

Workshop 5: Programming in R

QCBS R Workshop Series

Québec Centre for Biodiversity Science



About this workshop



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Learning Objectives

- 1. Recognizing control flow;
- 2. Getting comfortable with testing conditions and performing iterations;
- 3. Developing your first functions in R;
- 4. Discovering how to accelerate your code;
- 5. Demonstrating useful R packages for biologists.

Review

Objects

Review: Vectors

Recall Workshop #1?

Numeric vectors

```
num.vector <- c(1, 4, 3,
9, 32, -4)
num.vector
# [1] 1 4 3 9 32 -4
```

Character vector

Logical vector

```
bool_vector <- c(TRUE, TRUE, FALSE) # or c(T, T, F)
bool_vector
# [1] TRUE TRUE FALSE</pre>
```

Review: Data frames

We can begin by creating multiple vectors (remember Workshop #1):

```
siteID <- c("A1.01", "A1.02", "B1.01", "B1.02")
soil_pH <- c(5.6, 7.3, 4.1, 6.0)
num.sp <- c(17, 23, 15, 7)
treatment <- c("Fert", "Fert", "No_fert", "No_fert")</pre>
We then combine them using the function data.frame().
my.first.df <- data.frame(siteID, soil_pH, num.sp, treatment)</pre>
my.first.df
   siteID soil_pH num.sp treatment
# 1 A1.01 5.6 17 Fert
# 2 A1.02 7.3 23 Fert
# 3 B1.01 4.1 15 No_fert
# 4 B1.02 6.0 7 No_fert
```

Review: Lists

We can also create lists by combining the vectors we created before.

```
my.first.list <- list(siteID, soil_pH, num.sp, treatment)</pre>
```

```
my.first.list
# [[1]]
# [1] "A1.01" "A1.02" "B1.01" "B1.02"
#
# [[2]]
# [1] 5.6 7.3 4.1 6.0
#
# [[3]]
# [1] 17 23 15 7
# [[4]]
# [1] "Fert" "Fert" "No_fert" "No_fert"
```

Control flow

Control flow

Program flow control can be simply defined as the order in which a program is executed.

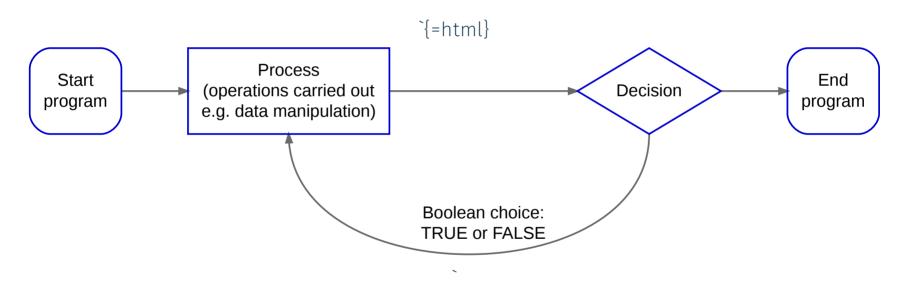
Why is it advantageous to have structured programs?

- It decreases the complexity and time of the task at hand;
- This logical structure also means that the code has increased clarity;
- It also means that many programmers can work on one program.

This means increased productivity.

Control flow

Program flowcharts can be used to plan programs and represent their structure.



Representing structure

The two basic building blocks of codes are the following:

Selection

Program's execution determined by statements

```
if() {}
if() {} else {}
```

Iteration

Repetition, where the statement will **loop** until a criteria is met

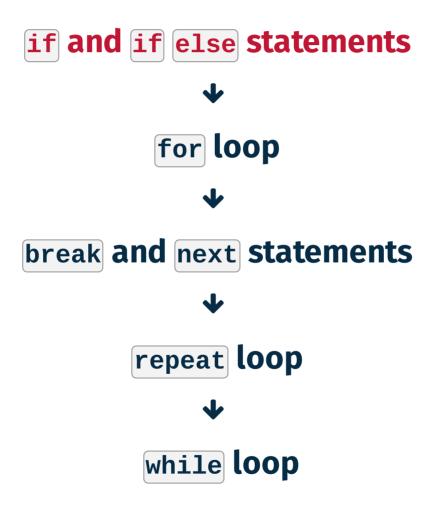
```
for() {}
while() {}
repeat {}
```

Selection and iterative statements can also be controlled by termination and jump statements:

Termination and Jump

break next

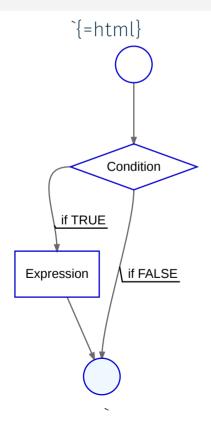
Control flow roadmap



Decision making

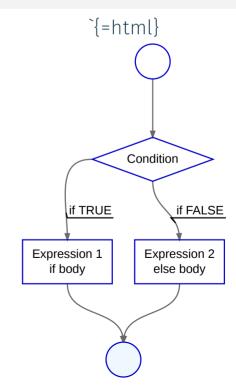
if() statement

```
if(condition) {
  expression
}
```



if() else statement

```
if(condition) {
   expression 1
} else {
   expression 2
}
```



What if you want to test more than one condition?

- if() and if() else test a single condition.
- You can also use the ifelse() function to:
 - test a vector of conditions;
 - apply a function only under certain conditions.

Nested if() else statement

While the if() and if() else statements leave you with exactly two options, nested if() else statement allows you consider more alternatives.

```
if(test_expression1) {
   statement1
} else if(test_expression2) {
   statement2
} else if(test_expression3) {
   statement3
} else {
   statement4
}
```

`{=html}

Beware of R's expression parsing!

What do you think will happen if we try the code below?

This does not work because R evaluates the first line and does not know that you are going to use an else statement

Use curly brackets { } so that **R** knows to expect more input. Try:

```
if(2+2 == 4) {
  print("Arithmetic works.")
} else {
  print("Houston, we have a problem.")
}
# [1] "Arithmetic works."
```

Challenge 1



Consider the following objects:

```
Paws <- "cat"
Scruffy <- "dog"
Sassy <- "cat"
animals <- c(Paws, Scruffy, Sassy)
```

- 1. Use an **if()** statement to print "meow" if **Paws** is a "cat".
- 2. Use an if() else statement to print "woof" if you supply an object that is a "dog" and "meow" if it is not. Try it out with Paws and Scruffy.
- 3. Use the ifelse() function to
 display "woof" for animals that
 are dogs and "meow" for animals
 that are cats.

Remember the logical operators

Command	Meaning
	equal to
[=	not equal to
	less than
<=	less than or equal to
\triangleright	greater than
>=	greater than or equal to
x&y	X AND y
x y	x OR y
<pre>isTRUE(x)</pre>	test if $\widehat{\mathbf{x}}$ is true

Challenge 1 - Solution



1. Use an [if()] statement to print "meow" if Paws is a "cat".

```
if(Paws == 'cat') {
  print("meow")
}
# [1] "meow"
```

2. Use an if() else statement to print "woof" if you supply an object that is a "dog" and "meow" if it is not. Try it out with Paws and Scruffy.

```
x = Paws
# x = Scruffy
if(x == 'cat') {
   print("meow")
} else {
   print("woof")
}
# [1] "meow"
```

Challenge 1 - Solution



3. Use the <code>ifelse()</code> function to display <code>"woof"</code> for <code>animals</code> that are dogs and <code>"meow"</code> for <code>animals</code> that are cats.

```
animals <- c(Paws, Scruffy, Sassy)

ifelse(animals == 'dog', "woof", "meow")
# [1] "meow" "woof" "meow"</pre>
```

Or

```
for(val in 1:3) {
   if(animals[val] == 'cat') {
      print("meow")
   } else if(animals[val] == 'dog') {
      print("woof")
   } else print("what?")
}
# [1] "meow"
# [1] "woof"
# [1] "meow"
```

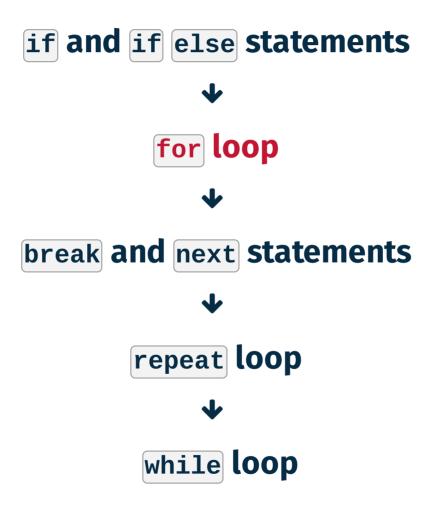
Iteration

Every time some operations have to be repeated, a loop may come in handy.

Loops are good for:

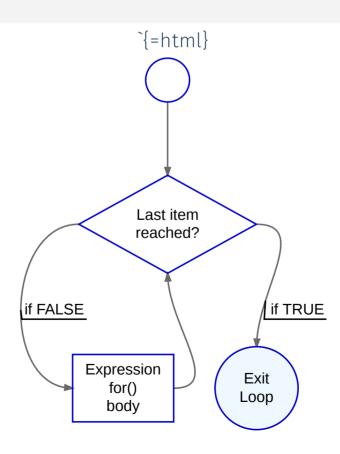
- Doing something for every element of an object;
- Doing something until the processed data runs out;
- Doing something for every file in a folder;
- Doing something that can fail, until it succeeds;
- Iterating a calculation until it converges.

Control flow roadmap



A for() loop works in the following way:

```
for(i in sequence) {
  expression
  }
```



A for() loop works in the following way:

```
for(i in sequence) {
  expression
  }
```

The letter i can be replaced with any variable name, sequence can be elements or the position of these elements, and expression can be anything:

```
# [1] "Hello"
# [1] "R"
# [1] "Programmers"
```

```
# [1] 5.966294
# [1] 6.8881
# [1] 13.18717
# [1] 17.97483
```

A **for()** loop works in the following way:

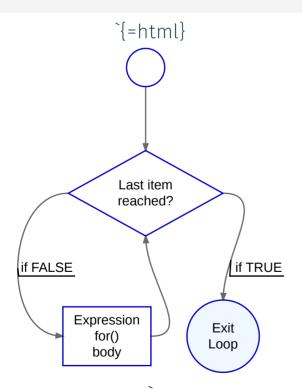
```
for(i in sequence) {
  expression
  }
```

As expected, you can use **for()** loops in different object types and classes, such as a **list**:

```
for(element in elements) {
   print(element*2)
}
# [1] 2 4 6
# [1] 8 10 12 14 16 18 20
# [1] 14 12 10 8 6 4 2 0 -2
```

Remember our flowchart?

```
for(i in sequence) {
  expression
  }
```



We will apply the following example into our flow chart!

Here, every instance of m is being replaced by each number between 1 and 7, until it reaches the last element of the sequence:

```
y <- 2
for(m in 1:6) {
  print(y*m)
}

# [1] 2
# [1] 4
# [1] 6
# [1] 8
# [1] 10
# [1] 12</pre>
```

Remember our flowchart?

```
for(i in sequence) {
  expression
  }
  `{=html}
```

We will apply the following example into our flow chart!

Here, every instance of m is being replaced by each number between 1 and 7, until it reaches the last element of the sequence:

```
y <- 2
for(m in 1:6) {
  print(y*m)
}

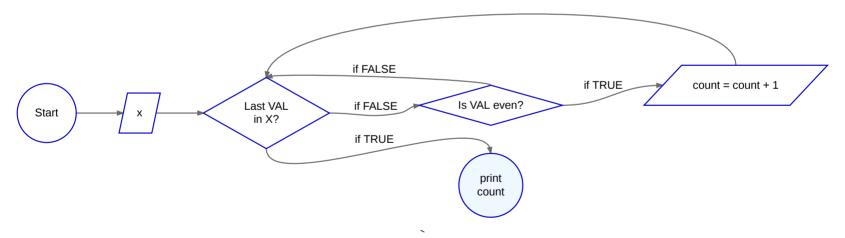
# [1] 2
# [1] 4
# [1] 6
# [1] 8
# [1] 10
# [1] 12</pre>
```

Let us perform operations for even elements within x using the modulo operator (%%):

```
x <- c(2, 5, 3, 9, 6)
count <- 0
```

`{=html}

```
for(val in x) {
   if(val %% 2 == 0) {
     count <- count + 1
   }
}
print(count)</pre>
```



for() loops are often used to loop over a dataset. We will use loops to perform functions on the **co2** dataset which is built in **R**.

```
data(CO2) # This loads the built in dataset

for(i in 1:length(CO2[,1])) { # for each row in the CO2 dataset
   print(CO2$conc[i]) # print the CO2 concentration
}
```

First 40 outputs:

```
# [1] 95
                   # [1] 350
                                            [1] 1000
                                                              # [1] 250
  [1] 175
                   # [1] 500
                                          # [1] 95
                                                                [1] 350
  [1] 250
                   # [1] 675
                                            [1] 175
                                                                [1] 500
  [1] 350
                   # [1] 1000
                                            [1] 250
                                                                [1] 675
  [1] 500
                   # [1] 95
                                            [1] 350
                                                                [1]
                                                                    1000
 [1] 675
                   # [1] 175
                                            [1] 500
                                                              # [1] 95
 [1] 1000
                                                                [1] 175
                   # [1] 250
                                            [1] 675
# [1] 95
                   # [1] 350
                                          # [1] 1000
                                                              # [1] 250
 [1] 175
                   # [1] 500
                                          # [1] 95
                                                                    350
                                                                [1]
                                          # [1] 175
 [1] 250
                   # [1] 675
                                                              # [1] 500
```

Another example:

```
for(i in 1:length(CO2[,1])) { # for each row in the CO2 dataset
  if(CO2$Type[i] == "Quebec") { # if the type is "Quebec"
    print(CO2$conc[i]) # print the CO2 concentration
  }
}
```

Outputs:

[1] 350

```
# [1] 95
                    # [1] 500
                                           # [1] 175
                                                               # [1] 675
  [1] 175
                    # [1] 675
                                             [1] 250
                                                                  [1]
                                                                      1000
  [1] 250
                    # [1] 1000
                                           # [1] 350
                                                               # [1]
                                                                      95
                                                                  [1]
  [1] 350
                    # [1] 95
                                             [1] 500
                                                                      175
  [1] 500
                                                                  [1]
                                                                      250
                    # [1] 175
                                             [1] 675
                      [1] 250
                                                                      350
  [1] 675
                                             [1] 1000
                                                                  [1]
  [1] 1000
                      [1] 350
                                             [1] 95
                                                                  [1]
                                                                      500
  [1] 95
                    # [1] 500
                                             [1] 175
                                                                  [1] 675
  [1] 175
                      [1] 675
                                             [1] 250
                                                                  [1] 1000
 [1] 250
                    # [1] 1000
                                             [1] 350
```

[1] 500

[1] 95

```
Tip 1. To loop over the number of rows of a data frame, we can use the function (nrow()).
```

```
for(i in 1:nrow(CO2)) {
    # for each row in
    # the CO2 dataset
  print(CO2$conc[i])
    # print the CO2
    # concentration
}
```

```
# [1] 95
                       [1]
                            1000
  [1] 175
                            95
                       [1]
  [1] 250
                       [1]
                            175
                            250
  [1] 350
                       [1]
  [1] 500
                       [1] 350
  [1] 675
                       [1]
                            500
  [1] 1000
                       [1] 675
  [1] 95
                       [1]
                            1000
  [1] 175
                       [1] 95
  [1] 250
                       [1]
                            175
  [1] 350
                       [1]
                            250
  [1] 500
                            350
                       [1]
  [1] 675
                            500
                       [1]
  [1] 1000
                       [1]
                            675
  [1] 95
                       [1]
                            1000
  [1] 175
                       [1] 95
  [1] 250
                       [1]
                            175
                            250
  [1] 350
                       [1]
  [1] 500
                       [1]
                            350
                       [1] 500
  [1] 675
```

Tip 2. To perform operations on the elements of one column, we can directly iterate over it.

```
for(p in CO2$conc) {
    # for each element of
    # the column "conc" of
    # the CO2 df
  print(p)
    # print the p-th element
}
```

```
# [1] 95
                       [1]
                           1000
  [1] 175
                           95
                       [1]
  [1] 250
                       [1]
                           175
  [1] 350
                       [1]
                           250
  [1] 500
                       [1] 350
  [1] 675
                       [1]
                           500
  [1] 1000
                       [1] 675
  [1] 95
                       [1]
                           1000
  [1] 175
                     # [1] 95
  [1] 250
                       [1]
                           175
  [1] 350
                       [1]
                           250
  [1] 500
                           350
                       [1]
  [1] 675
                           500
                       [1]
  [1] 1000
                       [1]
                           675
  [1] 95
                       [1]
                           1000
  [1] 175
                       [1] 95
  [1] 250
                       [1]
                           175
  [1] 350
                       [1]
                           250
  [1] 500
                       [1]
                           350
                       [1] 500
  [1] 675
```

The expression within the loop can be almost anything and is usually a compound statement containing many commands.

```
for(i in 4:5) { # for i in 4 to 5
  print(colnames(CO2)[i])
  print(mean(CO2[,i])) # print the mean of that column from the CO2 datas
}
```

Output:

```
# [1] "conc"
# [1] 435
# [1] "uptake"
# [1] 27.2131
```

Nested for() loops within for() loops

In some cases, you may want to use nested loops to accomplish a task. When using nested loops, it is important to use different variables as counters for each of your loops. Here we used **i** and **n**:

```
for(i in 1:3) {
  for(n in 1:3) {
    print(i*n)
  }
}
```

```
# Output
# [1] 1
# [1] 2
# [1] 3
# [1] 4
# [1] 6
# [1] 3
# [1] 9
```

Getting good: Using the apply() family

R disposes of the apply() function family, which consists of iterative functions that aim at minimizing your need to explicitly create loops.

apply() can be used to apply functions to a matrix.

lapply()

lapply() applies a function to every element of a list.

It may be used for other objects like dataframes, lists or vectors.

The output returned is a **list** (explaining the "**l**" in **lapply**) and has the same number of elements as the object passed to it.

```
SimulatedData <- list(
   SimpleSequence = 1:4,
   Norm10 = rnorm(10),
   Norm20 = rnorm(20, 1),
   Norm100 = rnorm(100, 5)
   )

# Apply mean to each element
## of the list
lapply(SimulatedData, mean)</pre>
```

```
# $SimpleSequence
# [1] 2.5
#
# $Norm10
# [1] -0.05895691
#
# $Norm20
# [1] 0.8336549
#
# $Norm100
# [1] 5.00853
```

sapply()

sapply() is a 'wrapper' function for lapply(), but returns a simplified output as
a vector, instead of a list.

mapply()

mapply() works as a multivariate version of sapply().

It will apply a given function to the first element of each argument first, followed by the second element, and so on. For example:

```
lilySeeds <- c(80, 65, 89, 23, 21)
poppySeeds <- c(20, 35, 11, 77, 79)

# Output
mapply(sum, lilySeeds, poppySeeds)
# [1] 100 100 100 100 100</pre>
```

tapply()

tapply() is used to apply a function over subsets of a vector.

It is primarily used when the dataset contains different groups (*i.e.* levels or factors), and we want to apply a function to each of these groups.

```
mtcars[1:10, c("hp", "cyl")]
#
                  hp cvl
                 110
# Mazda RX4
 Mazda RX4 Wag 110 6
 Datsun 710
            93 4
# Hornet 4 Drive 110
                       6
# Hornet Sportabout 175
                       8
# Valiant
                 105
                       6
# Duster 360 245
# Merc 240D
                 62
                       4
# Merc 230
                 95
                       4
# Merc 280
                 123
                       6
```

Challenge 2



After coming back from the field, you have realized that your tool for measuring ${\bf CO_2}$ uptake was not calibrated properly at Quebec sites and all measurements are 2 units higher than they should be.

Now, you must do the following:

- 1. Use a loop to correct these measurements for all Quebec sites;
- 2. Use an apply() family function to calculate the average CO_2 uptake in both Québec and Mississipi sampled sites.

For this, you must load the CO_2 dataset using $data(CO_2)$, and then use the object CO_2 .

Get your hands dirty!

Challenge 2: Solution



1. Using for() and if() to correct the measurements:

```
for(i in 1:dim(CO2)[1]) {
   if(CO2$Type[i] == "Quebec") {
      CO2$uptake[i] <- CO2$uptake[i] - 2
   }
}</pre>
```

2. Using tapply() to calculate the mean for each group:

```
tapply(CO2$uptake, CO2$Type, mean)
# Quebec Mississippi
# 31.54286 20.88333
```

Modifying iterations

Normally, loops iterate over and over until they finish.

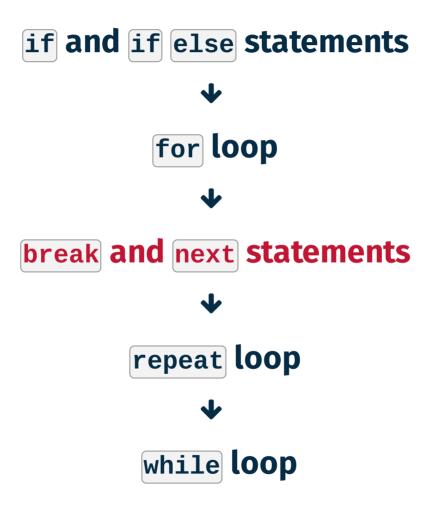
Sometimes you may be interested in breaking this behaviour.

For example, you may want to tell **R** to stop executing the iteration when it reaches a given element or condition.

You may also want **R** to jump certain elements when certain conditions are met.

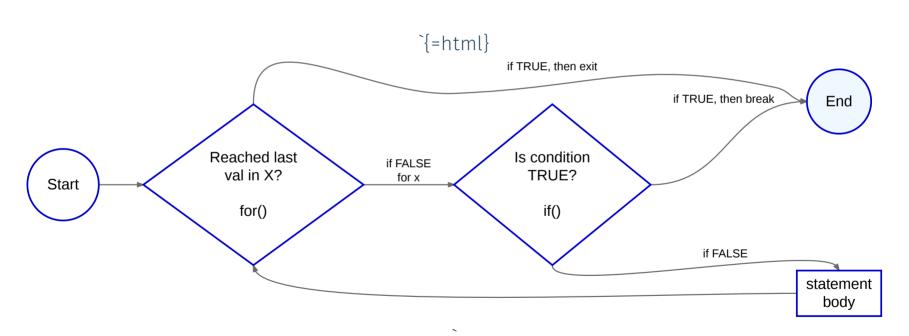
For this, we will introduce break, next and while.

Control flow roadmap



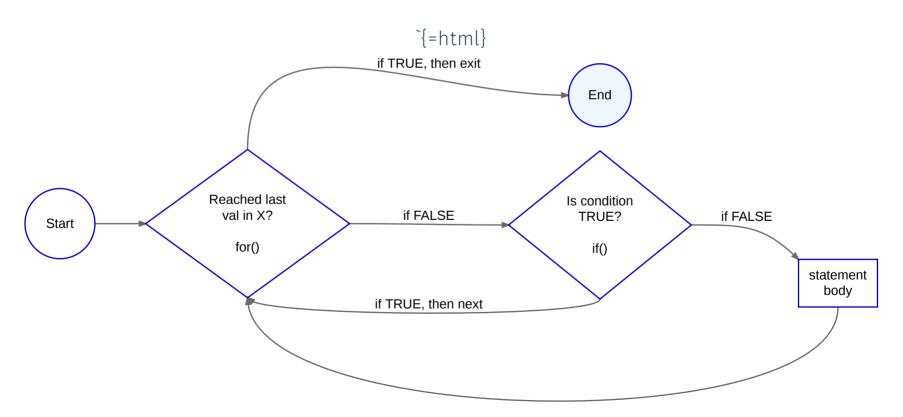
Modifying iterations: break statement

```
for(val in x) {
  if(condition) { break }
  statement
}
```



Modifying iterations: next statement

```
for(val in x) {
  if(condition) { next }
  statement
}
```

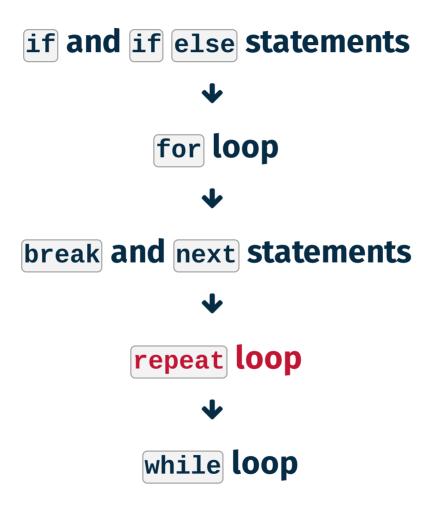


Modifying iterations: next statement

Print the CO_2 concentrations for *chilled* treatments and keep count of how many replications were done.

```
count <- 0
 for(i in 1:nrow(CO2)) {
   if(CO2$Treatment[i] == "nonchilled") next
   # Skip to next iteration if treatment is nonchilled
   count <- count + 1
 # print(CO2$conc[i]) # You can turn this on if you want to
 print(count) # The count and print command were performed 42 times.
# [1] 42
 sum(CO2$Treatment == "chilled")
 # [1] 42
```

Control flow roadmap



Modifying iterations: repeat loop

Under repeat, a task is performed until it is deliberately stopped:

```
repeat {
  print("Press 'Esc' to stop me!"
}
```

```
[1] "Press 'Esc' to stop me!"
[1] "Press 'Esc' to stop me!"
...
[1] "Press 'Esc' to stop me!"
```

You can stop this cycle using a condition and break:

```
repeat {
  expression
  if {
    condition
    } break
}
```

`{=html}

Modifying iterations: repeat loop

Remember our code that used the next statement to print the total counts of CO₂ concentrations taken for *chilled* treatments?

```
count <- 0

for(i in 1:nrow(CO2)) {
   if(CO2$Treatment[i] == "nonchil
   count <- count + 1
}

print(count)
# [1] 42</pre>
```

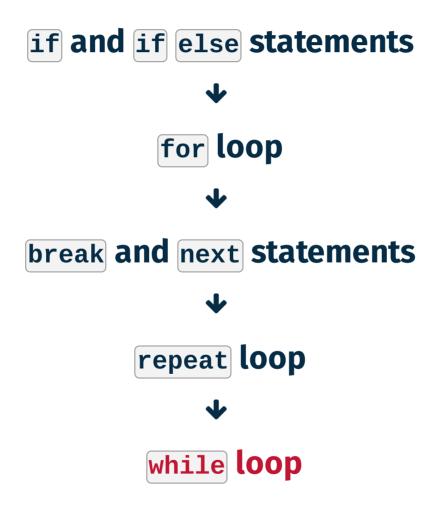
It could had been equivalently written using repeat and break:

```
count <- 0
i <- 0

repeat {
    i <- i + 1
    if(C02$Treatment[i] == "non
    count <- count + 1
    if(i == nrow(C02)) break
}

print(count)
# [1] 42</pre>
```

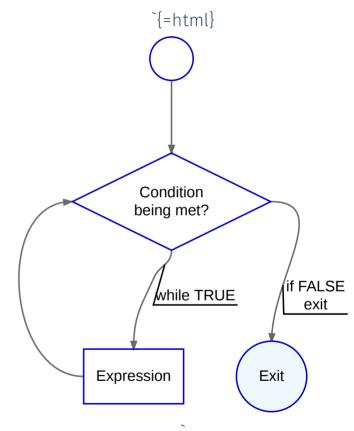
Control flow roadmap



Modifying iterations: while loop

Within the while loop, an expression happens while a condition is met.

```
while(condition){
  expression
}
```



Modifying iterations: while loop

Again, remember our code that used the **next** statement to print the total counts of CO_2 concentrations taken for *chilled* treatments?

```
count <- 0
for(i in 1:nrow(CO2)) {
   if(CO2$Treatment[i] == "nonchil
   count <- count + 1
}
print(count)</pre>
```

It could had been equivalently written using repeat and break:

```
count <- 0
i <- 0
repeat {
    i <- i + 1
    if(C02$Treatment[i] == "non
    count <- count + 1
    if(i == nrow(C02)) break
}
print(count)</pre>
```

Or, using a while loop:

```
i <- 0
count <- 0
while(i < nrow(CO2))
{
    i <- i + 1
    if(CO2$Treatment[i] == "nonchilled") next</pre>
```

Challenge 3



You have realized that another of your tools was not working properly!

At **Mississippi** sites, **concentrations** less than 300 were measured correctly, but concentrations equal or higher than 300 were overestimated by 20 units!

Your *mission* is to use a loop to correct these measurements for all Mississippi sites.

Tip. Make sure you reload the CO_2 data so that we are working with the raw data for the rest of the exercise:

data(CO2)

Challenge 3: Solution



```
for(i in 1:nrow(CO2)) {
   if(CO2$Type[i] == "Mississippi") {
      if(CO2$conc[i] < 300) next
      CO2$conc[i] <- CO2$conc[i] - 20
   }
}</pre>
```

Note: We could also have written it in this way, which is more concise and clearer:

```
for(i in 1:nrow(CO2)) {
   if(CO2$Type[i] == "Mississippi" && CO2$conc[i] >= 300) {
      CO2$conc[i] <- CO2$conc[i] - 20
   }
}</pre>
```

Writing functions

Why write functions?

Imagine that we would like to rescale variables to the range of \bigcirc to \bigcirc 1.

Our dataset has four variables:

```
our.dataset <- data.frame(</pre>
  a = rnorm(10),
  b = rnorm(10),
  c = rnorm(10),
  d = rnorm(10)
```

The equation for doing that is:

$$x_{ ext{new}} = rac{x_i - \min(x)}{\max(x) - \min(x)}$$

We can rescale these four variables to **0** and **1** by doing this:

```
our.dataset$a <- (our.dataset$a - min(our.dataset$a, na.rm = TRUE)) /
  (\max(\text{our.dataset}\$a, \text{na.rm} = \text{TRUE}) - \min(\text{our.dataset}\$a, \text{na.rm} = \text{TRUE}))
our.dataset$b <- (our.dataset$b - min(our.dataset$b, na.rm = TRUE)) /
  (max(our.dataset$b, na.rm = TRUE) - min(our.dataset$a, na.rm = TRUE))
our.dataset$c <- (our.dataset$c - min(our.dataset$c, na.rm = TRUE)) /
  (max(our.dataset$c, na.rm = TRUE) - min(our.dataset$c, na.rm = TRUE))
our.dataset$d <- (our.dataset$d - min(our.dataset$d, na.rm = TRUE)) /
  (max(our.dataset$d, na.rm = TRUE) - min(our.dataset$d, na.rm = TRUE))
                                                                        68 / 106
```

Why write functions?

Imagine that we would like to rescale variables to the range of **0** to **1**.

Our dataset has four variables:

The equation for doing that is:

```
our.dataset <- data.frame(
    a = rnorm(10),
    b = rnorm(10),
    c = rnorm(10),
    d = rnorm(10)
)</pre>
```

$$x_{
m new} = rac{x_i - \min(x)}{\max(x) - \min(x)}$$

Repeating that equation and this chunk of code 31 times could be a bit tedious:

```
our.dataset$a <- (our.dataset$a - min(our.dataset$a, na.rm = TRUE)) /
  (max(our.dataset$a, na.rm = TRUE) - min(our.dataset$a, na.rm = TRUE))</pre>
```

But, we can see that, except from the *input*, the code was pratically the same among the variables. We can then use our ninja abilities and **create a function** that will do that task for us:

```
# our
```

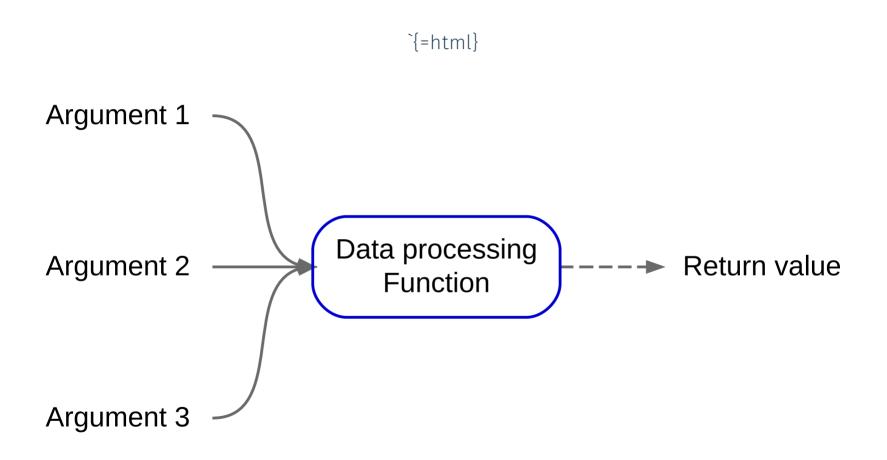
Why write functions?

Much of the heavy lifting in ${\Bbb R}$ is done by functions!

Functions are useful for:

- 1. Performing a task repeatedly, but configurably;
- 2. Making code more readable;
- 3. Making code easier to modify and maintain;
- 4. Sharing code between different analyses;
- 5. Sharing code with other people;
- 6. Modifying **R**'s built-in functionality.

What is a function?



Syntax of a function: function()

```
function_name <- function(argument1, argument2, ...) {
  body # What we want the function to do
  return(values) # Optional
}</pre>
```

- **function_name** is the name of the function, and will be stored in the **R** environemnt as an object with this name;
- **argument**'s take the defined values that can be used within the function;
- **body** contains the statements that define what the function does;
- **output** contains the returned value from the function. If **return()** is absent, then the last expression is returned.

Arguments of a function()

```
function_name <- function(argument1, argument2, ...) {
  body # What we want the function to do
  return(values) # Optional
}</pre>
```

Arguments are the *input* values of your function and will have the information your function needs to be able to perform correctly.

A function can have between zero and an infinity of arguments. See the following example:

```
operations <- function(number1, number2, number3) {
  result <- (number1 + number2) * number3
  print(result)
}</pre>
```

```
operations(1, 2, 3)
# [1] 9
```

Challenge 4



Using what you learned previously on flow control, create a function print_animal() that takes an animal as argument and gives the following results:

```
Scruffy <- "dog"
Paws <- "cat"

print_animal(Scruffy)
# [1] "woof"

print_animal(Paws)
# [1] "meow"</pre>
```

Challenge 4: Solution



Using what you learned previously on flow control, create a function print_animal() that takes an animal as argument and gives the following results:

```
Scruffy <- "dog"
Paws <- "cat"
print_animal(Scruffy)
# [1] "woof"
print_animal(Paws)
# [1] "meow"
print_animal <- function(animal) {</pre>
  if(animal == "dog") {
    print("woof")
  } else if(animal == "cat") {
    print("meow")
```

Default argument values in a function

Arguments can be provided with a **default value**, or even be optional.

Default values are useful when using a function with the same settings. The flexibility to depart from default values is still there, if needed.

```
operations <- function(number1, number2, number3 = 3) {</pre>
  result <- (number1 + number2) * number3</pre>
  print(result)
operations(number1 = 1, number2 = 2, number3 = 3)
# [1] 9
# is equivalent to
operations(1, 2)
# [1] 9
operations(1, 2, 2) # we can still change the value of number3 if needed
# [1] 6
```

The ellipsis argument:

The special argument ... allows you to pass arguments from other undefined functions, *i.e.* allowing for an indefinite number of arguments to be inputed.

```
percentages <- function(x, mult =</pre>
 percent <- round(x * mult, ...)</pre>
 paste(percent, "%", sep = "")
percentages(c(.543, .534, .466))
# [1] "54%" "53%" "47%"
# ?round
percentages(c(.543, .534, .466),
# [1] "54.3%" "53.4%" "46.6%"
```

The ellipsis argument:

The special argument ... allows you to pass on arguments to another function used inside your function. Here we use ... to pass on arguments to plot() and points().

```
plot.C02 <- function(C02, ...) {
  plot(x=C02$conc, y = C02$uptake, type = "n", ...)
  for(i in 1:length(C02[,1])){
    if(C02$Type[i] == "Quebec") {
      points(C02$conc[i], C02$uptake[i], col = "red", type = "p", ...)
    } else if(C02$Type[i] == "Mississippi") {
      points(C02$conc[i], C02$uptake[i], col = "blue", type = "p", ...)
    }
}</pre>
```

Return values

The last expression evaluated in a **function** becomes the return value:

```
myfun <- function(x) {
   if(x < 10) {
      0
   } else {
      10
   }
}
myfun(5)
# [1] 0
myfun(15)
# [1] 10</pre>
```

function() itself returns the last evaluated value even without including return() function.

Return values

It can be useful to explicitly **return()** if the routine should end early, jump out of the function and return a value.

```
simplefun1 <- function(x) {
  if(x < 0)
  return(x)
}</pre>
```

Functions can return only a single object (and text). But this is not a limitation because you can return a **list** containing any number of objects.

```
simplefun2(<mark>1</mark>,
```

```
# $result
# [1] 3
#
# $x
# [1] 1
#
# $y
# [1] 2
```

Challenge 5



Using what you have just learned on functions and control flow, create a function named **bigsum** that takes two arguments **a** and **b** and:

- 1. Returns 0 if the sum of a and b is strictly less than 50;
- 2. Else, returns the sum of **a** and **b**.

Challenge 5: Solution



Using what you have just learned on functions and control flow, create a function named **bigsum** that takes two arguments **a** and **b** and:

- 1. Returns **o** if the sum of **a** and **b** is strictly less than **50**;
- 2. Else, returns the sum of **a** and **b**.

Answer 1

```
bigsum <- function(a, b) {
  result <- a + b
  if(result < 50) {
    return(0)
  } else {
    return(result)
  }
}</pre>
```

Answer 2

```
bigsum <- function(a, b) {
    result <- a + b
    if(result < 50) {
        0
        } else {
        result
        }
}</pre>
```

Accessibility of variables

It is essential to always keep in mind where your variables are, and whether they are defined and accessible:

- 1. Variables defined **inside** a function are not accessible outside from it!
- 2. Variables defined **outside** a function are accessible inside, and are not modified, even if they have the same name.

```
out_val <- 3
vartest <- function() {
   in_val <- 4
   print(in_val)
   print(out_val)
}
vartest()
# [1] 4
# [1] 3</pre>
```

```
in_val; out_val
# Error in eval(expr, envir, encl
# [1] 3
```

```
out_val_2 <- 3
vartest <- function(out_val_2) {
   print(out_val_2)
}

vartest(8)
# [1] 8

out_val_2
# [1] 3</pre>
```

What happens in the function club, stays in the function club.

Accessibility of variables

```
var1 <- 3
vartest <- function() {</pre>
 a <- 4 # 'a' is defined inside
 print(a) # print 'a'
  print(var1) # print var1
             # we cannot print 'a' as it exists only inside the function
a
# [1] 17.97483
vartest() # calling vartest() will print a and var1
# [1] 4
# [1] 3
rm(var1) # remove var1
vartest() # calling the function again does not work anymore
# [1] 4
# Error in print(var1): object 'var1' not found
```

Accessibility of variables

Tip. Be very careful when creating variables inside a conditional statement as the variable may never have been created and cause (sometimes imperceptible) errors.

Tip. It is good practice to define variables outside the conditions and then modify their values to avoid any problems.

```
a <- 3
if(a > 5) {
   b <- 2
}
a + b</pre>
```

```
# Error: object 'b' not found
```

If you had **b** already assigned in your environment, with a different value, you could have had a **bigger** problem!

No error would have been shown and **a** + **b** would have meant another thing!

Additional good programming practices

Why should I care about programming practices?

- It makes your life easier;
- It helps you achieve greater readability and makes sharing and reusing your code a lot less painful;
- It helps reduce the time you will spend remembering and understanding your own code.

Pay attention to the next tips!

Keep a clean and nice code

Proper indentation and spacing is the first step to get an easy to read code:

- Use spaces between and after your operators;
 - \circ x>=1&x<=10 is more difficult to read then x>=1 & x<=10
- Use consistently the same assignation operator;
 - <- is often preferred. = is sometimes OK, but do not switch all the time between the two.
- Use brackets and returns when using flow control statements;
 - Inside brackets, indent by at least two returns;
 - Put closing brackets on a separate line, except when preceding an else statement.
- Define each variable on its own line;
- Use **Cmd** + **I** or **Ctrl** + **I** in RStudio to indent the highlighted code automatically;

```
Nay!
```

Yay!

```
if((a[x,y]>1.0)&(a[x,y]<2.0)){pri</pre>
```

```
if((a[x, y] > 1.0) & (a[x, y] <
    print("Between 1 and 2")
}</pre>
```

Keep a clean and nice code

On the left, code is not spaced, nor indented. All brackets are in the same line, and it looks "messy".

```
a<-4;b=3
if(a<b){
if(a==0)print("a zero")}else{
if(b==0){print("b zero")}else pri</pre>
```

Keep a clean and nice code

On the left, code is not spaced, nor indented. All brackets are in the same line, and it looks "messy". On the right, it looks more organized, no?

```
a<-4;b=3
if(a<b){
if(a==0)print("a zero")}else{
if(b==0){print("b zero")}else pri</pre>
```

```
a <- 4
h <- 3
if(a < b) {
  if(a == 0) {
    print("a zero")
} else {
  if(b == 0) {
    print("b zero")
  } else {
    print(b)
```

Use functions to simplify your code

Write your own function:

- 1. When portion of the code is repeated more than a few times in your script;
- 2. If only a part of the code changes and includes options for different arguments.

This would also reduce the number of potential errors done by copy-pasting, and the time needed to correct them.

Use functions to simplify your code

Let's modify the example from **Challenge 3** and suppose that all CO_2 uptake from Mississipi plants was overestimated by 20 and Quebec underestimated by 50.

We could write this:

```
for(i in 1:length(CO2[,1])) {
   if(CO2$Type[i] == "Mississippi"
        CO2$conc[i] <- CO2$conc[i] -
    }
}
for(i in 1:length(CO2[,1])) {
   if(CO2$Type[i] == "Quebec") {
        CO2$conc[i] <- CO2$conc[i] +
   }
}</pre>
```

Or this:

```
recalibrate <- function(CO2, type
  for(i in 1:nrow(CO2)) {
    if(C02$Type[i] == type) {
      CO2$conc[i] <- CO2$conc[i]
  return(CO2)
newC02 <- recalibrate(C02 = C02,</pre>
                        type = "Mis
                        bias = -20)
newC02 <- recalibrate(newC02, "Qu</pre>
```

Use meaningful names for functions

Same function as before, but with vague names:

```
rc <- function(c, t, b) {
   for(i in 1:nrow(c)) {
     if(c$Type[i] == t) {
      c$uptake[i] <- c$uptake[i]
     }
   }
   return(c)
}</pre>
```

That being said:



programmers: nooooo you have to give your variables descriptive names mathematicians: haha function go f(r,r,r,r,r,r,r,r,r,r,r)

What is c and rc?

Whenever possible, avoid using names of existing **R** functions and variables to avoid confusion and conflits.

Use comments:

Final tip. Add comments to describe what your code does, how to use its arguments or a detailed step-by-step description of the function.

```
# Recalibrates the CO2 dataset by modifying the CO2 uptake concentration
# by a fixed amount depending on the region of sampling.
# Arguments
# CO2: the CO2 dataset
# type: the type ("Mississippi" or "Quebec") that need to be recalibrated
# bias: the amount to add or remove to the concentration uptake
recalibrate <- function(CO2, type, bias) {</pre>
  for(i in 1:nrow(CO2)) {
    if(CO2$Type[i] == type) {
      CO2$uptake[i] <- CO2$uptake[i] + bias
  return(CO2)
```

Challenge 6: Group exercise

Using what you learned, write an if() statement that tests whether a numeric variable x is 0. If not, it assigns cos(x)/x to z, otherwise it assigns 1 to z.

Create a function called $my_function()$ that takes the variable x as argument and returns z.

If we assign 45, 20, and 0 to \boxed{x} respectively, which of the following options would represent the results?

- **1.** 0.054, 0.012, and 0;
- **2.** 0.020, 0.054, and 1;
- **3.** 0.012, 0.020, and 1.

In addition to this, discuss with your group about a function that you would like to create (it can or it may not be related to your research). Be prepared to briefly describe it to us!

Group exercise: Solution



Correct answer is option 3 (0.12, 0.20, and 1).

```
my_function <- function(x) {</pre>
  if(x != 0) {
  z < - \cos(x)/x
  } else { z <- 1 }</pre>
  return(z)
my_function(45)
# [1] 0.01167382
my_function(20)
# [1] 0.0204041
my_function(0)
# [1] 1
```

Thank you for attending this workshop!

