Probability - Tidyverse 2

Reducing duplication - writing functions and introducton to purrr

Xiaolong Yang

Introduction to Quantitative Social Science

University of Tokyo

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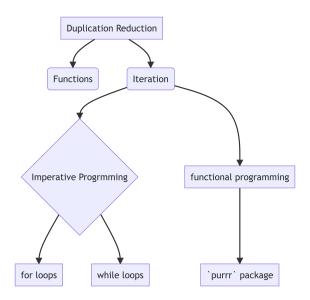
Today's Game Plan

- reducing duplication: functions
- purrr package
 - purrr: map_df() (introduced in Chapter 6: Probability (sections 6.3-6.4))
- i Today's in-class assignment: intrade-prob

Section 1

Functions

Landscape of Duplication Reduction in R



What does this code chunk do?

```
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>
```

• re-scaling each column to have a range from 0 - 1

```
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>
```

- re-scaling each column to have a range from 0 1
- But there is a mistake!!

```
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>
```

- re-scaling each column to have a range from 0 1
- But there is a mistake!!
 - df\$b: did not change a to b

```
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>
```

What if we have a function?

```
df$a <- rescale01(df$a)
df$b <- rescale01(df$b)
df$c <- rescale01(df$c)
df$d <- rescale01(df$d)</pre>
```

Advantages of functions over copy-paste

- easier to see the intent of your code: eyes on **difference** not **similarity**
- easier to respond to changes in requirements
- fewer bugs (i.e. updating a variable name in one place, but not in another).

You should consider writing a function whenever you've copied and pasted a block of code more than twice (i.e. you now have three copies of the same code).

3 key steps to create a function

pick a name for the function

square

3 key steps to create a function

- pick a name for the function
- ② list the inputs, or arguments, to the function inside function

```
square <- function(x) {}</pre>
```

3 key steps to create a function

- 1 pick a name for the function
- Iist the inputs, or arguments, to the function inside function
- place the code you have developed in body of the function

It's easier to start with working code and turn it into a function; it's harder to create a function and then try to make it work

```
square <- function(x) {
  x^2
}
square(13)</pre>
```

[1] 169

Function arguments

- data arguments: supplies the data to compute on
- details arguments: supplies arguments that control the details of the computation
- lm() as an example



```
function (formula, data, subset, weights, na.action, method =
  model = TRUE, x = FALSE, y = FALSE, qr = TRUE, singular.ok =
  contrasts = NULL, offset, ...)
{
    ...
    ...
    ...
}
```

Conditional execution

• an if statement allows you to conditionally execute code

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

Multiple conditions

• chain multiple if statements together

```
if (condition 1) {
    # do this if TRUE
} else if (condition 2) {
    # do that if TRUE
} else {
    # do something else if FALSE
}
```

Good practices

Functions are for humans and computers

- if and function should always be followed by {}
 - {: never go on its own line
 - }: always go on its own line (unless followed by else)

```
# Good
if (y == 0) {
  log(x)
} else {
  y ^ x
}

y ^ x
}
# Bad
if (y == 0) {
  log(x)
} # with else in this case
else {
  y ^ x
}
```

Good practices

Functions are for humans and computers

- function names: verbs
- argument names: nouns
- use inline code to explain the "why"
 - avoid the "what" or "how"

```
# Good # Bad
input_select() f()
input_checkbox() my_awesome_function()
```

Section 2

Brief introduction to purrr

Overview



Figure 1: purrr

- R as a functional programming (FP) language
- purrr provides complete and consistent tools for working with functions and different data types (vectors) → enhances R's efficiency
 - \bullet the family of map() function \to replace many for loops with succinct code
- purrr **cheatsheet**

purrr package: map_df()

- transforms the input by applying a function to each variable of a data frame or tibble (each element of a list or atomic vector)
- returns a data frame/tibble
- arguments
 - .x = a data frame/tibble (list or atomic vector)
 - .f = a function

```
map_df(.x, .f)
```

purrr package: map_df()

Example 1

[1] "data.frame"

purrr package: map_df()

Example 2

```
FLVoters %>%
  map(unique) %>%
  map_df(length)
```

Summary

What we learnt

- how to write functions
 - how to communicate functions by following good practices
- purrr (map_df())

Future Game Plan

- data types in R: vector
- reducing duplication: iteration
- new functions in Chapter 7: Uncertainty

Source

- Quantitative Social Science: An Introduction in tidyverse
- R for Data Sciene
- Advanced R