Stat 260, Lecture 10, Programming, Pipes and Functions

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Load packages and datasets

```
library(tidyverse) # loads the %>% pipe
```

Reading

- Programming overview, Pipes and Functions: Intro to Part II, and Chapters 14 and 15 of the printed text, Chapters 17 to 19 of online text.
- Optional: Wickham (2014), Advanced R, Chapter 6, [http://adv-r.had.co.nz/Functions.html]

Programming in R

- We have already been programming in R, mostly by writing chunks of code that use tidyverse functions to do data visualization or wrangling.
- Now we discuss strategies for making our code easier to read and less prone to errors.
- One useful tool for this purpose is the forward pipe, %>%, which we'll discuss in more detail.
- A principle in programming is "Do not Repeat Yourself (DRY)", and writing functions can help us stay DRY.

Pipes

- The forward pipe %>% is implemented in the `magrittr` package, and is loaded when we load the tidyverse.
- Forward pipes are useful for combining a linear sequence of data processing steps, when we won't need the intermediate steps.
  - Tidyverse functions are typically named as actions, or verbs, and a linear sequence of such actions reads like a sentence.

We can use pipes too much

- However, we can take the idea too far and make the code difficult to debug.
- **Exercise** Debug the following code

```
read_csv("API_ILO_country_YU.csv") %>%
  gather(year,`Unemployment Rate`,`2010`:`2014`,convert=TRUE) %>%
  filter(str_detect(`Country Name`," \\(IDA.*\\)")) %>%
  mutate(`Country Name` =
           str_replace(`Country Name`,"", "\\(IDA.*\\)")) %>%
  select(-`Country Code`) %>%
  ggplot(aes(x=year,y=`Unemployment Rate`, color=`Country Name`)) %>%
  + geom_point() + geom_line()
```

Pipes are not good when there are multiple inputs

- Code may involve parallel computations that are assembled at the end.
- For example, suppose you need to read in two tibbles, manipulate each with actions like filter/gather/mutate, and then join them together.
  - We should save the tibbles to intermediate objects before joining them.

Other Tools from `magrittr`

- See the text's section – I don't use them myself
- Note: %$% is similar to the base R function `with()`

```
library(magrittr)
mtcars %$% cor(disp,mpg)
```

```
## [1] -0.8475514
```

```
with(mtcars,cor(disp,mpg))
```

```
## [1] -0.8475514
```

R functions – overview

- Encapsuling code in a function has several advantages:
  - can be used multiple times on different inputs
  - can compartmentalize computations and give them a name
  - can help you break a complicated task down into more manageable pieces.
- We will discuss:
  - when to write a function,
  - components of a function
  - writing pipeable functions,

When to write a function

- If you find yourself cutting and pasting the same code multiple times (more than twice, according to the text), then you should consider writing a function.
- See text for one example. Here is another.
- The `Boston` dataset in the `MASS` package includes data on house prices (`medv`) and characteristics of different neighborhoods in Boston [https://stat.ethz.ch/R-manual/R-devel/library/MASS/html/Boston.html]
- Certain kinds of statistical analyses of the relationship between `medv` and the other variables require that the other variables be standardized, by subtracting the mean values and dividing by the SD.
- I have made the dataset available in our Notes folder with the following

```
library(MASS)
write_csv(Boston,file="Boston.csv")
```

Boston dataset

```
Boston <- read_csv("Boston.csv")
dim(Boston)
```

```
## [1] 506  14
```

```
Boston
```

```
## # A tibble: 506 x 14
##       crim    zn indus  chas   nox    rm   age   dis   rad   tax ptratio
##      <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>   <dbl>
##  1 0.00632  18    2.31     0 0.538  6.58  65.2  4.09     1   296    15.3
##  2 0.0273    0    7.07     0 0.469  6.42  78.9  4.97     2   242    17.8
##  3 0.0273    0    7.07     0 0.469  7.18  61.1  4.97     2   242    17.8
##  4 0.0324    0    2.18     0 0.458  7.00  45.8  6.06     3   222    18.7
##  5 0.0690    0    2.18     0 0.458  7.15  54.2  6.06     3   222    18.7
##  6 0.0298    0    2.18     0 0.458  6.43  58.7  6.06     3   222    18.7
##  7 0.0883   12.5  7.87     0 0.524  6.01  66.6  5.56     5   311    15.2
##  8 0.145    12.5  7.87     0 0.524  6.17  96.1  5.95     5   311    15.2
##  9 0.211    12.5  7.87     0 0.524  5.63 100    6.08     5   311    15.2
## 10 0.170    12.5  7.87     0 0.524  6.00  85.9  6.59     5   311    15.2
## # ... with 496 more rows, and 3 more variables: black <dbl>, lstat <dbl>,
## #   medv <dbl>
```

Standardize columns of Boston

- You can standardize the first column of Boston with

```
Boston$crim <- (Boston$crim - mean(Boston$crim, na.rm=TRUE))/
  sd(Boston$crim,na.rm=TRUE)
```

- Now cut-and-paste 12 times to standardize the remaining predictors of medv

```
Boston$zn <- (Boston$zn - mean(Boston$zn, na.rm=TRUE))/
  sd(Boston$zn,na.rm=TRUE)
Boston$indus <- (Boston$indus - mean(Boston$indus, na.rm=TRUE))/
  sd(Boston$indus,na.rm=TRUE)
# Etc.
```

- Not only is this tedious, it is error-prone. Plus, we will need to do the same operation on other datasets.

A standardization function

- The following function standardizes a vector.
  - We'll learn more about the components of a function in the slides to follow.

```
standardize <- function(x) {
  (x - mean(x,na.rm=TRUE))/sd(x,na.rm=TRUE)
}
Boston$crim <- standardize(Boston$crim)
Boston$zn <- standardize(Boston$zn)
# Etc
```

- This has reduced the amount of code and chances for cut-and-paste errors.

Components of a function

- In R, functions are objects with three essential components:
  - the code inside the function, or `body`,
  - the list of arguments to the function, and
  - a data structure called an `environment` which is like a map to the memory locations of all objects defined in the function.

Example function

```
f <- function(x) {
  x^2
}
f
```

```
## function(x) {
##   x^2
## }
```

The function arguments

- These are the arguments to the function.
- Function arguments can have default values, as in:

```
f <- function(x=0) { x^2}
```

- **Exercise** Re-write our `standardize()` function to have an additional argument `na.rm`, set to `TRUE` by default.

Argument defaults

- Argument defaults can be defined in terms of other arguments:

```
f <- function(x=0,y=3*x) { x^2 + y^2 }
f()
```

```
## [1] 0
```

```
f(x=1)
```

```
## [1] 10
```

```
f(y=1)
```

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

Argument matching

- When you call a function, the arguments are matched first by name, then by "prefix" matching and finally by position:

```
f <- function(firstarg,secondarg) {
  firstarg^2 + secondarg
}
f(firstarg=1,secondarg=2)
```

```
## [1] 3
```

```
f(s=2,f=1)
```

```
## [1] 3
```

```
f(2,f=1)
```

```
## [1] 3
```

```
f(1,2)
```

```
## [1] 3
```

The function body

- This is the code we want to execute.
- When the end of a function is reached, the value of the last line is returned.
  - If you prefer, you can end a function with `return()` to signal the function's returned value.

```
f <- function(x=0) { x^2}
f <- function(x=0) { return(x^2)}
```

Control Flow

- Code within a function is not always executed linearly from start to end.
- We may need to execute different code chunks depending on the function inputs.
- We may need to repeat certain calculations, or loop.
- Such constructs are called control flow.
- We'll touch on some of the basics.

if and if-else

- `if` tests a condition and executes code if the condition is true. Optionaly, can couple with an `else` to specify code to execute when condition is false.

```
if("cat" == "dog") {
  print("cat is dog")
} else {
  print("cat is not dog")
}
```

```
## [1] "cat is not dog"
```

Conditions require TRUE or FALSE

- If not, you will get a warning or error.

```
if(c(TRUE,FALSE)) {} # Throws the following warning:
#Warning message:
#In if (c(TRUE, FALSE)) { :
#  the condition has length > 1 and only the first element will be used
if(NA) {} # Throws the following error:
#Error in if (NA) { : missing value where TRUE/FALSE needed
```

for loops

- Example: In the following, `1:nreps` is the "index set".

```
n <- 10; nreps <- 100; x <- vector(mode="numeric",length=nreps)
for(i in 1:nreps) {
  x[i] <- mean(rnorm(n))
}
summary(x)
```

```
##     Min.  1st Qu.   Median     Mean  3rd Qu.     Max.
## -1.14038 -0.18771 -0.03117 -0.03749  0.13917  0.92148
```

```
print(i)
```

```
## [1] 100
```

- **Exercise** Write a function `standardize_tibble()` that loops through the columns of a `tibble` and standardizes each with your `standardize()` function. Hints: If `tt` is a tibble, `ncol(tt)` is the number of columns, and `1:ncol(tt)` is an appropriate index set. If `tt` is a tibble, `tt[[1]]` is the first column.

`for` loop index set

- Index sets are sometimes the indeces of a vector, and can also be the elements of the vector.

```
ind <- c("cat","dog","mouse")
for(i in seq_along(ind)) {
  print(paste("There is a",ind[i],"in my house"))
}
```

```
## [1] "There is a cat in my house"
## [1] "There is a dog in my house"
## [1] "There is a mouse in my house"
```

```
for(i in ind) {
  print(paste("There is a",i,"in my house"))
}
```

```
## [1] "There is a cat in my house"
## [1] "There is a dog in my house"
## [1] "There is a mouse in my house"
```

- We'll learn more about this kind of iterating in the chapter **Iteration with `purrr`**.

while loops

- Use a `while` loop when you want to continue until some logical condition is met.

```
set.seed(1)
# Number of coin tosses until first success (geometric distn)
p <- 0.1; counter <- 0; success <- FALSE
while(!success)  {
  success <- as.logical(rbinom(n=1,size=1,prob=p))
  counter <- counter + 1
}
counter
```

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

break

- `break` can be used to break out of a `for` or `while` loop.

```
for(i in 1:100) {
  if(i>3) break
  print(i)
}
```

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

The function environment

- The environment within a function is like a map to the memory locations of all its variables.
- The function arguments are "passed by value", meaning that a copy is made and stored in the function's environment.
- Variables created within the function are also store in its environment

```
f <- function(x) {
  y <- x^2
  ee <- environment() # Returns ID of environment w/in f
  print(ls(ee)) # list objects in ee
  ee
}
f(1) # function call
```

```
## [1] "ee" "x"  "y"
```

```
## <environment: 0x7f924dcb0130>
```

Enclosing environments

- Our function `f` was defined in the global environment, `.GlobalEnv`, which "encloses" the environment within `f`.
- If `f` needs a variable and can't find it whithin `f`'s environment, it will look for it in the enclosing environment, and then the enclosing environment of `.GlobalEnv`, and so on.
- The `search()` function lists the heirarchy of environments that enclose `.GlobalEnv`.

```
search()
```

```
##  [1] ".GlobalEnv"        "package:magrittr"  "package:forcats"
##  [4] "package:stringr"   "package:dplyr"     "package:purrr"
##  [7] "package:readr"     "package:tidyr"     "package:tibble"
## [10] "package:ggplot2"   "package:tidyverse" "package:stats"
## [13] "package:graphics"  "package:grDevices" "package:utils"
## [16] "package:datasets"  "package:methods"   "Autoloads"
## [19] "package:base"
```

- To facilitate this search, each environment includes a pointer to its enclosing environment.

Exercise

- The following code chunk is typed into the R Console.
  - What is the output of the function call `f(5)`?
  - What is the enclosing environment of `f()`?
  - What is the enclosing environment of `g()`?
  - What search order does R use to find the value of `x` when it is needed in `g()`?

```
x <- 1
f <- function(y) {
  g <- function(z) {
    (x+z)^2
  }
  g(y)
}
```

Writing readable code

- The more your code does, the harder it is for others to read.
  - Here "others" includes you some time in the future.
  - Wickham says we should write code that future-you can understand, because past-you doesn't answer emails.
- See the **Functions are for Humans and Computers** section of text for tips on writing readable code.

Other reasons to write functions

- Functions can be used to prevent repetition, but even if used only once they can improve code readability.
  - For example, you are writing a function `func()` that computes a statistic `mystat` that takes 10 lines of code to calculate.
  - The rest of your function is only 5 lines.
  - Write a function called `mystat()` and call it from `func()`.
  - If you define `func()` first, it will be easier to document `mystat()`.
- Writing code in a top-down way is like writing an outline for an essay and then filling in the details.
  - The main function is the outline.
  - the sub-functions are the details of each topic.

Exercise

- Create an R script that first defines `standardize_tibble()` and then `standardize()`. In `standardize()`, replace `mean()` by a function `center()` and `sd()` by a function `spread()`, where `center()` and `spread()` are functions that you write to compute the mean and SD using only the `sum()` function. `center()` and `spread()` should remove missing values by default.

R packages

- Use the `library()` command to load packages.
- When we load a package it is inserted in position 2 of the search list, just after `.GlobalEnv`.

```
# install.packages("hapassoc")
library(hapassoc)
search()
```

```
##  [1] ".GlobalEnv"        "package:hapassoc"  "package:magrittr"
##  [4] "package:forcats"   "package:stringr"   "package:dplyr"
##  [7] "package:purrr"     "package:readr"     "package:tidyr"
## [10] "package:tibble"    "package:ggplot2"   "package:tidyverse"
## [13] "package:stats"     "package:graphics"  "package:grDevices"
## [16] "package:utils"     "package:datasets"  "package:methods"
## [19] "Autoloads"         "package:base"
```

Detaching packages

- Detach a package from the search list with `detach()`

```
detach("package:hapassoc")
search()
```

```
##  [1] ".GlobalEnv"        "package:magrittr"  "package:forcats"
##  [4] "package:stringr"   "package:dplyr"     "package:purrr"
##  [7] "package:readr"     "package:tidyr"     "package:tibble"
## [10] "package:ggplot2"   "package:tidyverse" "package:stats"
## [13] "package:graphics"  "package:grDevices" "package:utils"
## [16] "package:datasets"  "package:methods"   "Autoloads"
## [19] "package:base"
```

Package namespaces

- Package authors create a list of objects that will be visible to users when the package is loaded. This list is called the package namespace.
- You can access functions in a package's namespace without loading the package using the `::` operator.

```
set.seed(321)
n<-30; x<-(1:n)/n; y<-rnorm(n,mean=x); ff<-lm(y~x)
car::sigmaHat(ff)
```

```
## [1] 0.926726
```

- Doing so does not add the package to the search list.