

Programming with R

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My first function

The best way:

```
my_function <- function(input1, input2) {  
  output <- input1 + input2  
  return(output)  
}
```

```
my_function(input1 = 1, input2 = 3)
```

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

My first function

The best way:

```
my_function <- function(input1, input2) {  
  output <- input1 + input2  
  return(output)  
}  
  
my_function(input1 = 1, input2 = 3)  
## [1] 4
```

If no return(), then it returns the last row:

```
my_function <- function(input1, input2) {  
  input1 + input2  
}  
  
my_function(input1 = 1, input2 = 3)  
## [1] 4
```

My first function

The best way:

```
my_function <- function(input1, input2) {
  output <- input1 + input2
  return(output)
}

my_function(input1 = 1, input2 = 3)
## [1] 4
```

If no return(), then it returns the last row:

```
my_function <- function(input1, input2) {
  input1 + input2
}

my_function(input1 = 1, input2 = 3)
## [1] 4
```

Inline shortcut (usefull in *apply()):

```
my_function <- function(input1, input2) input1 + input2

my_function(input1 = 1, input2 = 3)
## [1] 4
```

My first function

You can set defaults for your inputs:

```
my_function <- function(input1, input2 = 10) {  
  output <- input1 + input2  
  return(output)  
}  
  
my_function(input1 = 1)  
## [1] 11  
  
my_function(input1 = 1, input2 = 0)  
## [1] 1  
  
my_function(input2 = 1)  
## Error in my_function(input2 = 1): argument "input1" is missing, with no default
```

My first function

You can pass optional arguments to another function via "...":

```
my_function <- function(input1, ...) {  
  output <- mean(input1, ...)  
  return(output)  
}  
  
x <- c(1, 2, 3, 4, 5, NA)  
my_function(input1 = x)  
## [1] NA  
  
my_function(input1 = x, na.rm = TRUE)  
## [1] 3
```

Note: this is particularly usefull when designing plotting functions!

My first function

You can print things while the function runs:

```
my_function <- function() {  
  print("This function will output 2")  
  return(2)  
}  
  
my_function()  
## [1] "This function will output 2"  
## [1] 2
```

Note: observe also that it is possible to create functions that do not consider any input!

My first function

You can return only a single object:

```
my_function <- function(input1, ...) {  
  output1 <- min(input1, ...)  
  output2 <- max(input1, ...)  
  output <- list(output1 = output1, output2 = output2)  
  return(output)  
}  
  
x <- c(1, 2, 3, 4, 5, NA)  
my_function(input1 = x, na.rm = TRUE)  
  
## $output1  
## [1] 1  
##  
## $output2  
## [1] 5
```

So if you need several outputs, you must combine them (as vector, matrix, dataframe, list...)!

My first function

You can return return things “invisibly”:

```
my_function <- function(input1) {  
  output <- min(input1)  
  return(invisible(output))  
}  
  
my_function(input1 = c(1, 2, 3, -4))
```

My first function

You can return return things “invisibly”:

```
my_function <- function(input1) {  
  output <- min(input1)  
  return(invisible(output))  
}  
  
my_function(input1 = c(1, 2, 3, -4))
```

```
foo <- my_function(input1 = c(1, 2, 3, -4))  
foo  
## [1] -4
```

My first function

You don't have to return something:

```
my_function <- function(input1) {  
  the_min <- min(input1)  
  print(paste("The minimum is:", the_min))  
  return(invisible(NULL))  
}  
  
my_function(input1 = c(1, 2, 3, -4))  
## [1] "The minimum is: -4"
```

My first function

You don't have to return something:

```
my_function <- function(input1) {  
  the_min <- min(input1)  
  print(paste("The minimum is:", the_min))  
  return(invisible(NULL))  
}  
  
my_function(input1 = c(1, 2, 3, -4))  
## [1] "The minimum is: -4"
```

```
foo <- my_function(input1 = c(1, 2, 3, -4))  
## [1] "The minimum is: -4"  
  
foo  
## NULL
```

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Why writing your own R functions?

Using your own functions makes your scripts

- easier to understand
- safer to (re)use
- shorter to write (often)

What do you prefer?

```
d <- data.frame(proba = c(0.1, 0.5, 0.4), group = factor(c("A", "B", "C")))

with(data = d, (proba[group == "B"] / (1 - proba[group == "B"])) / (proba[group == "A"] / (1 - proba[group == "A"])))
## [1] 9

with(data = d, (proba[group == "C"] / (1 - proba[group == "C"])) / (proba[group == "A"] / (1 - proba[group == "A"])))
## [1] 6

with(data = d, (proba[group == "B"] / (1 - proba[group == "B"])) / (proba[group == "C"] / (1 - proba[group == "C"])))
## [1] 1.5
```

What do you prefer?

```
d <- data.frame(proba = c(0.1, 0.5, 0.4), group = factor(c("A", "B", "C")))

with(data = d, (proba[group == "B"] / (1 - proba[group == "B"])) / (proba[group == "A"] / (1 - proba[group == "A"])))
## [1] 9

with(data = d, (proba[group == "C"] / (1 - proba[group == "C"])) / (proba[group == "A"] / (1 - proba[group == "A"])))
## [1] 6

with(data = d, (proba[group == "B"] / (1 - proba[group == "B"])) / (proba[group == "C"] / (1 - proba[group == "C"])))
## [1] 1.5
```

Or

```
odds_ratio <- function(group1, group2){
  with(data = d, (proba[group == group1] / (1 - proba[group == group1])) / (proba[group == group2] / (1 - proba[group == group2])))
}

odds_ratio(group1 = "B", group2 = "A")
## [1] 9

odds_ratio(group1 = "C", group2 = "A")
## [1] 6

odds_ratio(group1 = "B", group2 = "C")
## [1] 1.5
```

What do you prefer?

Still not convinced? Let's compute all pairwise comparisons:

```
with(data = d, (proba[group == "A"] / (1 - proba[group == "A"])) / (proba[group == "A"] / (1 - proba[group == "A"])))
## [1] 1

with(data = d, (proba[group == "B"] / (1 - proba[group == "B"])) / (proba[group == "A"] / (1 - proba[group == "A"])))
## [1] 9

with(data = d, (proba[group == "C"] / (1 - proba[group == "C"])) / (proba[group == "A"] / (1 - proba[group == "A"])))
## [1] 6

with(data = d, (proba[group == "A"] / (1 - proba[group == "A"])) / (proba[group == "B"] / (1 - proba[group == "B"])))
## [1] 0.1111111

with(data = d, (proba[group == "B"] / (1 - proba[group == "B"])) / (proba[group == "B"] / (1 - proba[group == "B"])))
## [1] 1

with(data = d, (proba[group == "C"] / (1 - proba[group == "C"])) / (proba[group == "B"] / (1 - proba[group == "B"])))
## [1] 0.6666667

with(data = d, (proba[group == "A"] / (1 - proba[group == "A"])) / (proba[group == "C"] / (1 - proba[group == "C"])))
## [1] 0.1666667

with(data = d, (proba[group == "B"] / (1 - proba[group == "B"])) / (proba[group == "C"] / (1 - proba[group == "C"])))
## [1] 1.5

with(data = d, (proba[group == "C"] / (1 - proba[group == "C"])) / (proba[group == "C"] / (1 - proba[group == "C"])))
## [1] 1
```

What do you prefer?

Still not convinced? Let's compute all pairwise comparisons:

```
for (group2 in d$group) {  
  for (group1 in d$group) {  
    print(paste(group1, group2, odds_ratio(group1 = group1, group2 = group2)))  
  }  
}  
  
## [1] "A A 1"  
## [1] "B A 9"  
## [1] "C A 6"  
## [1] "A B 0.111111111111111"  
## [1] "B B 1"  
## [1] "C B 0.666666666666667"  
## [1] "A C 0.166666666666667"  
## [1] "B C 1.5"  
## [1] "C C 1"
```

Note: this is bad code, we will come back on this!

When to write your own functions?

Don't Repeat Yourself

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Control flow: if()

```
i <- 1
a <- 2

if (i == 1) {
  a <- 1
}

a

## [1] 1
```

Control flow: if()

```
i <- 1
a <- 2

if (i == 1) {
  a <- 1
}

a

## [1] 1
```

```
i <- 5
a <- 2

if (i == 1) {
  a <- 1
} else {
  a <- 2
}

a

## [1] 2
```


Control flow: for()

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

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

Control flow: while()

```
i <- 1

while (i < 5) {
  print(i)
  i <- i + 1
}

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

Random number generators

```
runif(5)
## [1] 0.96463192 0.49758115 0.55721983 0.04147941 0.29552567
runif(5)
## [1] 0.3423236 0.2061392 0.6612697 0.6651510 0.8388332
```

Random number generators

```
runif(5)
## [1] 0.96463192 0.49758115 0.55721983 0.04147941 0.29552567
runif(5)
## [1] 0.3423236 0.2061392 0.6612697 0.6651510 0.8388332
```

You can have reproducible results by setting a seed:

```
set.seed(10132)
runif(5)
## [1] 0.7258731 0.3086039 0.4393603 0.7952702 0.5325773
set.seed(10132)
runif(5)
## [1] 0.7258731 0.3086039 0.4393603 0.7952702 0.5325773
```

Random number generators

```
runif(5)
## [1] 0.96463192 0.49758115 0.55721983 0.04147941 0.29552567
runif(5)
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You can have reproducible results by setting a seed:

```
set.seed(10132)
runif(5)
## [1] 0.7258731 0.3086039 0.4393603 0.7952702 0.5325773
set.seed(10132)
runif(5)
## [1] 0.7258731 0.3086039 0.4393603 0.7952702 0.5325773
```

Note: check `?Distributions` for more distributions.

Errors and Warnings

```
process_factor <- function(x) {  
  if (!is.factor(x) & !is.character(x)) {  
    stop("Your input for factor is not a factor")  
  }  
  if (!is.factor(x) & is.character(x)) {  
    x <- as.factor(x)  
    warning("Your input has been converted to factor")  
  }  
  return(x)  
}
```

Errors and Warnings

```
process_factor <- function(x) {  
  if (!is.factor(x) & !is.character(x)) {  
    stop("Your input for factor is not a factor")  
  }  
  if (!is.factor(x) & is.character(x)) {  
    x <- as.factor(x)  
    warning("Your input has been converted to factor")  
  }  
  return(x)  
}
```

```
process_factor(x = factor(c("A", "B")))  
## [1] A B  
## Levels: A B
```

Errors and Warnings

```
process_factor <- function(x) {
  if (!is.factor(x) & !is.character(x)) {
    stop("Your input for factor is not a factor")
  }
  if (!is.factor(x) & is.character(x)) {
    x <- as.factor(x)
    warning("Your input has been converted to factor")
  }
  return(x)
}
```

```
process_factor(x = factor(c("A", "B")))
## [1] A B
## Levels: A B
```

```
process_factor(x = c("A", "B"))
## Warning in process_factor(x = c("A", "B")): Your input has been converted to factor
## [1] A B
## Levels: A B
```


Errors and Warnings

```
process_factor <- function(x) {
  if (!is.factor(x) & !is.character(x)) {
    stop("Your input for factor is not a factor")
  }
  if (!is.factor(x) & is.character(x)) {
    x <- as.factor(x)
    warning("Your input has been converted to factor")
  }
  return(x)
}
```

```
process_factor(x = factor(c("A", "B")))
## [1] A B
## Levels: A B
```

```
process_factor(x = c("A", "B"))
## Warning in process_factor(x = c("A", "B")): Your input has been converted to factor
## [1] A B
## Levels: A B
```

```
process_factor(x = c(2, 3))
## Error in process_factor(x = c(2, 3)): Your input for factor is not a factor
```

Scope

```
i <- 1
i <- i + 1
i
## [1] 2

e1 <- environment()
environmentName(e1)
## [1] "R_GlobalEnv"

get("i", envir = e1)
## [1] 2
```

Note: every object in R belongs to an environment!

Scope

```
i <- 1

f <- function(i) {
  i <- i + 1
  e <- environment()
  return(e)
}

e2 <- f(i)
i
## [1] 1
e2
## <environment: 0x55749b7a2670>
get("i", envir = e1)
## [1] 1
get("i", envir = e2)
## [1] 2
```

As a rule, anything created inside a function and not exported stays inside the function!

Scope

There are a few exceptions:

```
i <- 1

for (j in 1:10) {
  i <- i + 1
}

i
## [1] 11
```

And yet, for is actually a function:

```
i <- 1

`for`(j, 1:10, i <- i + 1)
i
## [1] 11
```

R has the same numerical issues as most programming languages

```
x <- 0.7 - 0.4 - 0.3  
x == 0  
## [1] FALSE
```

R has the same numerical issues as most programming languages

```
x <- 0.7 - 0.4 - 0.3
```

```
x == 0
```

```
## [1] FALSE
```

```
print(x, digits = 22)
```

```
## [1] -5.551115123125782702118e-17
```

R has the same numerical issues as most programming languages

```
x <- 0.7 - 0.4 - 0.3
```

```
x == 0
```

```
## [1] FALSE
```

```
print(x, digits = 22)
```

```
## [1] -5.551115123125782702118e-17
```

```
print(seq(0, 1, 0.1), digits = 22)
```

```
## [1] 0.00000000000000000000 0.10000000000000000055511
```

```
## [3] 0.200000000000000000111022 0.3000000000000000444089
```

```
## [5] 0.40000000000000000222045 0.500000000000000000000
```

```
## [7] 0.60000000000000000888178 0.7000000000000000666134
```

```
## [9] 0.80000000000000000444089 0.9000000000000000222045
```

```
## [11] 1.000000000000000000000
```

NB: same kind of thing can happen in Excel too (<https://support.microsoft.com/en-us/kb/214118>)

R has the same numerical issues as most programming languages

```
??"equality"
```

Help files with alias or concept or title matching 'equality' using fuzzy matching:

FactoMineR::prefpls	Scatter plot and additional variables with quality of representation contour lines
base::all.equal	Test if Two Objects are (Nearly) Equal
base::identical	Test Objects for Exact Equality
datasets::airquality	New York Air Quality Measurements

```
?all.equal
all.equal(target = 0, current = x)
## [1] TRUE
```


R has the same numerical issues as most programming languages

J.M Muller's Serie: $u_0 = 2$; $u_1 = -4$; $u_{n+1} = 111 - \frac{1130}{u_n} + \frac{3000}{u_n * u_{n-1}}$

```
u <- c(2, -4)
new.u <- function(u) 111 - 1130/u[length(u)] + 3000/(u[length(u)]*u[length(u) - 1])
for (i in 1:40) u <- c(u, new.u(u))
```

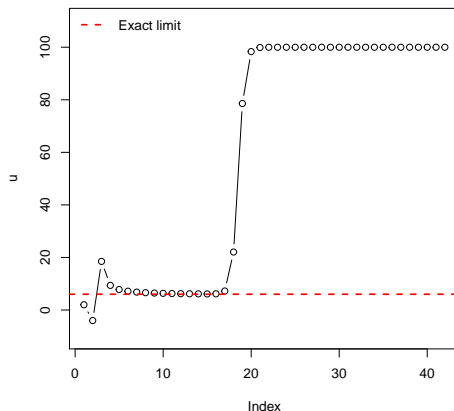


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Some key advices

- everything you use in the body must pass through the inputs
- everything you output must pass through the return
- try to write functions that you could reuse in other situations

Bad:

```
d <- data.frame(proba = c(0.1, 0.5, 0.4), group = factor(c("A", "B", "C")))

odds_ratio <- function(group1, group2){
  with(data = d, (proba[group == group1] / (1 - proba[group == group1])) / (proba[group == group2] / (1 - proba[group == group2])))
}

for (group2 in d$group) {
  for (group1 in d$group) {
    print(paste(group1, group2, odds_ratio(group1 = group1, group2 = group2)))
  }
}
```

Why is this bad?

A better implementation of functions to compute odd ratios

Better because the function does not rely on d:

```
odds_ratio <- function(group1, group2, data){
  p1 <- data$proba[data$group == group1]
  p2 <- data$proba[data$group == group2]
  OR <- (p1/(1 - p1)) / (p2/(1 - p2))
  return(OR)
}

for (group2 in d$group) {
  for (group1 in d$group) {
    print(paste(group1, group2, odds_ratio(group1 = group1, group2 = group2, data = d)))
  }
}
```

...but it stills rely on the name of the variable proba and group,
and the second part of the code is not a function.

A better implementation of functions to compute odd ratios

Better because it is very general and based on functions (example 1):

```
all_pairwise <- function(proba, groups){
  groups_id <- unique(groups)

  results <- matrix(NA, ncol = length(groups_id), nrow = length(groups_id))
  colnames(results) <- groups_id
  rownames(results) <- groups_id

  for(group1 in groups_id) {
    for(group2 in groups_id) {
      results[group1, group2] <- odds_ratio(p1 = proba[groups == group1], p2 = proba[groups == group2])
    }
  }

  return(results)
}

odds_ratio <- function(p1, p2){
  OR <- (p1/(1 - p1)) / (p2/(1 - p2))
  return(OR)
}

all_pairwise(proba = d$proba, groups = d$group)
```

	A	B	C
A	1	0.1111111	0.1666667
B	9	1.0000000	1.5000000
C	6	0.6666667	1.0000000

A better implementation of functions to compute odd ratios

Better because it is very general and based on functions (example 2):

```
all_pairwise2 <- function(proba, groups){
  groups_id <- unique(groups)
  to_do <- expand.grid(groups_id, groups_id)

  OR <- apply(to_do, 1, function(gr) {
    odds_ratio(p1 = proba[groups == gr[1]], p2 = proba[groups == gr[2]])
  })

  return(data.frame(group1 = to_do[, 1],
                    group2 = to_do[, 2],
                    OR = OR))
}

odds_ratio <- function(p1, p2){
  OR <- (p1/(1 - p1)) / (p2/(1 - p2))
  return(OR)
}

all_pairwise2(proba = d$proba, groups = d$group)
```

##	group1	group2	OR
## 1	A	A	1.0000000
## 2	B	A	9.0000000
## 3	C	A	6.0000000
## 4	A	B	0.1111111
## 5	B	B	1.0000000
## 6	C	B	0.6666667
## 7	A	C	0.1666667
## 8	B	C	1.5000000
## 9	C	C	1.0000000

Some key advices

- everything you use in the body must pass through the inputs
- everything you output must pass through the return
- try to write functions that you could reuse in other situations
- there are many ways to reach the same outcome; experiment a bit to find something you like/understand

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Challenge

The t-test and the Mann-Whitney U tests are two tests aiming at comparing 2 groups. We want to compare the risk of false positives and true positives of these two tests in the following conditions:

- assuming that height of males is gaussian with mean 180 cm and SD 6 cm and that the height of females is gaussian with mean 170 cm and SD 5 cm, how many individuals (sex-ratio = 1) do I need to get a power of 80% (risk of false negative = 20%)?
- assuming that the null hypothesis is true and that you have sampled 20 males and 20 females, what is the probability of false positives for the threshold $\alpha = 0.05$? (And for any threshold between 0 and 1%?)

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Learning by mimicking

Looking at code written by others will teach you

- how their functions work
- how to code
- new functions or packages that could be usefull for you

How to get the code behind a function?

Start by simply typing the function name (without brackets):

```
mosaic::oddsRatio  
## function (x, conf.level = 0.95, verbose = !quiet, quiet = TRUE,  
##     digits = 3)  
## {  
##     orrr(x, conf.level = conf.level, verbose = verbose, digits = digits,  
##         relrisk = FALSE)  
## }  
## <bytecode: 0x5574a1d02158>  
## <environment: namespace:mosaic>
```

How to get the code behind a function?

Then, follows the successive calls:

```
mosaic::orrr
## function (x, conf.level = 0.95, verbose = !quiet, quiet = TRUE,
##   digits = 3, relrisk = FALSE)
## {
##   if (any(dim(x) != c(2, 2))) {
##     stop("expecting something 2 x 2")
##   }
##   names(x) <- NULL
##   row.names(x) <- NULL
##   colnames(x) <- NULL
##   rowsums <- rowSums(x)
##   p1 <- x[1, 1]/rowsums[1]
##   p2 <- x[2, 1]/rowsums[2]
##   o1 <- p1/(1 - p1)
##   o2 <- p2/(1 - p2)
##   RR <- p2/p1
##   OR <- o2/o1
##   crit <- qnorm((1 - conf.level)/2, lower.tail = FALSE)
##   names(RR) <- "RR"
##   log.RR <- log(RR)
##   SE.log.RR <- sqrt(sum(x[, 2]/x[, 1]/rowsums))
##   log.lower.RR <- log.RR - crit * SE.log.RR
##   log.upper.RR <- log.RR + crit * SE.log.RR
##   lower.RR <- exp(log.lower.RR)
##   upper.RR <- exp(log.upper.RR)
##   names(OR) <- "OR"
##   log.OR <- log(OR)
##   SE.log.OR <- sqrt(sum(1/x))
##   log.lower.OR <- log.OR - crit * SE.log.OR
##   log.upper.OR <- log.OR + crit * SE.log.OR
##   lower.OR <- exp(log.lower.OR)
##   upper.OR <- exp(log.upper.OR)
```

How to get the code behind a function?

Sometimes, the code is not directly displayed... e.g. R methods (S3):

```
residuals
## function (object, ...)
## UseMethod("residuals")
## <bytecode: 0x557498868908>
## <environment: namespace:stats>
```

`residuals()` is a *generic* function which rely on class specific *methods*:

```
methods(residuals)
## [1] residuals.default*      residuals.glm
## [3] residuals.gls*          residuals.glsStruct*
## [5] residuals.gnls*         residuals.gnlsStruct*
## [7] residuals.HoltWinters*   residuals.isoreg*
## [9] residuals.lm             residuals.lme*
## [11] residuals.lmeStruct*     residuals.lmList*
## [13] residuals.loglm*        residuals.nlmeStruct*
## [15] residuals.nls*          residuals.psych*
## [17] residuals.smooth.spline* residuals.tukeyline*
## see '?methods' for accessing help and source code
```

The methods with a * are not exported from their package namespace!

How to get the code behind a function?

Add the class at the end of the function name to get the code for exported R (S3) methods:

```
residuals.lm
## function (object, type = c("working", "response", "deviance",
##      "pearson", "partial"), ...)
## {
##   type <- match.arg(type)
##   r <- object$residuals
##   res <- switch(type, working = , response = r, deviance = ,
##       pearson = if (is.null(object$weights)) r else r * sqrt(object$weights),
##       partial = r)
##   res <- naresid(object$na.action, res)
##   if (type == "partial")
##       res <- res + predict(object, type = "terms")
##   res
## }
## <bytecode: 0x5574a278fc38>
## <environment: namespace:stats>
```

Note: this requires to know the class of the object you work with!
You can use `class()` on your input to figure this out.

How to get the code behind a function?

It is also possible to get the code of non-exported R methods (S3):

```
residuals.nls
## Error in eval(expr, envir, enclos): object 'residuals.nls' not found
```

```
getAnywhere("residuals.nls") # or getS3method("residuals", "nls")

## A single object matching 'residuals.nls' was found
## It was found in the following places
##   registered S3 method for residuals from namespace stats
##   namespace:stats
## with value
##
## function (object, type = c("response", "pearson"), ...)
## {
##   type <- match.arg(type)
##   if (type == "pearson") {
##     val <- as.vector(object$m$resid())
##     std <- sqrt(sum(val^2)/(length(val) - length(coef(object))))
##     val <- val/std
##     if (!is.null(object$na.action))
##       val <- naresid(object$na.action, val)
##     attr(val, "label") <- "Standardized residuals"
##   }
##   else {
##     val <- as.vector(object$m$lhs() - object$m$fitted())
##     if (!is.null(object$na.action))
##       val <- naresid(object$na.action, val)
##     lab <- "Residuals"
##     if (!is.null(aux <- attr(object, "units")$y))
##       lab <- paste(lab, aux)
##     attr(val, "label") <- lab
##   }
## }
```


Challenge

Which function actually computes the numbers behind `boxplot()`?

Challenge

What is the code behind `t.test()`?

How to get the code behind a function?

Some functions – the interfaces – call functions that are written in other languages. The source code of these latter functions is not directly visible (spotted as `.C()`, `.Fortran()`, `.Call()`, `.Primitive()`, `.Internal()`, `.External()`).

```
dnorm
## function (x, mean = 0, sd = 1, log = FALSE)
## .Call(C_dnorm, x, mean, sd, log)
## <bytecode: 0x55749d43f748>
## <environment: namespace:stats>
```

In these cases, the easiest is to use the read-only mirror for R (<https://github.com/wch/r-source>) or the relevant package on Github! (here, the answer lies in `r-source/src/nmath/dnorm.c`)

Challenge

What is the code really estimating coefficients behind `lm()`?

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Debugging a faulty function

```
pythagora <- function(x, y) {  
  x2 <- x^2  
  y2 <- y^2  
  hyp <- (x^2 + y^2)^1/2  
  return(hyp)  
}  
  
pythagora(x = 2, y = 2)  
## [1] 4
```

Debugging a faulty function

```
pythagora <- function(x, y) {  
  x2 <- x^2  
  y2 <- y^2  
  hyp <- (x^2 + y^2)^1/2  
  return(hyp)  
}  
  
pythagora(x = 2, y = 2)  
## [1] 4
```

```
pythagora <- function(x, y) {  
  x2 <- x^2  
  y2 <- y^2  
  browser()  
  hyp <- (x^2 + y^2)^1/2  
  return(hyp)  
}  
  
pythagora(x = 2, y = 2)
```

Note: this is very usefull but you need to have access to the code!

Debugging a faulty function

```
pythagora <- function(x, y) {  
  x2 <- x^2  
  y2 <- y^2  
  hyp <- (x^2 + y^2)^1/2  
  return(hyp)  
}  
  
pythagora(x = 2, y = 2)  
## [1] 4
```


Debugging a faulty function

```
pythagora <- function(x, y) {  
  x2 <- x^2  
  y2 <- y^2  
  hyp <- (x^2 + y^2)^1/2  
  return(hyp)  
}  
  
pythagora(x = 2, y = 2)  
## [1] 4
```

```
debug(pythagora)  
  
pythagora(x = 2, y = 2)  
  
undebug(pythagora) ## when you are done, or use debugonce() above, or reload the function
```

Note: this can work on any function without having to mess with the code!

Debugging a faulty function

There are plenty more debugging possibilities out there!

Check:

- the nice possibility with R studio:
`https://support.rstudio.com/hc/en-us/articles/205612627-Debugging-with-RStudio`
- `?option` and look at error
- `?traceback`
- `?trace`

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Working on the language

Recursions

Using C++ code on the fly!

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There are other object systems than S3!