use it" mostly refers to me using the  $\{\text{tidyverse}\}$ , which "is an opinionated collection of R packages designed for data science. All packages share an underlying design philosophy, grammar, and data structures" and has become quite popular over the last years. Here is a direct comparison of how to do things in R with base R and via the tidyverse.

### 1 Let's go

You can use R like a calculator. It does not matter<sup>1</sup> whether you put spaces in or not as shown here:

```
2+3
[1] 5
2 * 6
[1] 12
```

Similar to what you may know from Excel, you can use functions like e.g. sqrt() to obtain the square root of a number:

```
sqrt(9)
```

You can find out more about a function by running the command with a question mark in front: ?sqrt().

Besides built-in functions, R also knows certain things like  $\pi$  or the alphabet, which are stored in the built-in constants named pi and letters:

```
pi
[1] 3.141593

letters

[1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k" "l" "m" "n" "o" "p" "q" "r" "s" [20] "t" "u" "v" "w" "x" "y" "z"
```

Importantly, you may also define your own variables via = or <-:

 $<sup>^{1}</sup>$ As long as you are dealing with numbers and calculations. Spaces do matter when we are talking about strings (i.e. text).

```
x = 3
x

[1] 3

x = 20
x

[1] 20

a_long_variable_name <- 2 + 4
a_long_variable_name

[1] 6

x + a_long_variable_name

[1] 26

mytext <- "This is my text"
mytext

[1] "This is my text"</pre>
```

Note that the variable x was overwritten - at first it was 3, but then it was 20. Also, as can be seen when the variable a\_long\_variable\_name is defined, you may put more than just a simple number on the right side of the <- or =.

## 2 data types

As you just saw, R can deal with both numbers and text. We can check the data type via the typeof() function:

```
typeof(x)
[1] "double"
```

```
typeof(mytext)
[1] "character"
```

Here is a simplified overview over some of R's data types you may see more often:

- Numbers
  - integer/int: whole number
    e.g. 42, -1504
    numeric/num & double/dbl: real number
- Text
  - character/chr: string values e.g. "hello", "Two words"

e.g. 3.14, 0.051795

- Factor
  - factor/fct: categorical variable that stores both string and integer data values as "levels"
     e.g. Control, Treatment
- TRUE/FALSE
  - logical/logi: logical value either TRUE or FALSE

#### 3 vectors

Instead of dealing with single numbers, we obviously want to deal with entire datasets. Before we get to an entire table with multiple rows and columns, the first step is to understand what a vector is in R: It is a sequence of elements that share the same data type. I often think about them as a single column in my dataset. Above, we actually already looked at a vector: letters is a built-in vector with 26 elements of the data type character. We could check this via the length() or str() functions:

```
length(letters)
[1] 26
str(letters)
```

```
 \texttt{chr} \ \ [1:26] \ \ "a" \ \ "b" \ \ "c" \ \ "d" \ \ "e" \ \ "f" \ \ "g" \ \ "h" \ \ "i" \ \ "j" \ \ "k" \ \ "l" \ \ "m" \ \ "n" \ \ "o" \ \ "p" \ \dots
```

Note how the built-in constant pi is not a vector, because it's just a single number (a.k.a. a scalar).

```
str(pi)
num 3.14
length(pi)
[1] 1
```

If you want to create your own vector from scratch, you must put all the elements together in the c() function and separate them with commas:

```
mynumbers <- c(1, 4, 9, 12, 12, 12, 16)
mynumbers

[1] 1 4 9 12 12 12 16

mywords <- c("Hakuna", "Matata", "Simba")
mywords

[1] "Hakuna" "Matata" "Simba"</pre>
```

Interestingly, we can still apply the sqrt() function we used above to a vector with numbers and it will simply take the square-root of every element:

```
sqrt(mynumbers)
[1] 1.000000 2.000000 3.000000 3.464102 3.464102 3.464102 4.000000
```

However, there are also functions like mean() which return the mean of all numbers in a vector as a single output element:

```
mean(mynumbers)
```

### 4 function arguments

So far, the functions we used had in common that they required only one input. The really good stuff in R happens with more complex functions which need multiple inputs. Let us use seq() as an example, which seems simple enough, because it generates a sequence of numbers:

```
seq(1, 10)
[1] 1 2 3 4 5 6 7 8 9 10
```

As you can see, putting in 1 and 10 separated by a comma generates a numeric vector with numbers from 1 to 10. However, I would like you to fully understand what is going on here, because it will help a lot with more complex functions.

See, we could switch the numbers and the function will work as expected:

```
seq(10, 1)
[1] 10 9 8 7 6 5 4 3 2 1
```

So this means that the first input is always the starting point and the second one is always the end point of the sequence, right? Well, yes by default, but you can have it your way if you specifically use the names of the arguments.

Looking at ?seq() it says seq(from = 1, to = 1, by = ...) so this seq(10, 1) is more explicitly this: seq(from = 1, to = 10, by = 1). Here is proof:

```
seq(from = 1, to = 10, by = 1)
[1] 1 2 3 4 5 6 7 8 9 10
```

Again, if you do not write out the arguments like this, it will simply assume the default order: The first number supplied is from = the second is to = and the third is by =. However, if we write out the arguments, we can use any order we like:

```
seq(1, 9, 2)
```

```
[1] 1 3 5 7 9

seq(from = 1, to = 9, by = 2)

[1] 1 3 5 7 9

seq(from = 1, by = 2, to = 9)

[1] 1 3 5 7 9

seq(1, 2, 9)

[1] 1
```

In short: If you understand why the first three lines of the code above produce the same result, but the last one does not, you are good to go!

### 5 R packages

Any function is always part of an R package.

#### 5.1 base R

After installing R there are many functions etc. you can use right away - which is what we did above. For example, when running ?mean, the help page will tell you that this function is part of the {base} package. As the name suggests, this package is built-in and its functions are ready to use the moment you have installed R. You can verify this by going to the "Packages" tab in RStudio - you will find the base package and it will have a checked box next to it.

#### 5.2 loading packages

When looking at the "Packages" tab in RStudio you may notice that some packages are listed, but do not have a check mark in the box next to them. These are packages that are installed, but not loaded. When a package is not loaded, its functions cannot be used. In order to load a package, the default command is library(package\_name). This command must be run once every time you open a new R session.

#### 5.3 installing additional packages

R really shines because of the ability to install additional packages from external sources. Basically, anyone can create a function, put it in a package and make it available online. Some packages are very sophisticated and popular - e.g. the package {ggplot2}, which is not built-in, has been downloaded 75 million times. In order to install a package, the default command is install.packages(package\_name). Alternatively, you can also click on the "Install" button in the top left of the "Packages" tab and type in the package\_name there.

```
A package only needs to be installed once, but
A package must be loaded every time you open a new R session!
```

Here is a curated list of R packages and tools for different areas.

### 5.4 p\_load() vs. library()

You now know how to install and load R packages the standard way. However, over the years I switched to using the function p\_load() from the {pacman} package instead of library() and install.packages(). The reason is simple: Usually R-scripts start with multiple lines of library() statements that load the necessary packages. However, when this code is run on a different computer, the user may not have all these packages installed and will therefore get an error message. This can be avoided by using the p\_load(), because it

- loads all packages that are installed and
- installs and loads all packages that are not installed.

Obviously, {pacman} itself must first be installed (the standard way). Moreover, you may now think that in order to use p\_load() we do need a single library(pacman) first. However, we can avoid this by writing pacman::p\_load() instead. Simply put, writing package\_name::function\_name() makes sure that this explicit function from this explicit package is being used. Additionally, R actually lets you use this function without loading the corresponding package. Thus, we now arrived at the way I handle packages at the beginning of all my R-scripts:

```
pacman::p_load(
   package_name_1,
   package_name_2,
   package_name_3
)
```

## 6 Keyboard shortcuts

Here are shortcuts I actually use regularly in RStudio:

Shortcut	Description		
CTRL+ENTER	Run selected lines of code		
CTRL+SHIFT+M	Insert $\%$ $\%$ <sup>2</sup>		
CTRL+SHIFT+R	Insert code section header		
CTRL+LEFT/RIGHT	Jump to Word		
CTRL+SHIFT+LEFT/RIGHT Select Word			
ALT+LEFT/RIGHT	Jump to Line Start/End		
ALT+SHIFT+LEFT/RIGHT	Select to Line Start/End		
CTRL+A	Highlight everything (to run the entire code)		
CTRL+Z	Undo		

Keyboard shortcuts can be customized in RStudio as described here.

 $<sup>^2</sup>$ The pipe operator %>% is introduced in the next chapter.