# Pipes, Functions, Conditionals

### 2023-07-04

# Learning Objectives

- Piping.
- Creating Functions.
- Chapters 17 through 19 in RDS.
- Magrittr Overview.

# Magrittr and the %>%

- When doing multiple operations to a dataset, we can:
  - create intermediate objects
  - overwrite the original object
  - nest commands
  - use pipe
- Suppose we want to calculate the geometric mean of a vector of positive numbers. Then the numerically stable steps are
  - 1. Take the log of all of the elements of the vector.
  - 2. Average these logged elements.
  - 3. Exponentiate this average.
- We'll create some random data using rnorm() to look at these steps:

```
set.seed(1)
  x <- abs(rnorm(100))
  head(x)</pre>
```

## [1] 0.6264538 0.1836433 0.8356286 1.5952808 0.3295078 0.8204684

$$\left(\prod_{i=1}^{n} a_{i}\right)^{\frac{1}{n}} = \sqrt[n]{a_{1} \cdot a_{2} \cdot \cdot \cdot a_{n}}, b$$
 Geometric mean

```
gm <- prod(x)^(1/100)
gm
```

## [1] 0.4640516

$$\left(\Pi_{i=1}^n \, a_i\right)^{\frac{1}{n}} = \exp(\frac{1}{n} \sum_{i=1}^n \, \ln a_i)$$
Arithmatic Mean of Logarithm

• Create intermediate objects

```
log_x <- log(x)
mean_lx <- mean(log_x)
l_geo_mean <- exp(mean_lx)
l_geo_mean</pre>
```

## [1] 0.4640516

• Nest functions

```
geo_mean <- exp(mean(log(x)))
geo_mean</pre>
```

## [1] 0.4640516

• Use pipe: we need to install and load tidyverse.

```
library(tidyverse)
x %%
log() %>%
mean() %>%
exp() ->
geo_mean
geo_mean
```

## [1] 0.4640516

• The term on the left of the pipe becomes the *first* argument of the function on the right of the pipe. Thus, the following two expressions are the exact same:

```
4 %%
  log(base = 2)

## [1] 2

# Without pipe
log(4, base = 2)
```

## [1] 2

• Note that the output is only placed in the *first* argument of the next function. We can use the pipe while specifying other arguments..

```
x %>%
  log() %>%
  mean(na.rm = TRUE) %>%
  exp() ->
  geo_mean
geo_mean
```

## [1] 0.4640516

• Exercise: Using the pipe, create a vector of 100 random draws from the normal distribution with standard deviation 10, sort this vector in increasing order, calculate the lagged differences between these elements, average these lagged differences, then round this average to the nearest tenth decimal place.

Useful functions: diff(), rnorm(), sort(), round(), mean().

## **Functions**

- When to write a function and why to write a function
- Steps to creating a function:
  - 1. figure out the logic in a simple case
  - 2. name it something meaningful usually a verb
  - 3. list the inputs inside function(x,y,z)
  - 4. place code for function in a {} block
  - 5. test your function with some different inputs

- 6. add error-checking of inputs
- Coding standards

```
1. Save as text file
```

- 2. Indent code
- 3. Limit width of code (80 columns?)
- 4. Limit the length of individual functions
- 5. Frequent comments

```
add_two <- function(a, b) {
  return(a + b) # adding two numbers
}
add_two(2, 4)</pre>
```

## [1] 6

```
adding <- function(a,b) {
  a + b
}
adding(2, 4)</pre>
```

#### ## [1] 6

• Example from our book follows.

```
df <- tibble::tibble(
    a = rnorm(10),
    b = rnorm(10),
    c = rnorm(10),
    d = rnorm(10)
)

df$a <- (df$a - min(df$a, na.rm = TRUE)) /
    (max(df$a, na.rm = TRUE) - min(df$a, na.rm = TRUE))

df$b <- (df$b - min(df$b, na.rm = TRUE)) /
    (max(df$b, na.rm = TRUE) - min(df$a, na.rm = TRUE))

df$c <- (df$c - min(df$c, na.rm = TRUE)) /
    (max(df$c, na.rm = TRUE) - min(df$c, na.rm = TRUE))

df$d <- (df$d - min(df$d, na.rm = TRUE)) /
    (max(df$d, na.rm = TRUE) - min(df$d, na.rm = TRUE))</pre>
```

df

```
## # A tibble: 10 x 4
##
          a
                 b
                        С
                                d
##
       <dbl> <dbl> <dbl>
                            <dbl>
   1 0.108  0.0104  0.0868  0.772
   2 0.356 0.132 1
                          0.458
##
##
   3 0
            1.45
                   0.231 1
##
   4 0.399 0
                   0.248
                          0.00873
                          0.891
##
   5 0.0957 0.310 0.287
##
   6 1
            0.180 0.689
                          0
##
   7 0.608 0.231 0.300 0.597
## 8 0.680 0.259 0.318 0.488
## 9 0.484 0.799 0
                          0.428
## 10 0.968 0.330 0.177 0.715
```

```
df$a <- (df$a - min(df$a, na.rm = TRUE)) /</pre>
      (max(df$a, na.rm = TRUE) - min(df$a, na.rm = TRUE))
df$a
  [1] 0.1084885 0.3558488 0.0000000 0.3991288 0.0957121 1.0000000 0.6077305
##
## [8] 0.6799678 0.4835721 0.9682209
my_fun(c)
  • How many inputs does each line have?
     x \leftarrow df$a
     (x - min(x, na.rm = TRUE)) / (max(x, na.rm = TRUE) - min(x, na.rm = TRUE))
     ## [1] 0.1084885 0.3558488 0.0000000 0.3991288 0.0957121 1.0000000 0.6077305
     ## [8] 0.6799678 0.4835721 0.9682209
     # get rid of duplication
     rng <- range(x, na.rm = TRUE)</pre>
     (x - rng[1]) / (rng[2] - rng[1])
         [1] 0.1084885 0.3558488 0.0000000 0.3991288 0.0957121 1.0000000 0.6077305
     ## [8] 0.6799678 0.4835721 0.9682209
     # make it into a function and test it
     rescale01 <- function(x) {
       rng <- range(x, na.rm = TRUE)</pre>
       (x - rng[1]) / (rng[2] - rng[1])
     # inside function find the values of x
     rescale01(c(0, 5, 10))
     ## [1] 0.0 0.5 1.0
     rescale01(c(-10, 0, 10))
     ## [1] 0.0 0.5 1.0
     rescale01(c(1, 2, 3, NA, 5))
     ## [1] 0.00 0.25 0.50
                             NA 1.00
     df$a <- rescale01(df$a)
     df$b <- rescale01(df$b)</pre>
     df$c <- rescale01(df$c)</pre>
     df$d <- rescale01(df$d)</pre>
  • Now, if we have a change in requirements, we only have to change it in one place. For instance, perhaps
     we want to handle columns that have Inf as one of the values.
     x \leftarrow c(1:10, Inf)
     rescale01(x)
     ## [1] 0
                        0
                                     0
                                         0
                                             0 0
                                                      0 NaN
     rescale01 <- function(x) {
       rng <- range(x, na.rm = TRUE, finite = TRUE)</pre>
       # finite: indicates if all non-finite elements should be omitted
       (x - rng[1]) / (rng[2] - rng[1])
     }
     rescale01(x)
```

```
## [1] 0.0000000 0.1111111 0.2222222 0.3333333 0.4444444 0.5555556 0.6666667
## [8] 0.7777778 0.8888889 1.0000000 Inf
```

- Do's and do not's of function naming:
  - pick either snake case or camelCase but don't use both
  - meaningful names (preferably verbs)
  - for a family of functions, start with the same word
  - try not to overwrite common functions or variables
  - use lots of comments in your code, particularly to explain the "why" of your code or to break up your code into sections using something like # load data ------

#### **Function Documentation**

- This is a DCG opinion section.
- It is good practice to always precede a function with documentation on
  - 1. What does it do?
  - 2. What are the inputs?
  - 3. What are the outputs?
- Use comments to do this.
- Here is an example of me re-writing the diff() function.

```
# Calculates a vector of lags from a numeric vector
#
# x: a vector of numerics to be lagged
#
# returns: The lags of x
diff2 <- function(x) {
    stopifnot(is.numeric(x))
    lagvec <- x[2:length(x)] - x[1:(length(x) - 1)]
    return(lagvec)
}
diff2(c(1, 5, 21))
## [1] 4 16
diff(c(1, 5, 21))
## [1] 4 16
# Note: & and && indicate logical AND and | and || indicate logical OR. The shorter
# form performs elementwise comparisons in much the same way as arithmetic operators.
# The longer form evaluates left to right examining only the first element of each vector.</pre>
```

- Exercise: Re-write the function range(), including documentation at the top of the function. Use functions: min(), max()
- Exercise: Write both\_na(), a function that takes two vectors of the same length and returns the number of positions that have an NA in both vectors. Include documentation. Useful functions: is.na(), sum(), logical operators.
- Exercise: Write pos\_na(), a function that takes two vectors of the same length and returns the positions that have an NA in both vectors. Include documentation. Useful functions: is.na(), which(), logical operators.

Without if function

• Exercise: Read the source code for each of the following three functions, puzzle out what they do, and then brainstorm better names. Then add documentation to each function.

```
f1 <- function(string, prefix) {</pre>
       substr(string, 1, nchar(prefix)) == prefix
     }
     f2 <- function(x) {
       if (length(x) <= 1) return(NULL)</pre>
       x[-length(x)]
     f3 <- function(x, y) {
       rep(y, length.out = length(x))
   • Let us see what does function f1
f1 <- function(string, prefix) {</pre>
      substr(string, 1, nchar(prefix)) == prefix
f1("Hi its ME", "Hi")
## [1] TRUE
f1("Hi its ME", "Hey")
## [1] FALSE
f1("Hi its ME", "Hi its")
## [1] TRUE
f1("Wednesday", "Wed")
## [1] TRUE
f1("Wednesday", "w")
## [1] FALSE
f1(c(1, 2, 3, 3), c(1, 3, 2, 4))
## [1] TRUE FALSE FALSE FALSE
f1(c(1, 2, 3, 3), c(1, 3))
## [1] TRUE FALSE FALSE TRUE
f1(c(1,3), c(1, 2, 3, 3))
## [1] TRUE FALSE FALSE TRUE
   • Let us see what does function f2
    f2 <- function(x) {
      if (length(x) <= 1) return(NULL)</pre>
      x[-length(x)]
    }
```

## NULL

f2(c())

```
f2(c(5))
## NULL
f2(c(1,0, 25, 27))
## [1] 1 0 25
f2(c("a", "b", "c", "d"))
## [1] "a" "b" "c"
  • Let us see what does function f3
f3 <- function(x, y) {
      rep(y, length.out = length(x))
}
f3(c(1, 2, 3, 4), c(11, 12, 13, 14, 15, 16, 17))
## [1] 11 12 13 14
f3(c(11, 12, 13, 14, 15, 16, 17), c(1, 2, 3, 4))
## [1] 1 2 3 4 1 2 3
f3(c(1,2, 3, 7), (3:9))
## [1] 3 4 5 6
f3((3:9), c(1,2, 3, 7))
## [1] 1 2 3 7 1 2 3
```

# Conditional Execution

• Conditional if-then statements are useful to evaluate code only if certain conditions are met. The syntax is:

```
if (condition) {
    # code executed when condition is TRUE
} else {
    # code executed when condition is FALSE
}

# input: x a numeric
# output: TRUE if x > 0, FALSE if x <= 0
is_positive <- function(x) {
    if (x > 0) {
        pos_logic <- TRUE
    } else {
        pos_logic <- FALSE
    }
    return(pos_logic)
}

is_positive(10)

## [1] TRUE
is_positive(-1)</pre>
```

```
## [1] FALSE
is_positive(0)
## [1] FALSE
```

• The condition for an if statement must be a single logical element, not a vector. For this reason, if combining them, use | | which returns TRUE at first occurrence of TRUE or && which return FALSE at first occurrence of FALSE. Can use any() or all() to collapse a logical vector. Similarly, for equality, use identical() or dplyr::near(). You can use ==, but it is vectorized and might give you errors if you use it with a vector.

## Multiple conditions

• You may chain multiple if statements or use switch

```
do_op <- function(x, y, op) {</pre>
  switch(op,
     plus = x + y,
     minus = x - y,
     times = x * y,
     divide = x / y,
     stop("Unknown op!")
 )
}
do_op(2,4,"plus")
## [1] 6
do_op(2,4,"divide")
## [1] 0.5
do_{op}(2,4,1)
## [1] 6
do op(2,4,"mod")
## Error in do_op(2, 4, "mod"): Unknown op!
# note that we had to change the chunk option to error=TRUE in
# order to test our error output
```

- Some code style standards:
  - Put function code and if statement code inside {}
  - opening { never on own line and always followed by a new line
  - closing } on own line unless followed by else
  - use indentation
- Exercise: Implement a fizzbuzz() function. It takes a single number as input. If the number is divisible by three, it returns "fizz". If it's divisible by five it returns "buzz". If it's divisible by three and five, it returns "fizzbuzz". Otherwise, it returns the number. Make sure you first write working code before you create the function.

### **Function inputs**

• First inputs listed should be the data to compute on. Later arguments should be parameters that control the details of the computation and should generally have defaults specified. The default should

be the most often used value unless there is a safety reason to do differently. Look at some functions and see what the defaults are. Try log(), mean(), t.test(), cor().

- When calling a function, omit the names of the data arguments, but specify the names of the parameters. White space: use space after , and around = or other operators.
- Use descriptive names for your arguments with the following exceptions:

```
x, y, z: vectors.
w: a vector of weights.
df: a data frame.
i, j: numeric indices (typically rows and columns).
n: length, or number of rows.
p: number of columns.
```

# Idiot proof your functions

• Check validity of inputs and use stop() to output an error message

```
wt_mean <- function(x, w) {
  if (length(x) != length(w)) {
    stop("`x` and `w` must be the same length", call. = TRUE)
    # call.: logical, indicating if the call should become part of the error message.
}
  sum(w * x) / sum(w)
}
wt_mean(1:4,1:2)
## Error in wt_mean(1:4, 1:2): `x` and `w` must be the same length
wt_mean(3:1,1:2)
## Error in wt_mean(3:1, 1:2): `x` and `w` must be the same length
wt_mean(5:4,1:2)
## [1] 4.333333</pre>
```

• Another option is to use stopifnot() which checks that each argument is true and issues a generic error message if there is a problem

```
#na. rm is FALSE, then it returns NA from the calculation being done
#on the entire row or column.If rm is TRUE, the function skips over any NA values.
wt_mean <- function(x, w, na.rm = FALSE) {
    stopifnot(is.logical(na.rm), length(na.rm) == 1)
    stopifnot(length(x) == length(w))

if (na.rm) {
    #miss will consist of TRUE or FALSE values
    miss <- is.na(x) | is.na(w)
    x <- x[!miss]
    w <- w[!miss]
}

sum(w * x) / sum(w)
}
#Comments/Trials:</pre>
```

```
#wt_mean(1:6, 6:1, na.rm = "foo")
#wt_mean(1:6, 6:1, na.rm = FALSE)
#wt_mean(5:4, 1:2, na.rm = FALSE)
#Try the following to understand the if statement
\#x \leftarrow c(NA, 5, NA)
\#is.na(x)
#w <- c(NA,3,2)
\#is.na(w)
\#miss \leftarrow is.na(x) \mid is.na(w)
\#x \leftarrow x[!miss]
#w <- w[!miss]
#wt mean(x, w, na.rm = FALSE)
#Another Example
\#x \leftarrow c(1,5,2)
#w \leftarrow c(4,3,1)
\#miss \leftarrow is.na(x) \mid is.na(w)
\#x \leftarrow x[!miss]
#w <- w[!miss]
#wt_mean(x, w, na.rm = FALSE)
```

# **Environment**

• R will look for functions and variables in the current environment before it looks elsewhere. This allows you to use variables in the environment that are not explicitly in your function or to overwrite things that may be general functions in R with local versions. However, in general, this is not a good idea.