

First steps with R

Probabilités et statistique pour la biologie (STAT1)

Jacques van Helden

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Goal of this tutorial

This tutorial aims at discovering the fundamental elements of the **R** statistical language.

We will briefly survey the following concepts.

- Handling variables
 - Assigning a value to a variable
 - Basic operations on numbers

- Basic data structures
 - * vectors
 - * matrices
 - * data.frames
 - * lists
- Using functions
- Graph drawing
- Distributions of probabilities

Using *R* as a calculator

Example: an addition with R.

At the R prompt, type the following instruction and press the Enter key.

```
2 + 5
```

```
[1] 7
```

The result (7) of the addition is printed out, preceded by an index [1] (we will explain later why this index appears).

Assigning a value to a variable

In *R*, the succession of a the hyphen and “smaller than” characters (`<-`) serves to assign a value to a variable. If the variable does not exist yet, it is created.

For example

```
a <- 2
```

creates a variable named *a*, and assigns it the value 2.

The result can be displayed with the `print()` function.

```
print(a)
```

```
[1] 2
```

Remark: R also allows to use the equal symbol (=) to assign a value to a variable. However, we prefer to use the original assignation (`<-`), to follow the R style recommendations.

Naming and syntactic conventions in R

A priori, are several conventions can be envisaged to ensure a consistent naming of variables, functions, operators, etc.

For each programming language, the community of programmers defines some standard(s) to ensure a consistency of the published code.

For this course, we will follow the recommendations of the **Google R style guide**:

<https://google.github.io/styleguide/Rguide.xml>

However, for variable identifiers, the traditional notation `variable.name` raises some issues for programmers who are familiar with object-oriented languages (e.g. java, python), where the point serves to apply a method (that follows the point) to an object (that precedes the point).

To avoid this confusion, we will use the alternative so-called *camel back* notation (e.g. `variableName`). Attention, according to this convention, variable names always start with a lower case, whereas function / method names start with an uppercase.

Computing with variables

- Create a variable named b with value 5
- Compute $a + b$ and store the result in a variable named c
- Print the result

```
b <- 5
c <- a + b
print(a)
```

```
[1] 2
```

```
print(b)
```

```
[1] 5
```

```
print(c)
```

```
[1] 7
```

Assignment \neq equality

Exercise

- Replace the value of a by 3
- Print out the value of c
- Is-it still true that $c = a + b$? Why?

Solution

```
a <- 3 ## Change the value of a
print(a)
```

```
[1] 3
```

```
print(b)
```

```
[1] 5
```

```
print(c)
```

```
[1] 7
```

```
## Check whether c equals a + b
c == a + b
```

```
[1] FALSE
```

Interpretation: `==` tests whether two variables have the same content. The result is a logical value (TRUE or FALSE).

Recomputing a result

When the content of a given variable a is changed, another variable (c) previously computed from it has no reason to be recomputed if not explicitly requested.

Example:

- Replace the value of a by 27,
- Recompute the value of c
- Test the equality $c = a + b$

```
a <- 27 ## Change the value of a
c <- a + b
print(c) ## Print the value of c
```

```
[1] 32
```

```
## Check whether c equals a + b
c == a + b
```

```
[1] TRUE
```

Vectors of values

In **R**, the simplest data structure is a **vector**.

- In the previous example, the variable a contained a single number, but in practice it was stored in a single-entry vector.
- The R function `print()` displays the indices at the beginning of each row. This is useful when displaying a vector with a large number of entries.

Example: create a variable named *threeNumbers*, and initialise it with a vector containing the values *27*, *12* and *3000*.

Tips:

- the function `c()` combines several values into a vector.

```
threeNumbers <- c(27,12,3000)
print(threeNumbers)
```

```
[1] 27 12 3000
```

Series of numbers

The simplest way to create a series of number is to use the column character `:`, which generates all integer values between two boundaries;

```
x <- 0:30
print(x)
```

```
[1] 0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
[24] 23 24 25 26 27 28 29 30
```

Note: if the printout of the values extends beyond the width of the console, R goes to the next row but displays between square brackets the index of the first element at the beginning of the new row.

Another example

```
print(58:157)
```

```
[1] 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74
[18] 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90 91
[35] 92 93 94 95 96 97 98 99 100 101 102 103 104 105 106 107 108
[52] 109 110 111 112 113 114 115 116 117 118 119 120 121 122 123 124 125
[69] 126 127 128 129 130 131 132 133 134 135 136 137 138 139 140 141 142
[86] 143 144 145 146 147 148 149 150 151 152 153 154 155 156 157
```

Computing with vectors

R enables to handle vectors in a every practical ways: mathematical operations involving a vector automatically apply to all its elements.

```
x <- 1:10 # Define a series from 1 to 10
print(x)
```

```
[1] 1 2 3 4 5 6 7 8 9 10
```

```
y <- x^2 # Compute the square of each number
print(y)
```

```
[1] 1 4 9 16 25 36 49 64 81 100
```

Sequences of numbers

The function `seq()` enables to generate series of bumbers separated by an arbitrary interval.

```
seq(from=-1, to=1, by=0.1)
```

```
[1] -1.0 -0.9 -0.8 -0.7 -0.6 -0.5 -0.4 -0.3 -0.2 -0.1 0.0 0.1 0.2 0.3
[15] 0.4 0.5 0.6 0.7 0.8 0.9 1.0
```

Variables can also contain text

Variables are not restricted to numbers: they can contain text (strings of characters).

We will now use the function `c()` to combine severak character strings into a vector.

```
# The # symbol allows to insert comments in R code
```

```
# Define a vector named "whoami", and
# containing two names
whoami <- c("Denis", "Siméon")
print(whoami) # Comment at the end of a line
```

```
[1] "Denis" "Siméon"
```

Sring concatenation

The function `paste()` enables to concatenate string-containing variables.

```
# To concatenate the elements of a vector in a single chain, use "collapse"
firstName <- paste(collapse = " ", whoami)
print(firstName)
```

```
[1] "Denis Siméon"
```

```
# To concatenate two vectors, use "sep"
lastName <- "Poisson"
print(paste(sep = " ", firstName, lastName))
```

```
[1] "Denis Siméon Poisson"
```

```
## Concatenate 2 vectors with 3 values each
firstNames <- c("George", "Alfred", "Frédéric")
lastNames <- c("Sand", "Musset", "Chopin")
fullNames <- paste(sep = " ", firstNames, lastNames)
print(fullNames)
```

```
[1] "George Sand"      "Alfred Musset"    "Frédéric Chopin"
```

Note that the `paste()` functions can also be used to concatenate all the values of a given vector, but this requires to use the `collapse` argument instead of `sep`.

```
paste(fullNames, collapse = ", ")
```

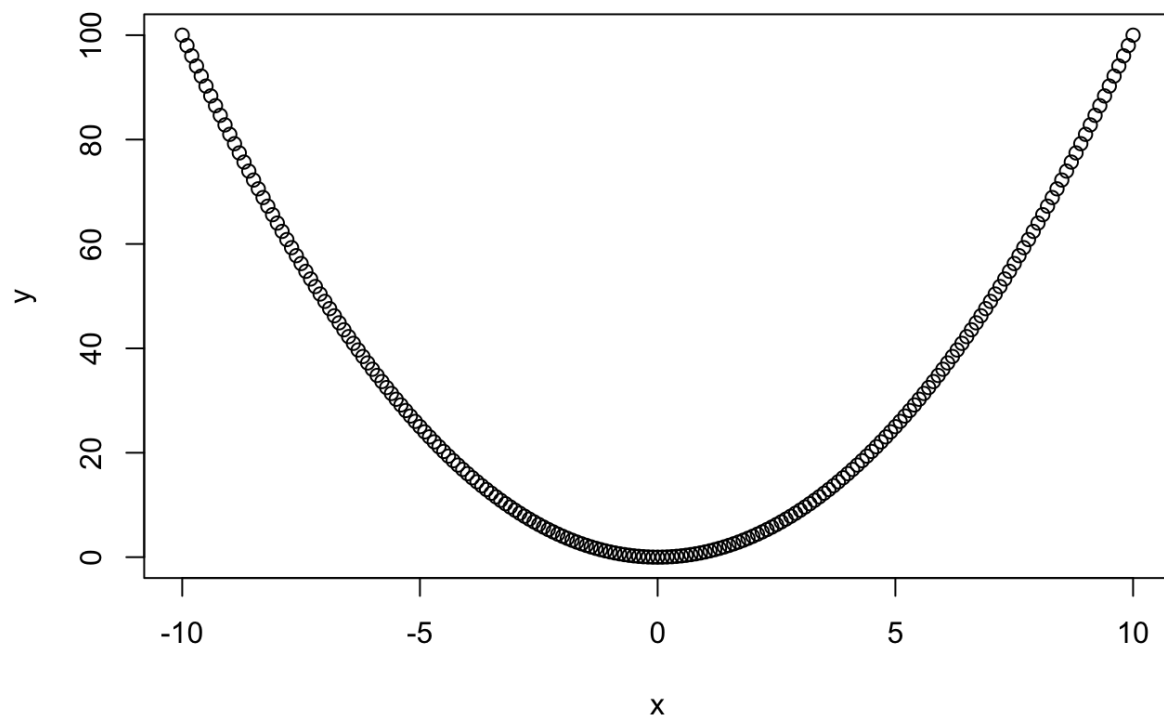
```
[1] "George Sand, Alfred Musset, Frédéric Chopin"
```

Graphical functions

R includes a large number of functions enabling to draw simple or elaborate graphics. We explore hereafter the simplest methods.

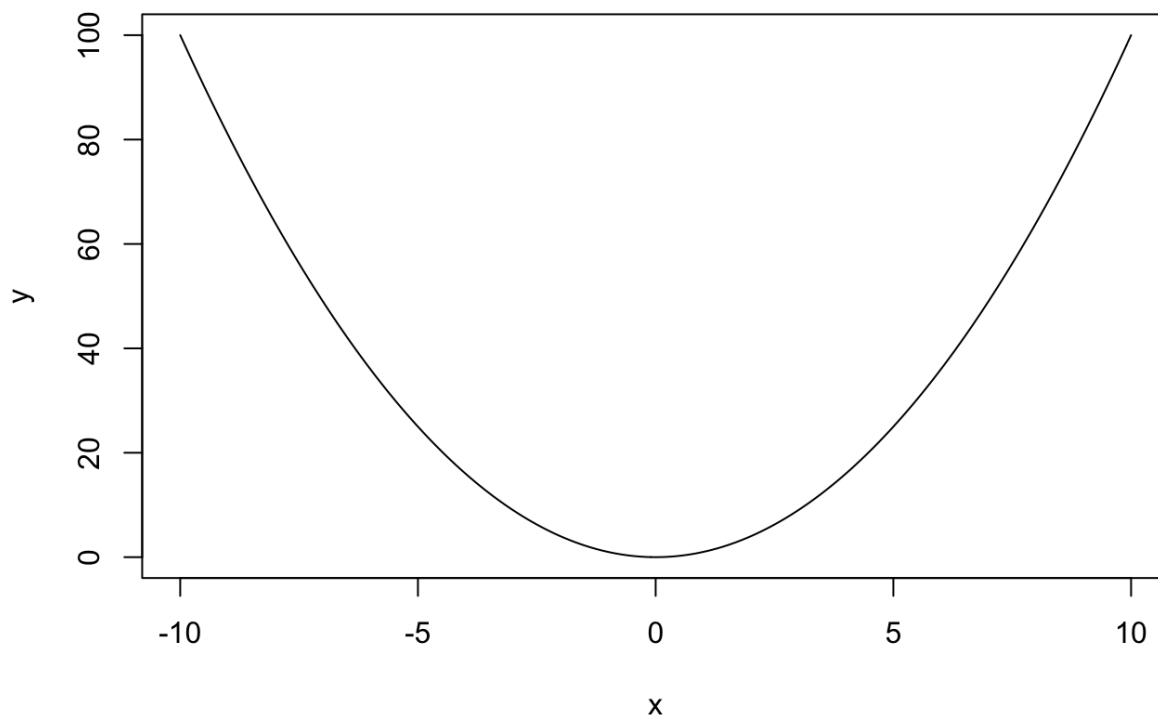
Scatter plot

```
x <- seq(from = -10, to = 10, by = 0.1)
y <- x^2
plot(x,y)
```



Curve (line plot)

```
x <- seq(from = -10, to = 10, by = 0.1)
y <- x^2
plot(x,y, type="l")
```



Improving a graphics

Exercise: on the R console, type `help(plot)`, read the help of the `plot()` function, and explore the parameters in order to improve the previous graphics. Also consult the help, function for the graphical parameters (`help(par)`).

You could for example, attempt to add the following elements to the figure:

- title
- axis labels
- line color
- line width
- grid
- trace an horizontal line to mark the Y axis at coordinate $X = 0$
- trace a vertical line to mark the X axis at coordinate $Y = 0$
- any other parameter that will improve the readability / interpretability of the resulting Figure

An example of solution is shown in the next section (don't look before having done the exercise!).

Solution

Run the following chunk of code to generate the improved figure.

```
x <- seq(from = -10, to = 10, by = 0.1)
y <- x^2
```



```

plot(x,y,
      type="l", # Plot type
      main = "Parabole", # Main title
      xlab = "x", #X label
      ylab = "y = x^2", # Y label
      col = "blue", # Curve color
      lwd = 3, # Line width
      las = 1 # display axis labels horizontally
    )
grid(lty = "dashed", col="gray") # Grid
abline(h = 0) # Horizontal line
abline(v = 0) # Vertical line

```

Probability distributions

For each one of the classical distribution of probabilities, **R** provides 4 functions-.

Before going any further, read carefully the help for the functions associated to the binomial distribution.

```
help("Binomial")
```

Questions:

1. Which R function enables to compute the probability mass function (also called “density” for convenience)?
2. Which R function corresponds to the cumulative distribution function (CDF)?
3. What is the role of the function *rbinom()*?

Drawing the binomial distribution

Exercise: assuming a DNA sequence with equiprobable nucleotides, draw the expected distribution for the number of Adenin (A) residues in an oligonucleotide of length 30 (count the adenines on a single strand).

- probability mass function
- cumulative distribution

In the forthcoming slides, we will guide you step by step to start the exercise, and you will then be invited to improve the result at your will.

Formula of the solution

The number of adenins can take any value between 0 and 30. The problem can be modeled as Bernoulli Schema with $n = 30$ trials, which can each result in either a success (observing an adenin, with $p = 0.25$) or a failure (any other nucleotide, with probability $q = 1 - p = 0.75$). The probability to observe exactly x adenins is thus:

$$P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x} = \frac{30!}{x!(30 - x)!} \cdot 0.25^x \cdot 0.75^{n-x}$$

where x can take any value between 0 and 30.

Computation of the probability mass function

```
## Define all possible values for X
n <- 30
x <- 0:n
p <- 0.25

## Compute the binomial PMF
pmf <- dbinom(x = x, size = n, prob = p)
```

Selecting a subset of entries from a vector

We can select the 4 first values of the `pmf` variable defined above (which correspond to number of successes x from 0 to 4) ...

```
pmf[1:5]
```

```
[1] 0.0001785821 0.0017858209 0.0086314677 0.0268534550 0.0604202738
```

... or the 4 last values (x values from 27 to 30 successes).

```
pmf[(n-3):n]
```

```
[1] 1.925374e-12 9.508019e-14 3.395721e-15 7.806256e-17
```

Restricting the number of decimals

The function `round()` rounds a result with a given number of digits.

```
round(pmf, digit=3)
```

```
[1] 0.000 0.002 0.009 0.027 0.060 0.105 0.145 0.166 0.159 0.130 0.091
[12] 0.055 0.029 0.013 0.005 0.002 0.001 0.000 0.000 0.000 0.000 0.000
[23] 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000
```

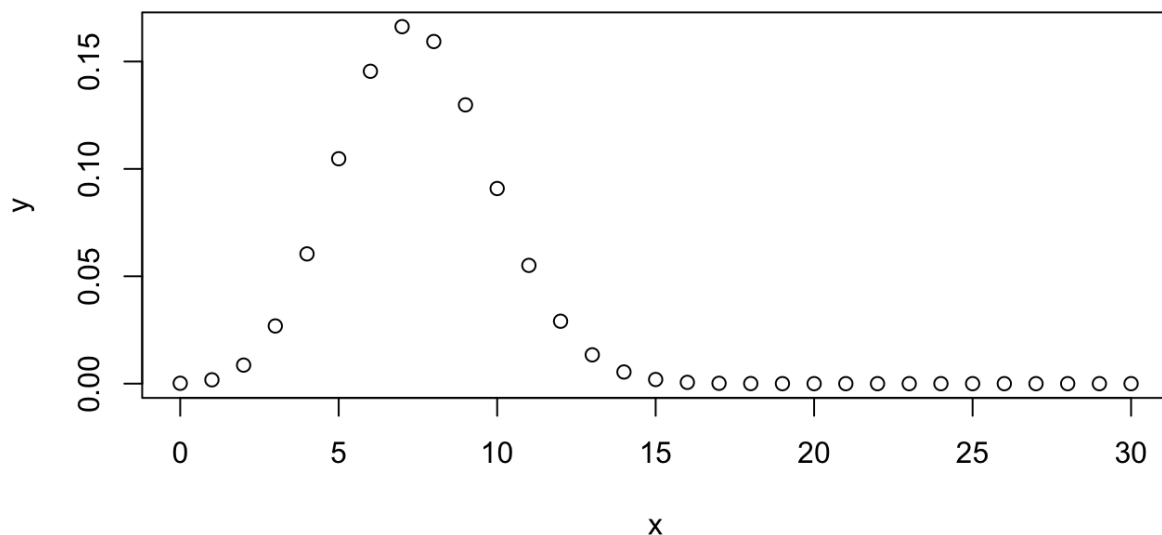
However, this is not always the best solution. Indeed, probability values can rapidly reach very small values, and we would thus not like to replace all these values by 0 if they are lower than the number of decimals. We would like to see both the order of magnitude, and to show the value with a reasonable number of significant digits. For this, we can use the `signif()` function.

```
signif(pmf, digit=3)
```

```
[1] 1.79e-04 1.79e-03 8.63e-03 2.69e-02 6.04e-02 1.05e-01 1.45e-01
[8] 1.66e-01 1.59e-01 1.30e-01 9.09e-02 5.51e-02 2.91e-02 1.34e-02
[15] 5.43e-03 1.93e-03 6.03e-04 1.66e-04 3.99e-05 8.39e-06 1.54e-06
[22] 2.44e-07 3.33e-08 3.86e-09 3.75e-10 3.00e-11 1.93e-12 9.51e-14
[29] 3.40e-15 7.81e-17 8.67e-19
```

Drawing the binomial distribution

```
n <- 30; x <- 0:n # Define the X values from 0 to 14
y <- dbinom(x = x, size = n, prob = 0.25) # Poisson density
plot(x,y) # Check the result
```



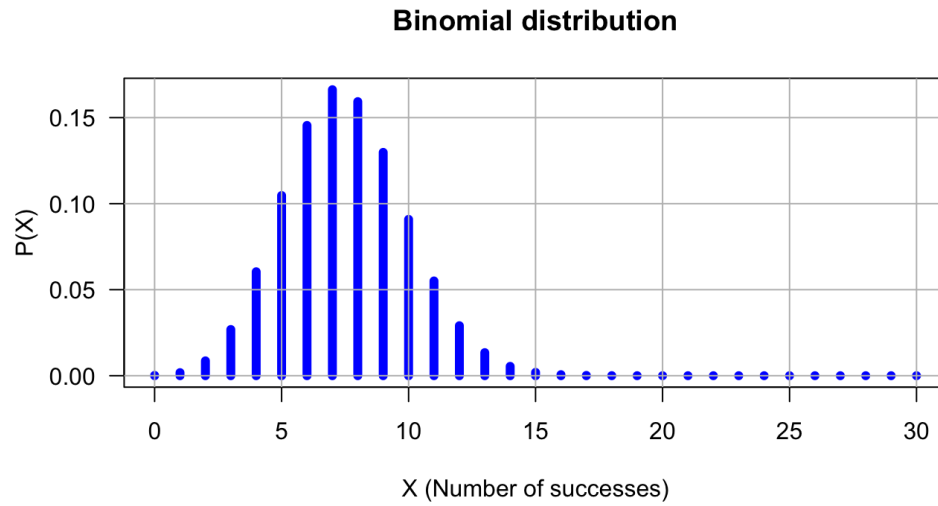
This first drawing is not very elegant. The empty circles do not make it easy to perceive the general shape of the distribution. We can attempt to improve its interpretability.

Exercise 1: improving the drawing of the binomial

Use the following options of the `plot()` function to highlight the shape of the distribution. To give you an idea of the target, the expected result is displayed on the next slide.

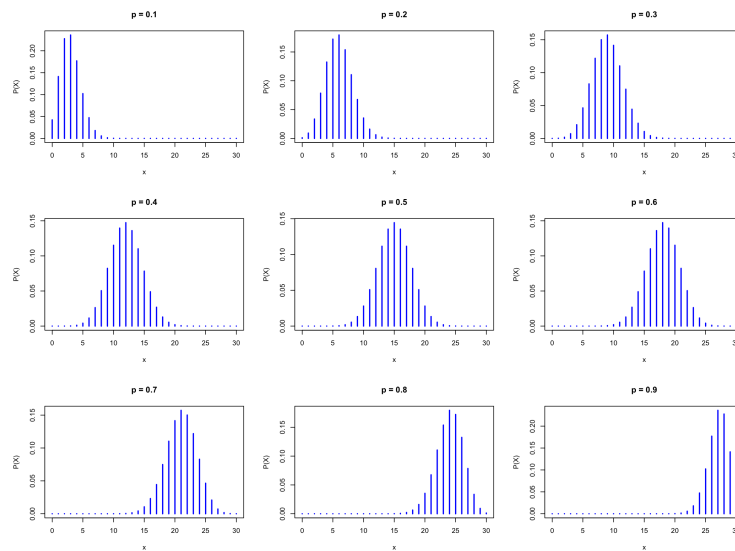
- Choose a type of points (option `type`) that reflects the height of the probability (valeur Y) at the different values of X .
- Add a title (option `main`) and specify relevant axis legends (options `xlab` and `ylab`)
- Color the drawing (option `col`)
- Thicken the lines (option `lwd`)
- Add an horizontal grid (function `grid()`)
- make sure that the axis labels are horizontal (option `las`).

Expected result

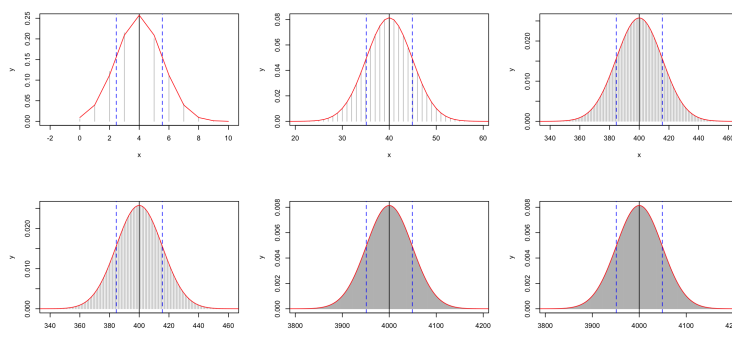


Exercise: series of binomial curves

Draw a series of binomial curves with $n = 30$ trials, and probabilities ranging from $p = 0.1$ to $p = 0.9$ by steps of 0.1.



Exercise: convergence of the binomial to the normal



Before finishing: keep a trace of your session!

Traceability and reproducibility of the analyses are crucial issues for sciences. The **R** function `sessionInfo()` summarises the software environment used for the current working session: R version, operating system, loaded libraries/packages and their versions.

```
sessionInfo()
```

```
R version 3.6.1 (2019-07-05)
```

```
Platform: x86_64-apple-darwin15.6.0 (64-bit)
```

```
Running under: macOS High Sierra 10.13.6
```

```
Matrix products: default
```

```
BLAS: /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRblas.0.dylib
```

```
LAPACK: /Library/Frameworks/R.framework/Versions/3.6/Resources/lib/libRlapack.dylib
```

```
locale:
```

```
[1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
```

```
attached base packages:
```

```
[1] stats      graphics  grDevices  utils      datasets  methods    base
```

```
other attached packages:
```

```
[1] knitr_1.23
```

```
loaded via a namespace (and not attached):
```

```
[1] compiler_3.6.1  magrittr_1.5    tools_3.6.1     htmltools_0.3.6
```

```
[5] yaml_2.2.0      Rcpp_1.0.2      stringi_1.4.3   rmarkdown_1.14
```

```
[9] stringr_1.4.0   xfun_0.8        digest_0.6.20   evaluate_0.14
```