# Functional programming in R

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You'll learn some of the powerful programming tools provided by purrr, one of the tidyverse core packages.

```
library(tidyverse)
```

## What is functional programming?

In computer science, functional programming is a programming paradigm—a style of building the structure and elements of computer programs—that treats computation as the evaluation of mathematical functions and avoids changing-state and mutable data.

To put is simple, it is a programming style without side effects.

```
f <- function(x) {
    x$a <- 2
    x
}

x <- list(a = 1)
isTRUE(all.equal(f(x), x))</pre>
```

## [1] FALSE

```
def f(x):
    x["a"] = 2
    return x

x = {"a": 1}
f(x) == x
```

## True

# Is R a functional language?

R, at its heart, is a functional programming language. Most base R functions are pure, with a few notable exceptions:

- library() which loads a package, and hence modifies the search path.
- setwd(), Sys.setenv(), Sys.setlocale() which change the working directory, environment variables, and the locale, respectively.
- plot() and friends which produce graphical output.
- write(), write.csv(), saveRDS(), etc. which save output to disk.
- options() and par() which modify global settings.

- S4 related functions which modify global tables of classes and methods.
- Random number generators which produce different numbers each time you run them

Almost everything else in R are functions.

```
x <- 1
y <- 2
x + y
## [1] 3
+(x, y)
## [1] 3
a <- 3:5
a[<mark>3</mark>]
## [1] 5
`[`(a, 3)
## [1] 5
z <- 1
<-^{(z, 1)}
for (i in 1:10) print(i)
## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
## [1] 7
## [1] 8
## [1] 9
## [1] 10
`for`(i, 1:10, print(i))
## [1] 1
## [1] 2
## [1] 3
## [1] 4
## [1] 5
## [1] 6
## [1] 7
## [1] 8
## [1] 9
## [1] 10
```

We could even change the body of previously defined functions

```
add <- function(x, y) x + y
add(3, 1)
## [1] 4
body(add) <- quote(x - y)</pre>
add(3, 1)
## [1] 2
```

## Revisit for loops

Suppose we want to compute the sample medians for each column of the data frame

```
df <- tibble(</pre>
  a = rnorm(10),
  b = rnorm(10),
  c = rnorm(10),
  d = rnorm(10)
```

```
A naive way to do so
median(df$a)
## [1] -0.4432872
median(df$b)
## [1] -0.09424343
median(df$c)
## [1] 0.2235735
median(df$d)
```

```
## [1] 0.06092786
```

```
# a base R for loop
output <- vector("double", ncol(df)) # 1. output</pre>
for (i in seq_along(df)) { # 2. sequence
  output[[i]] <- median(df[[i]]) # 3. body</pre>
output
```

```
## [1] -0.44328719 -0.09424343 0.22357347 0.06092786
```

Sometimes we would like to modify existing objects

```
rescale01 <- function(x) {
    rng <- range(x, na.rm = TRUE)
        (x - rng[1]) / (rng[2] - rng[1])
}

df$a <- rescale01(df$a)
df$b <- rescale01(df$b)
df$c <- rescale01(df$c)
df$d <- rescale01(df$d)

# or simply
for (i in seq_along(df)) {
    df[[i]] <- rescale01(df[[i]])
}</pre>
```

## For loops vs functionals

The pattern of looping over a vector, doing something to each element and saving the results is so common that the purrr package provides a function map to do it.

The basic syntax for map is

## [1] 0.742106

## [1] 0.6425789

## ## \$d

```
map(ITERATABLE, A_FUNCTION)
```

ITERATABLE's are objects like vectors, list and data frame.

```
map(df, median)
## $a
## [1] 0.5170176
##
## $b
## [1] 0.5628878
##
## $c
## [1] 0.742106
##
## $d
## [1] 0.6425789
df %>% map(median)
## $a
## [1] 0.5170176
##
## $b
## [1] 0.5628878
##
## $c
```

```
\# the input is not necessary to be a data frame
c(9, 16, 25) %>% map(sqrt)
## [[1]]
## [1] 3
##
## [[2]]
## [1] 4
##
## [[3]]
## [1] 5
fab_norm <- function(p) {</pre>
 x \leftarrow c(1, 1, 2, 3, 5, 8)
  (sum(abs(x)^p))^(1/p)
fab_norm(1)
## [1] 20
fab_norm(1.5)
## [1] 12.43168
fab_norm(2)
## [1] 10.19804
# but fab\_norm(c(1, 1.5, 2)) does not work
c(1, 1.5, 2) %>% map(fab_norm)
## [[1]]
## [1] 20
##
## [[2]]
## [1] 12.43168
##
## [[3]]
## [1] 10.19804
mean(df\$a, trim = 0.5)
## [1] 0.5170176
mean(df$b, trim = 0.5)
## [1] 0.5628878
```

```
mean(df$c, trim = 0.5)
## [1] 0.742106
mean(df$d, trim = 0.5)
## [1] 0.6425789
# pass additional arguments
map(df, mean, trim = 0.5)
## $a
## [1] 0.5170176
## $b
## [1] 0.5628878
##
## $c
## [1] 0.742106
##
## $d
## [1] 0.6425789
# equivalently
df %>% map(mean, trim = 0.5)
## $a
## [1] 0.5170176
##
## $b
## [1] 0.5628878
##
## $c
## [1] 0.742106
## $d
## [1] 0.6425789
# preserve names
z \leftarrow list(x = 1:3, y = 4:5)
map(z, length)
## $x
## [1] 3
##
## $y
## [1] 2
```

```
# equivalently
z %>% map(length)
## $x
## [1] 3
##
## $y
## [1] 2
The function group_split could be used to split a data frame into a list
(models <- mtcars %>%
  group_split(cyl) %>%
  map(function(df) lm(mpg ~ wt, data = df)))
## [[1]]
##
## Call:
## lm(formula = mpg ~ wt, data = df)
## Coefficients:
## (Intercept)
               -5.647
        39.571
##
##
##
## [[2]]
##
## lm(formula = mpg ~ wt, data = df)
##
## Coefficients:
## (Intercept)
                        wt
         28.41 -2.78
##
##
##
## [[3]]
##
## Call:
## lm(formula = mpg ~ wt, data = df)
##
## Coefficients:
## (Intercept)
                         wt
        23.868
                     -2.192
# or simply
(models <- mtcars %>%
  group_split(cyl) %>%
map(\sim lm(mpg \sim wt, data = .)))
## [[1]]
##
## Call:
```

```
## lm(formula = mpg ~ wt, data = .)
##
## Coefficients:
## (Intercept)
##
        39.571
                     -5.647
##
## [[2]]
##
## Call:
## lm(formula = mpg ~ wt, data = .)
##
## Coefficients:
## (Intercept)
                          wt
##
         28.41
                       -2.78
##
##
## [[3]]
##
## Call:
## lm(formula = mpg ~ wt, data = .)
## Coefficients:
## (Intercept)
                          wt
        23.868
##
                     -2.192
```

# Type-specific map

map() always returns a list, even if all the elements have the same flavor and are of length one. But in that case, you might prefer a simpler object.

- map\_lgl() makes a logical vector.
- map\_int() makes an integer vector.
- map\_dbl() makes a double vector.
- map\_chr() makes a character vector.

```
map_dbl(c(9, 16, 25), sqrt)
```

```
## [1] 3 4 5
```

#### Name and position shortcuts

The map\_ functions provide some shortcuts to extract elements from lists or vectors.

```
list(LETTERS, letters) %>% map_chr(2)
## [1] "B" "b"
```

```
mtcars %>%
  group_split(cyl) %>%
  map(~lm(mpg ~ wt, data = .)) %>%
  map(summary) %>%
  map_dbl("r.squared")
```

## [1] 0.5086326 0.4645102 0.4229655

# Map more than one argument

```
Two arguments
a \leftarrow c(7, 8, 9)
b \leftarrow c(2, 3, 4)
map2(a, b, choose)
## [[1]]
## [1] 21
##
## [[2]]
## [1] 56
##
## [[3]]
## [1] 126
a %>% map2(b, choose)
## [[1]]
## [1] 21
##
## [[2]]
## [1] 56
## [[3]]
## [1] 126
b %>% map2(a, . , choose)
## [[1]]
## [1] 21
## [[2]]
## [1] 56
##
## [[3]]
## [1] 126
```

```
# or equivalently
map2(a, b, ~ choose(.x, .y))
## [[1]]
## [1] 21
##
## [[2]]
## [1] 56
##
## [[3]]
## [1] 126
map2(a, b, function(x, y) choose(x, y))
## [[1]]
## [1] 21
##
## [[2]]
## [1] 56
##
## [[3]]
## [1] 126
More than two arguments
1 <- list(
     a = c(1, 2, 3),
     b = c(5, 4, 8),
     c = c(4, 7, 9))
1 %>% pmap(function(a, b, c) a / (b + c))
## [[1]]
## [1] 0.1111111
##
## [[2]]
## [1] 0.1818182
##
## [[3]]
## [1] 0.1764706
# or equivalently
1 %>% pmap(~ ..1 / (..2 + ..3))
## [[1]]
## [1] 0.1111111
##
## [[2]]
## [1] 0.1818182
##
## [[3]]
## [1] 0.1764706
```

# Enumerating over the input

```
a <- c(11, 12, 13)
b \leftarrow c(4, 5, 6)
a %>% imap_dbl(~ .x + b[.y])
## [1] 15 17 19
# equivalently
a %>% imap_dbl(function(v, i) v + b[i])
## [1] 15 17 19
Filtering
a <- 11:20
m <- list(lowers = letters, uppers = LETTERS)</pre>
a %>% pluck(1) # equivalent to a[1]
## [1] 11
m %>% pluck("lowers") # equivalent to m["lowers"]
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k" "l" "m" "n" "o" "p" "q" "r" "s"
## [20] "t" "u" "v" "w" "x" "y" "z"
m %>% pluck("lowers", 2) # equivalent to m["lowers"][2]
## [1] "b"
Keep and drop elements
is_even <- function(x) x \frac{2}{2} == 0
a %>% keep(is_even)
## [1] 12 14 16 18 20
# or equivalently
a \%% keep(~ . \%% 2 == 0)
## [1] 12 14 16 18 20
```

```
m %>% keep(~ "A" %in% .)
## $uppers
## [1] "A" "B" "C" "D" "E" "F" "G" "H" "I" "J" "K" "L" "M" "N" "O" "P" "Q" "R" "S"
## [20] "T" "U" "V" "W" "X" "Y" "Z"
a %>% keep(~!is_even(.))
## [1] 11 13 15 17 19
a %>% keep(negate(is_even))
## [1] 11 13 15 17 19
a %>% discard(is_even)
## [1] 11 13 15 17 19
Summerizing
# is every element in `a` even?
a %>% every(is_even) # maybe equivalent to all(is_even(a))
## [1] FALSE
# are some elements in `a` even?
a %>% some(is_even) # maybe equivalent to any(is_even(a))
## [1] TRUE
a %>% has_element(12) # 12 is a double
## [1] FALSE
a %>% has_element(12L)
## [1] TRUE
# the first even number is
a %>% detect(is_even)
## [1] 12
```

```
# the first even number divisible by 7 is
a %>% detect(~ . %% 7 == 0)

## [1] 14

# get the index of the first element divisible by 7, then extract it
a %>% detect_index(~ . %% 7 == 0) %>% pluck(a, .)

## [1] 14
```

## Modify elements selectively

```
# a %>% modify(~ . + 1) results in an error because a is an integer vector
a %>% modify(~ . + 1L)

## [1] 12 13 14 15 16 17 18 19 20 21

a %>% modify_if(is_even, ~ -.)

## [1] 11 -12 13 -14 15 -16 17 -18 19 -20

a %>% modify_if(~ !is_even(.), ~ -.)

## [1] -11 12 -13 14 -15 16 -17 18 -19 20

a %>% modify_at(c(1, 3, 5), ~ . + 2L)
## [1] 13 12 15 14 17 16 17 18 19 20
```

#### Reshape lists

```
m %>% flatten_chr()

## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j" "k" "l" "m" "n" "o" "p" "q" "r" "s"

## [20] "t" "u" "v" "w" "x" "y" "z" "A" "B" "C" "D" "E" "F" "G" "H" "I" "J" "K" "L"

## [39] "M" "N" "0" "P" "Q" "R" "S" "T" "U" "V" "W" "X" "Y" "Z"

(m1 <- m %>% map(~ .[1:5]))

## $lowers

## [1] "a" "b" "c" "d" "e"

##

## $uppers

## [1] "A" "B" "C" "D" "E"
```

```
m1 %>% pluck("lowers", 2)
## [1] "b"
m1 %>% transpose() %>% map(~ stringr::str_c(.$lowers, .$uppers))
## [[1]]
## [1] "aA"
##
## [[2]]
## [1] "bB"
##
## [[3]]
## [1] "cC"
##
## [[4]]
## [1] "dD"
##
## [[5]]
## [1] "eE"
```

### Add elements to lists

```
li <- list(a = 1, b = 2)
li %>%
  append(list(c = 3, d = 3)) %>%
  prepend(list(x = 0, y = 0)) %>%
  splice(u = 3, v = 4)
```

```
## $x
## [1] 0
##
## $y
## [1] 0
##
## $a
## [1] 1
##
## $b
## [1] 2
##
## $c
## [1] 3
##
## $d
## [1] 3
##
## $u
## [1] 3
##
## $v
## [1] 4
```

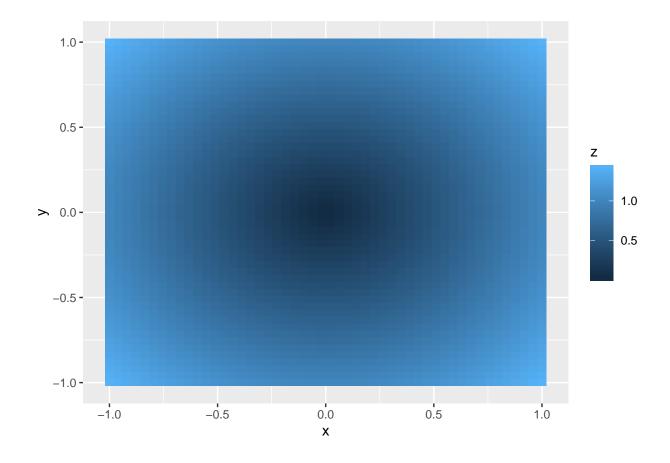
### Cross

The map function applies to the inputs elementwisely. Sometimes, we wish to consider all combinations of the inputs. The cross\* functions are designed to handle those situations.

cross2 usually is used together with lift

```
a \leftarrow c(7, 8, 9)
b < -c(2, 3, 4)
choose(7, 3)
## [1] 35
lift(choose)(list(7, 3))
## [1] 35
cross2(a, b) %>% map_dbl(lift(choose))
## [1] 21 28 36 35 56 84 35 70 126
li \leftarrow list(a = c(7, 8, 9),
           b = c(2, 3, 4))
li %>%
  set_names(c("n", "k")) %>%
  cross() %>%
  map(lift(choose))
## [[1]]
## [1] 21
##
## [[2]]
## [1] 28
##
## [[3]]
## [1] 36
##
## [[4]]
## [1] 35
##
## [[5]]
## [1] 56
##
## [[6]]
## [1] 84
##
## [[7]]
## [1] 35
##
## [[8]]
## [1] 70
##
## [[9]]
## [1] 126
```

```
dat <- tibble(</pre>
 x = seq(-1, 1, length = 50),
 y = seq(-1, 1, length = 50))
# we wish to plot a heat map of sqrt(x^2 + y^2)
dat \%% mutate(z = sqrt(x^2 + y^2)) # wrong
## # A tibble: 50 x 3
          X
              У
      <dbl> <dbl> <dbl>
##
## 1 -1 -1
                1.41
## 2 -0.959 -0.959 1.36
## 3 -0.918 -0.918 1.30
## 4 -0.878 -0.878 1.24
## 5 -0.837 -0.837 1.18
## 6 -0.796 -0.796 1.13
## 7 -0.755 -0.755 1.07
## 8 -0.714 -0.714 1.01
## 9 -0.673 -0.673 0.952
## 10 -0.633 -0.633 0.895
## # ... with 40 more rows
dat %>%
 cross_df() %>%
mutate(z = sqrt(x^2 + y^2)) \%
ggplot(aes(x, y, fill = z)) + geom_tile()
```



# Reduce iteratable

The reduce function combines the values of a vector by iteratively applying a binary function.

```
reduce(1:4, `+`)

## [1] 10

# it means ((1 + 2) + 3) + 4

reduce(1:4, `+`, .dir = "backward")

## [1] 10

# it means 1 + (2 + (3 + 4))

reduce(4:1, choose)

## [1] 6

# it means choose(choose(choose(4, 3), 2), 1)
```

```
# shortcircuit a reduction
sum_until <- function(x, y) {
   out <- x + y
   if (out > 30) {
      done(out)
   } else {
      out
   }
}
reduce(1:10, sum_until)

## [1] 36
accumulate works similarly but keep all the imediate calculations.
accumulate(1:4, `+`)

## [1] 1 3 6 10
accumulate(1:10, sum_until)
```

# Nesting

**##** [1] 1 3 6 10 15 21 28 36

```
mtcars %>%
nest(data = -cyl)
## # A tibble: 3 x 2
##
                     data
      cyl
     <dbl> <list<df[,10]>>
##
## 1
      6
                [7 x 10]
## 2
        4
                [11 x 10]
## 3
        8
                [14 x 10]
# or more readible
mtcars %>%
 group_by(cyl) %>%
 nest() %>%
 mutate(model = data %>% map(~ lm(mpg ~ wt, data = .))) %>%
 transmute(r_sq = model %>% map(summary) %>% map_dbl("r.squared"))
## # A tibble: 3 x 2
## # Groups: cyl [3]
##
      cyl r_sq
##
    <dbl> <dbl>
## 1 6 0.465
## 2
       4 0.509
## 3
       8 0.423
```

# References

- R for Data Science https://r4ds.had.co.nz/iteration.html
  purrr tutorial https://jennybc.github.io/purrr-tutorial/
  purrr cheatsheet https://github.com/rstudio/cheatsheets/blob/master/purrr.pdf