Programming Topics II: Writing Functions



Abbie M Popa BSDS 100 - Intro to Data Science with $\ensuremath{\mathbb{R}}$

Outline



- What is a function?
- Writing your own functions
- Using functionals

Functions



- We have already used many functions, including mean(), seq(),
 c(), data.frame(), and ggplot()
- Just about anything we have used that is followed by parantheses
 () is a function
- Functions tell R to perform a specific task

Functions



- Each function has several parts
 - A name (e.g., mean)
 - A list of arguments often called formals(), these can be mandatory (e.g., mean() needs a vector of numbers to find the mean of, or optional (e.g., mean() can take the argument na.rm = T)
 - A body (), which is the code a function executes

Functions



- Recall, we can find a list of a functions arguments with ?
- Any argument in the documentation followed by = carries a default value, that means if the user doesn't fill in that argument, the function can still run using the default (i.e., the argument is optional)
- We can also see the code inside the function by calling it's name without the ()

Writing a function



 We can write our own functions using assignment and the word "function"

```
name_of_func <- function(arguments_of_function) {
body_of_function }</pre>
```

- To the left of the assignment symbol we give it a name, inside the parentheses following the word "function" we list the arguments, and inside the curly brackets we write code
- A simple example

```
my_square <- function(x) {
x^2 }</pre>
```

Writing a function



- In R whatever the last line of the body to run is what the function will "return", or report to the user
- Exception! If the last line of a function body is an assignment the function will return nothing, be sure to avoid this so your function returns something

Writing a function



- Functions are useful because they allow you to repeat the same set of commands without copy and pasting code, which can be error prone and time consuming
- Functions can contain much more complex code than we are showing here, including calling other functions, looping, or if/then statements

Writing a function - Example



 Let's write a function that takes a numeric vector, removes any numbers lower than 1 or higher than 26, then returns the letter of the alphabet that matches each number in the resulting vector

Writing a function - Example



- So far, we have shown functions that take one argument, but functions can also take more than one argument
- This function returns the sum of two numbers a user passes it

```
sum_two <- function(a, b) {
a + b }</pre>
```

Writing a function - Practice

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 Arrange the follow lines to make a function named subtract_two that takes two numbers and subtracts the smaller number from the larger number (or returns zero if they are equal)

```
0
b - a
a - b
subtract_two <- function(a,b) {</li>
if (a == b) {
} else if (a < b) {</li>
} else if (b < a) {</li>
}
```

Test your new function on a couple values to make sure it works!

Writing Functions

Writing a function - Practice



- Write a function named square_df that takes a vector and returns a data frame with two columns where
 - the first column is named "original" and is the original vector
 - the second column is named "squared" and is each of the elements of the original vector squared
- Test your function on a couple vectors to make sure it works!

Writing a Function - Scoping



• We could write a function in the following manner:

```
x <- 1
return_one_two <- function(){
y <- 2
c(x, y)}</pre>
```

- Notice how the \times is written outside of the function even though it is used inside the function?
- But then, what would happen if we forgot to define x?

Writing a Function - Scoping



- In this case we would say $\mathbf x$ is **globally** scoped while $\mathbf y$ is **locally** scoped
- We would also say x is defined in the global environment while y
 is defined in the local environment
- Functions should be written with all variables locally scoped or passed as arguments to avoid errors
- How would we fix the previous function?

Scoping Practice



• We write a function which returns a number (chosen by the user) divided by 3, we initially write the function like so:

```
user_chosen_number <- 10
three <- 3
num_divide_three <- function() {
user_chosen_number / three }</pre>
```

 Rewrite the function so that the user passes their chosen number as an argument and three is locally scoped

Formal Arguments of a Function



- It is important to distinguish between the formal and actual arguments of a function
- Formal arguments are a property of the function

Arithmetic Mean

Description

Generic function for the (trimmed) arithmetic mean.

Usage

```
mean(x, ...)
## Default S3 method:
mean(x, trim = 0, na.rm = FALSE, ...)
```

Arguments

- x An R object. Currently there are methods for numeric/logical vectors and date, date-time and time interval objects. Complex vectors are allowed for trim = 0, only.
- trim the fraction (0 to 0.5) of observations to be trimmed from each end of x before the mean is computed. Values of trim outside that range are taken as the nearest endpoint.

 ${\tt na.rm}$ a logical value indicating whether NA values should be stripped before the computation proceeds.

... further arguments passed to or from other methods.

Calling Arguments of a Function



- It is important to distinguish between the formal and actual arguments of a function
- Actual or calling arguments can vary each time you call a function

```
> mean(x = 1:10)
[1] 5.5
> mean(x = 99:999)
[1] 549
```

 In the above examples, the calling arguments are 1:10 and 99:999 respectively

Default Arguments



- It is also possible to define default values for your function
- The default values will be used if the user doesn't specify a value

Default Arguments



```
# w/o default values
myFunc_10 <- function(a, b) {</pre>
  c(a, b)
> myFunc_10()
Error in myFunc_10(): argument "a" is missing, with no default
# with default values
myFunc_11 \leftarrow function(a = 1, b = 2) {
  c(a, b)
> myFunc_11()
[1] 1 2
```

Default Arguments



Function arguments in $\ensuremath{\mathbb{R}}$ can be defined in terms of other arguments

```
myFunc_12 \leftarrow function(a = 1, b = a * 2) {
  c(a, b)
> myFunc_12()
[1] 1 2
> myFunc_12(111)
[1] 111 222
> myFunc_12(99, 100)
[1] 99 100
```

Return Values



The last expression evaluated in a function becomes the return value

```
myFunc_18 <- function(xyz) {</pre>
  if (xyz < 10) {
    0
  } else {
    1.0
> myFunc_18(5)
[1] 0
> myFunc_18(10)
[1] 10
```

To return() or not to return()



- The last expression evaluated in a function is the return value
- You can always wrap the final expression in return() if you choose
- Using return() makes the code very slightly slower
- In simplistic functions, R programmers will typically omit return()
- In longer, more complicated functions, return() is often used when it makes the code easier to read

To return() or not to return()



```
# simple function, does not require a return()
myFunc_15 <- function(x) {</pre>
  x + 10
  a more complex function benefits visually from having return()
    but does not require return()
myFunc_18 <- function(xyz) {
  if (xyz < 10) {
    return(0)
  } else {
    return(10)
```

Functionals



- In addition to saving you time copying and pasting code, functions can be used with "functionals" to avoid loops
- Because loops (for, while, and repeat) are slow in R avoiding them is useful
- Also, functionals can make your code more readable

Functionals



The apply() family of functionals are often used in lieu of *for* loops, coming in a variety of flavors (not exhaustive)

Functional	Input	Output
apply()	Array/Matrix	Vector/Array
lapply()	Vector/List	List
sapply()	Vector/List	List
<pre>vapply()</pre>	Vector/List	Vector

There is also a functional do.call which is very flexible and can take and output most types. We will focus on the two most commonly used functionals, apply() and lapply()

apply()



- apply() is useful for applying the same function to every row or every column of a matrix
- e.g., apply(test_matrix, 1, mean) would find the mean of each row in test_matrix

apply()



- What if we want to include na.rm = T in our apply()?
- Just adding it doesn't work
- Adding it with the placeholder doesn't work
- What should we do?

apply()



- First option, make a custom function, say, na_mean() that does
 what we want
- Second option, make an anonymous or lambda function
 apply(test_matrix, 1, function(x) {mean(x, na.rm
 = T)})
- The anonymous function doesn't have a name, and can't be called again later, but this can be a useful method if you don't anticipate reusing the function

apply() practice



- Use apply() to get the standard deviation, removing NA's, from the test matrix
- You can either define a new function or use an anonymous function

lapply()



- lapply () takes and returns a list
- ullet This is useful to us because ${\mathbb R}$ considers data frames to be a type of list!
- Consider the following:



Assume you are given the following data frame

```
> myDataFrame_01
A B C D E G
1 1 6 1 5 -99 1
2 10 4 4 -99 9 3
3 7 9 5 4 1 4
4 2 9 3 8 6 8
5 1 10 5 9 8 6
6 6 2 1 3 8 5
```

Your objective is to replace all of the −99s with NAs



 You could—but shouldn't—iterate through each column manually, e.g.

```
myDataFrame_01$A[myDataFrame_01$A == -99] <- NA
myDataFrame_01$B[myDataFrame_01$B == -99] <- NA
...
myDataFrame_01$F[myDataFrame_01$F == -99] <- NA</pre>
```

Problems with Brute-Force Approaches



- It's easy to make copy-paste mistakes
- It will take you a really long time
- If you need to change the code later, you will have to change many lines of code rather than just a few



Let's write a function with the objective of replacing all -99s in a single column with NAs

```
fix99s_byCol <- function(myCol) {
  myCol[myCol == -99] <- NA
}</pre>
```

• Will the code above work as intended? Hint: no. Why not?



Let's write a function with the objective of replacing all -99s in a single column with NAs

```
fix99s_byCol <- function(myCol) {
  myCol[myCol == -99] <- NA
}</pre>
```

• Will the code above work as intended? Hint: no. Why not?



The following does work as intended:

```
fix99s_byCol <- function(myCol) {
  myCol[myCol == -99] <- NA
  myCol
}

myDataFrame_01$A <- fix99s_byCol(myDataFrame_01$A)
...

myDataFrame_01$F <- fix99s_byCol(myDataFrame_01$F)</pre>
```

- This reduces but doesn't eliminate the potential for errors
- There is no gain in efficiency (repetitive code is still required)
- We can instead use an lapply ()



```
fix99s_byCol <- function(myCol) {
 mvCol[mvCol == -991 <- NA
 myCol
> myDataFrame_02 <- lapply(myDataFrame_01, fix99s_byCol)
> str(myDataFrame_02)
List of 6
 $ A: num [1:6] 1 10 7 2 1 6
$ B: num [1:6] 6 4 9 9 10 2
 $ C: num [1:6] 1 4 5 3 5 1
 $ D: num [1:6] 5 NA 4 8 9 3
 $ E: num [1:6] NA 9 1 6 8 8
 $ F: num [1:6] 1 3 4 8 6 5
```

This almost worked...but not quite





Here are two ways to correct the previous function call so that it returns a data frame

```
> myDataFrame_03 <-
    as.data.frame(lapply(myDataFrame_01,
    fix99s_byCol))

> myDataFrame_01[] <- lapply(myDataFrame_01,
    fix99s_byCol)</pre>
```

Note, option 2 only works when replacing the old data frame with the new values, not for making a brand new data frame

lapply() practice



First, make a function that takes a column and changes all values
 4 or smaller to 0 and all values 5 or larger to 10 by filling in the
 blanks below:

```
big_small <- function(myCol) {
myCol[myCol < ___] <- ___
myCol[myCol > ___] <- ___
}</pre>
```

 Then, use your new function and lapply() to change myDataFrame_01 into myDataFrame_0_10, where all the numbers have been transformed to 0s and 10s

Lab



- Lab is on github, please submit to canvas for today's participation/attendance (as usual)
- Recall, you can request extra review through the canvas site (optional)
- If you prefer, you can also write an extra review request on a notecard (also optional)