

# Advanced R - Hadley Wickham Ch 9 Functionals

R Ladies Netherlands Bookclub

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# Welcome!

- This is joint effort between RLadies Nijmegen, Rotterdam, 's-Hertogenbosch (Den Bosch), Amsterdam and Utrecht
- We meet every 2 weeks to go through a chapter
- Use the HackMD to present yourself, ask questions and see your breakout room
- We split in breakout rooms after the presentation, and we return to the main jitsi link after xx min
- There are still possibilities to present a chapter :) Sign up at  
[https://rladiesnl.github.io/book\\_club/](https://rladiesnl.github.io/book_club/)
- <https://advanced-r-solutions.rbind.io/> has some answers and we could PR the ones missing
- The R4DS book club repo has a Q&A section.[https://github.com/r4ds/bookclub-Advanced\\_R](https://github.com/r4ds/bookclub-Advanced_R)



# Thanks to:

- Hadley Wickham for writing AdvancedR
- Allison Horst for the palmerpenguins package
- R Core Team (2020). R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. URL <https://www.R-project.org/>.
- Lionel Henry and Hadley Wickham (2020). purrr: Functional Programming Tools. R package version 0.3.4. <https://CRAN.R-project.org/package=purrr>
- Gorman KB, Williams TD, Fraser WR (2014) Ecological Sexual Dimorphism and Environmental Variability within a Community of Antarctic Penguins (Genus Pygoscelis). PLoS ONE 9(3): e90081.  
<https://doi.org/10.1371/journal.pone.0090081>
- Yihui Xie (2020). xaringan: Presentation Ninja. R package version 0.16. <https://CRAN.R-project.org/package=xaringan>
- Garrick Aden-Buie (2020). xaringanExtra: Extras And Extensions for Xaringan Slides. R package version 0.0.17. <https://github.com/gadenbuie/xaringanExtra>

and all the authors of R packages used in this presentation

# Functional programming - Chp 9, 10 and 11

R has **first-class** functions. This means the language supports:

- assigning functions to variables
- store funtions in a list
- pass functions as arguments to other functions
- create functions inside functions
- return functions as the result of a function

A functions is **pure**, if:

- the output only depends on the input (same input, same output)
- the function has no side effects (so no actions beside the output value)

R is not strictly a functional programming language,  
but it is very well possible to program in a **functional style**:

- decompose a problem in subproblems
- solve each subproblem with a set of functions.

# Functional programming - Chp 9, 10 and 11

In	Out	
	Vector	Function
Vector	Regular functions Chp 6	Function factories Chp 10
Function	<b>Functionals</b> <b>Chp 9</b>	Function operators Chp 11

# Functionals, `function(function) = vector`

```
let_us <- function(f, inputs) f(inputs)
```

```
let_us(sum, 1:4)
```

```
[1] 10
```

```
let_us(median, 1:4)
```

```
[1] 2.5
```

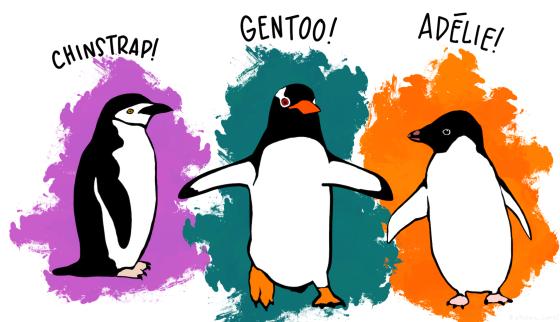
```
let_us(range, 1:4)
```

```
[1] 1 4
```

```
let_us(sqrt, 1:4)
```

```
[1] 1.000000 1.414214 1.732051 2.000000
```

# About the demo data: palmerpenguins::penguins

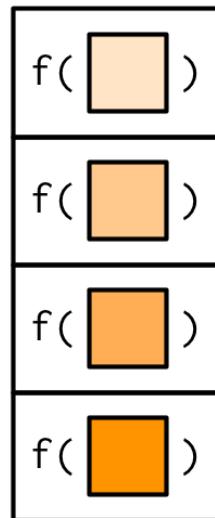
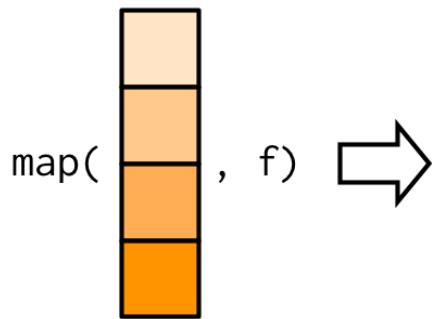


Artwork by @allison\_horst

```
str(penguins)
```

```
tibble [344 x 7] (S3: tbl_df/tbl/data.frame)
$ species      : Factor w/ 3 levels "Adelie", "Chinstrap", ...: 1 1 1 1 1 1 1 1 1 1 ...
$ island       : Factor w/ 3 levels "Biscoe", "Dream", ...: 3 3 3 3 3 3 3 3 3 3 ...
$ bill_length_mm: num [1:344] 39.1 39.5 40.3 NA 36.7 39.3 38.9 39.2 34.1 42 ...
$ bill_depth_mm: num [1:344] 18.7 17.4 18 NA 19.3 20.6 17.8 19.6 18.1 20.2 ...
$ flipper_length_mm: int [1:344] 181 186 195 NA 193 190 181 195 193 190 ...
$ body_mass_g   : int [1:344] 3750 3800 3250 NA 3450 3650 3625 4675 3475 4250 ...
$ sex          : Factor w/ 2 levels "female", "male": 2 1 1 NA 1 2 1 2 NA NA ...
```

# purrr::map()



input slot 1: vector  
input slot 2: function  
output: list

```
str(purrr::map)
```

```
function (.x, .f, ...)
```

```
map(penguins, typeof)
```

```
$species  
[1] "integer"  
  
$island  
[1] "integer"  
  
$bill_length_mm  
[1] "double"  
  
$bill_depth_mm  
[1] "double"
```

```
$flipper_length_mm  
[1] "integer"  
  
$body_mass_g  
[1] "integer"  
  
$sex  
[1] "integer"
```

# `purrr::map()` - atomic functions 1/3

## list as output

```
map(penguins, typeof)
```

```
$species  
[1] "integer"  
  
$island  
[1] "integer"  
  
$bill_length_mm  
[1] "double"  
  
$bill_depth_mm  
[1] "double"  
  
$flipper_length  
[1] "integer"  
  
$body_mass_g  
[1] "integer"  
  
$sex  
[1] "integer"
```

## chr vector as output

```
map_chr(penguins, typeof)
```

```
species           island      bill_length
"integer"        "integer"    "double"
flipper_length_mm body_mass_g "integer"
"integer"         "integer"    "integer"
```

# purrr::map() - atomic functions 2/3

logical vector as output

```
map_lgl(penguins, is.double)
```

species	island	bill_length_mm	bill_depth_mm
FALSE	FALSE	TRUE	TRUE
flipper_length_mm	body_mass_g	sex	
FALSE	FALSE	FALSE	

integer vector as output

```
n_unique <- function(x) length(unique(x))  
map_int(penguins, n_unique)
```

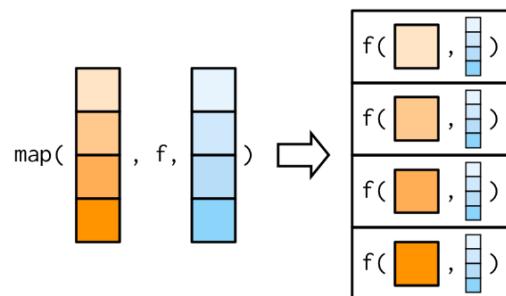
species	island	bill_length_mm	bill_depth_mm
3	3	165	81
flipper_length_mm	body_mass_g	sex	
56	95	3	

# purrr::map() - atomic functions 3/3

```
map_db1(penguins, mean, na.rm = TRUE)
```

	species	island	bill_length_mm	bill_depth_mm
	NA	NA	43.92193	17.15117
flipper_length_mm		body_mass_g		sex
200.91520		4201.75439		NA

This is an example of passing arguments:



Extra arguments (after `f`) will be passed along as is (no decomposition).

# Anonymous functions and shortcuts 1/

```
# instead of:  
n_unique <- function(x) length(unique(x))  
map_int(penguins, n_unique)
```

	species	island	bill_length_mm	bill_depth_mm
3	3	165	81	
56	body_mass_g	sex		
	95	3		

```
# you can create an inline anonymous function:  
map_int(penguins, function(x) length(unique(x)))
```

	species	island	bill_length_mm	bill_depth_mm
3	3	165	81	
56	body_mass_g	sex		
	95	3		

```
# or even with less characters, and with a tilde/twiddle:  
map_int(penguins, ~length(unique(.x)))
```

	species	island	bill_length_mm	bill_depth_mm
3	3	165	81	
56	body_mass_g	sex		
	95	3		

# Anonymous functions and shortcuts 2/

Very useful for making sets of random numbers

```
x <- map(1:3, ~ runif(.x))
str(x)
```

```
List of 3
$ : num 0.416
$ : num [1:2] 0.905 0.47
$ : num [1:3] 0.3311 0.0935 0.4597
```

If at the second spot there is a vector/list instead of function,  
result is extraction of elements from the input vector:

```
map_dbl(x, 1)
```

```
[1] 0.4164139 0.9046881 0.3310901
```

```
map_dbl(x, 2)
```

```
Error: Result 1 must be a single double, not NULL of length 0
```

# Anonymous functions and shortcuts 3/

```
map_dbl(x, 2)
```

Error: Result 1 must be a single double, not NULL of length 0

Prevent this error by supplying a default value:

```
map_dbl(x, 2, .default = NA)
```

```
[1] NA 0.47013588 0.09346625
```

Extract elements by:

- location, with an integer vector
- by name, with a character vector
- combi name/location, with a list



# Intermezzo on *trim*

Plain mean:

```
mean(c(1,1,2,2,2,2,2,2,2,1000,1000))
```

```
[1] 168.1667
```

Trim 20% of the observations at both sides:

```
mean(c(1,1,2,2,2,2,2,2,2,1000,1000), trim = 0.2)
```

```
[1] 2
```

It cuts off from the ordered vector:

```
mean(c(1000,1,2,2,2,2,2,2,2,1000,1), trim = 0.2)
```

```
[1] 2
```

So it trims the smallest and the largest values.

# Use argument names

Calculate the means for the numeric columns in `penguins`:

```
map_dbl(penguins[, 3:6], mean, TRUE)
```

Error in mean.default(.x[[i]], ...): 'trim' must be numeric of length one

What we wanted was to have the `na.rm = TRUE`:

```
map_dbl(penguins[, 3:6], mean, na.rm = TRUE)
```

bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
43.92193	17.15117	200.91520	4201.75439

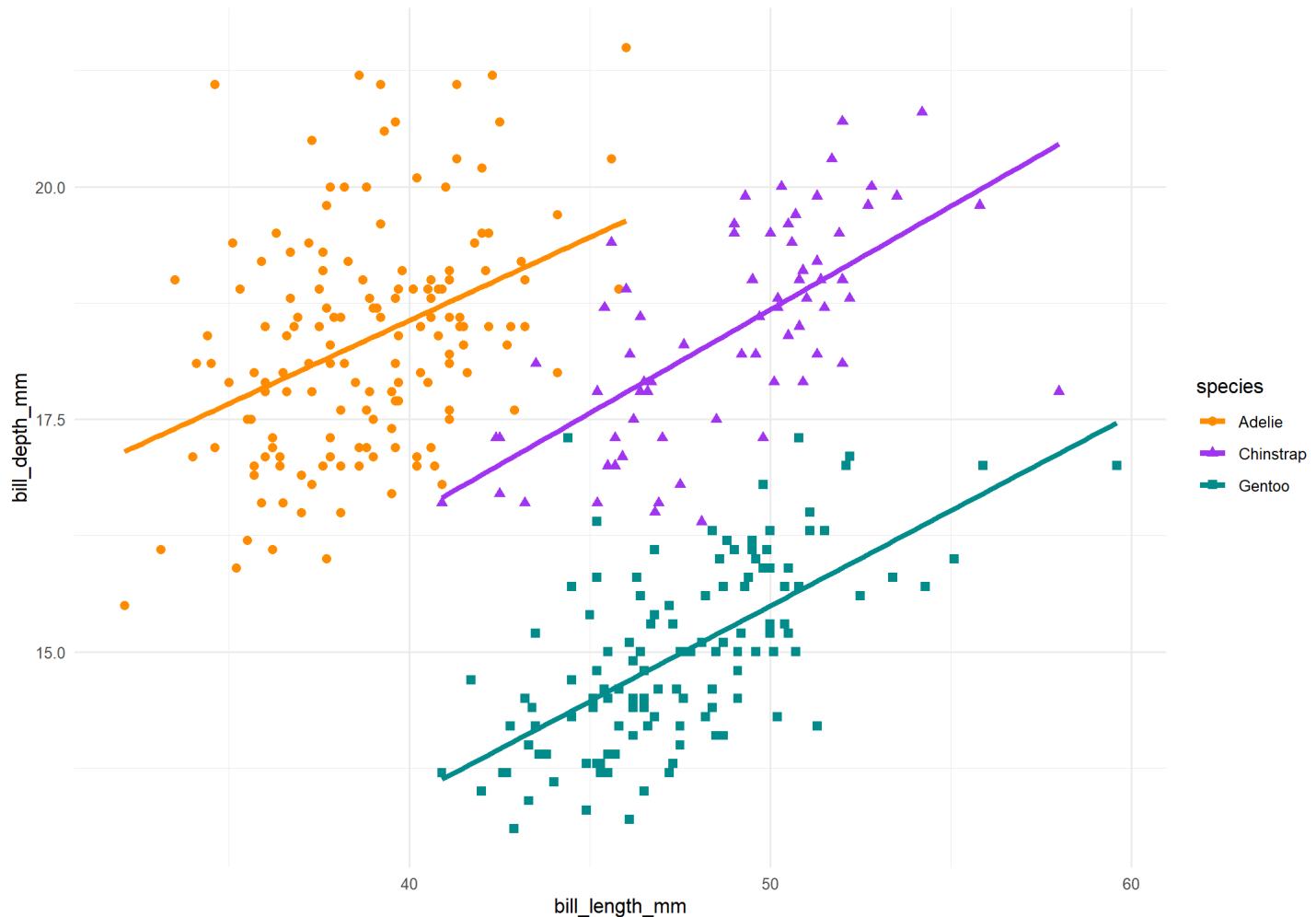
Another example:

```
map_dbl(penguins[, 3:6], mean, na.rm = TRUE, 0.1)
```

bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
43.90693	17.17263	200.33577	4154.01460

The 0.1 appears to be the *trim*. But would you know that by looking at the script?

# Regression example



# Regression example

```
penguins_species <- split(penguins, penguins$species)
```

```
penguins_species %>%  
  # inputs are the 3 splits  
  # of penguins  
  # function is the lm:  
  map(~ lm(bill_depth_mm ~  
           bill_length_mm,  
           data = .x))
```

\$Adelie

Call:

```
lm(formula = bill_depth_mm ~ bill_length_mm, dat
```

Coefficients:

(Intercept)	bill_length_mm
11.4091	0.1788

\$Chinstrap

Call:

```
lm(formula = bill_depth_mm ~ bill_length_mm, dat
```

Coefficients:

(Intercept)	bill_length_mm
7.5691	0.2222

\$Gentoo

Call:

```
lm(formula = bill_depth_mm ~ bill_length_mm, dat
```

# Regression example

```
penguins_species <- split(penguins, penguins$species)
```

```
penguins_species %>%
  # inputs are the 3 splits
  # of penguins
  # function is the lm:
  map(~ lm(bill_depth_mm ~
            bill_length_mm,
            data = .x)) %>%
  # inputs are the models
  # function is coef()
  map(coef)
```

```
$Adelie
(Intercept) bill_length_mm
11.4091245   0.1788343

$Chinstrap
(Intercept) bill_length_mm
7.5691401    0.2222117

$Gentoo
(Intercept) bill_length_mm
5.2510084    0.2048443
```

# Regression example

```
penguins_species <- split(penguins, penguins$species)
```

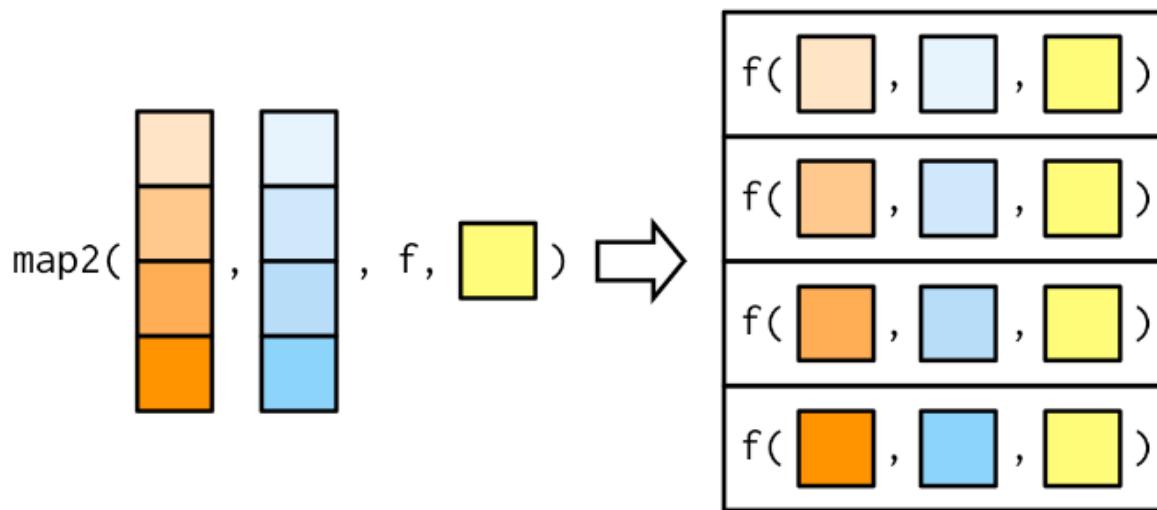
```
penguins_species %>%  
  # inputs are the 3 splits  
  # of penguins  
  # function is the lm:  
  map(~ lm(bill_depth_mm ~  
           bill_length_mm,  
           data = .x)) %>%  
  # inputs are the models  
  # function is coef()  
  map(coef) %>%  
  # inputs are sets of coefficients  
  # subset to get the slopes  
  map(2)
```

\$Adelie  
[1] 0.1788343

\$Chinstrap  
[1] 0.2222117

\$Gentoo  
[1] 0.2048443

# purrr::map2()



```
str(purrr::map2)
```

```
function (.x, .y, .f, ...)
```

input slot 1: vector

input slot 2: another vector

input slot 3: function

input slot 4: additional arguments to function

output: list

# purrr::map2() - examples

```
map2(1:3, 4:6, ~.x ^ .y)
```

```
[[1]]
```

```
[1] 1
```

```
[[2]]
```

```
[1] 32
```

```
[[3]]
```

```
[1] 729
```

```
map2_chr(penguins$species, penguins$island, str_c, sep = " - ") %>% head(30)
```

```
[1] "Adelie - Torgersen" "Adelie - Torgersen" "Adelie - Torgersen"  
[4] "Adelie - Torgersen" "Adelie - Torgersen" "Adelie - Torgersen"  
[7] "Adelie - Torgersen" "Adelie - Torgersen" "Adelie - Torgersen"  
[10] "Adelie - Torgersen" "Adelie - Torgersen" "Adelie - Torgersen"  
[13] "Adelie - Torgersen" "Adelie - Torgersen" "Adelie - Torgersen"  
[16] "Adelie - Torgersen" "Adelie - Torgersen" "Adelie - Torgersen"  
[19] "Adelie - Torgersen" "Adelie - Torgersen" "Adelie - Biscoe"  
[22] "Adelie - Biscoe" "Adelie - Biscoe" "Adelie - Biscoe"  
[25] "Adelie - Biscoe" "Adelie - Biscoe" "Adelie - Biscoe"  
[28] "Adelie - Biscoe" "Adelie - Biscoe" "Adelie - Biscoe"
```

# purrr::map2() recycles

```
map2(1:3, 4, ~.x ^ .y)
```

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

```
[[2]]  
[1] 16
```

```
[[3]]  
[1] 81
```

This is equivalent with  
treating the second argument as an additional argument to the function in a map():

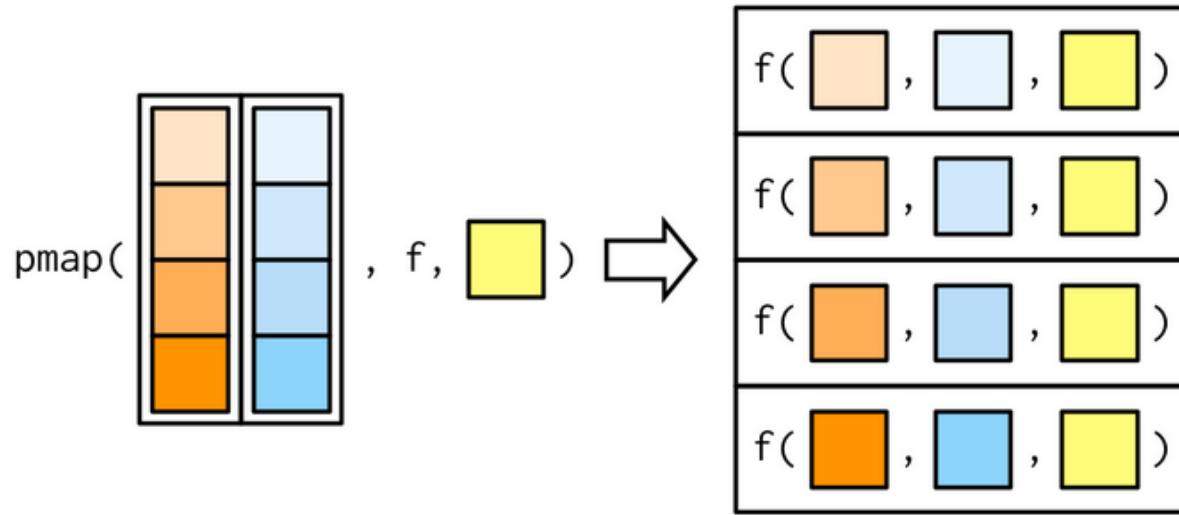
```
map(1:3, ~.x ^ .y, 4)
```

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

```
[[2]]  
[1] 16
```

```
[[3]]  
[1] 81
```

# purrr::pmap()



```
str(purrr::pmap)
```

```
function (.l, .f, ...)
```

input slot 1: list

input slot 2: function

input slot 3: additional arguments to function

output: list

# purrr::pmap() - example

Use pmap to draw sets of random samples from a number of normal distributions:

```
str(rnorm)
```

```
function (n, mean = 0, sd = 1)
```

So we need a tibble with the desired inputs:

```
params <- tibble(n = 3:5, mean = c(30, 20, 10), sd = 1:3)
```

Because the variable names in `params` match the argument names in `rnorm`, this works:

```
pmap(params, rnorm)
```

```
[[1]]  
[1] 29.13720 29.99630 29.05474
```

```
[[2]]  
[1] 21.70650 19.42706 19.65767 19.48892
```

```
[[3]]  
[1] 5.703459 14.434996 8.865970 8.252433 7.747931
```

# purrr::imap() - example names

```
str(imap)
```

```
function (.x, .f, ...)
```

```
imap(x, f):
```

- x has names → `map2(x, names(x), f)`
- x has NO names → `map2(x, seq_along(x), f)`

## Example

```
imap_chr(penguins, ~ str_c("First value of ", .y, " is ", .x[[1]]))
```

```
species
"First value of species is Adelie"
island
"First value of island is Torgersen"
bill_length_mm
"First value of bill_length_mm is 39.1"
bill_depth_mm
"First value of bill_depth_mm is 18.7"
flipper_length_mm
"First value of flipper_length_mm is 181"
body_mass_g
"First value of body_mass_g is 3750"
```

# purrr::imap() - example no names

```
x <- map(1:3, ~ sample(1000, 10))
```

```
x
```

```
[[1]]
[1] 391 564 895 218 150 244 372 190 227 640
```

```
[[2]]
[1] 956 363 183 676 424 785 859 490 602 400
```

```
[[3]]
[1] 520 197 360 957 817 83 162 760 753 351
```

```
imap_chr(x, ~ paste0("The highest value of ", .y, " is ", max(.x)))
```

```
[1] "The highest value of 1 is 895" "The highest value of 2 is 956"
[3] "The highest value of 3 is 957"
```



# modify()

Returns a modified copy, same type of output as input.

```
str(modify)
```

```
function (.x, .f, ...)
```

```
map(penguins[3:6], ~.x*2) %>% str()
```

List of 4

```
$ bill_length_mm    : num [1:344] 78.2 79 80.6 NA 73.4 78.6 77.8 78.4 68.2 84 ...
$ bill_depth_mm    : num [1:344] 37.4 34.8 36 NA 38.6 41.2 35.6 39.2 36.2 40.4 ...
$ flipper_length_mm: num [1:344] 362 372 390 NA 386 380 362 390 386 380 ...
$ body_mass_g       : num [1:344] 7500 7600 6500 NA 6900 7300 7250 9350 6950 8500 ...
```

```
modify(penguins[3:6], ~.x*2) %>% str()
```

tibble [344 x 4] (S3:tbl\_df/tbl/data.frame)

```
$ bill_length_mm    : num [1:344] 78.2 79 80.6 NA 73.4 78.6 77.8 78.4 68.2 84 ...
$ bill_depth_mm    : num [1:344] 37.4 34.8 36 NA 38.6 41.2 35.6 39.2 36.2 40.4 ...
$ flipper_length_mm: num [1:344] 362 372 390 NA 386 380 362 390 386 380 ...
$ body_mass_g       : num [1:344] 7500 7600 6500 NA 6900 7300 7250 9350 6950 8500 ...
```

# walk()

Most functions are used for their return value.

Some mainly for their side-effects, like `plot()`, `write.csv()`, ...

These latter functions do have an invisible return value (see 6.7.2),  
that become visible with `map( )`.

```
print(3:4)
```

```
[1] 3 4
```

```
map_chr(3:4, print)
```

```
[1] 3  
[1] 4
```

```
[1] "3" "4"
```

Use `walk` when you only want the side-effect:

```
walk(3:4, print)
```

```
[1] 3  
[1] 4
```

# walk() - example saving multiple files

```
temp <- tempfile()  
# returns a vector of character strings  
# which can be used as names for temporary files  
temp
```

```
[1] "C:\\\\Users\\\\871112\\\\AppData\\\\Local\\\\Temp\\\\RtmpoXF7rC\\\\file4aa064ba477f"
```

```
# make the temporary directory  
dir.create(temp)  
# and see that it is empty  
dir(temp)
```

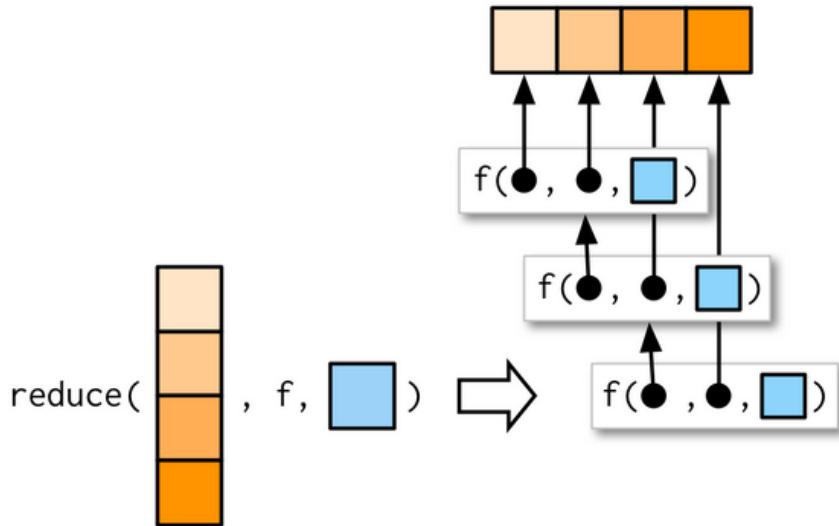
```
character(0)
```

```
peng <- split(penguins, penguins$species)  
paths <- file.path(temp, paste0("species-", names(peng), ".csv"))  
walk2(peng, paths, write.csv)  
dir(temp)
```

```
[1] "species-Adelie.csv"    "species-Chinstrap.csv" "species-Gentoo.csv"
```

```
# to clean up all the temp directories and content made  
unlink(temp, recursive = TRUE)
```

# reduce()



```
str(reduce)
```

```
function (.x, .f, ..., .init, .dir = c("forward", "backward"))
```

# reduce( ) - example intersect

```
mylist <- map(1:3, ~ sample(1:10, 15, replace = T))
str(mylist)
```

```
List of 3
$ : int [1:15] 9 7 8 10 2 6 4 3 8 4 ...
$ : int [1:15] 2 10 2 7 1 9 6 4 1 8 ...
$ : int [1:15] 3 3 1 6 8 10 4 8 5 4 ...
```

How to find the common numbers?

```
out <- mylist[[1]]
(out <- intersect(out, mylist[[2]]))
```

```
[1] 9 7 8 10 2 6 4 3
```

```
(out <- intersect(out, mylist[[3]]))
```

```
[1] 8 10 2 6 4 3
```

```
reduce(mylist, intersect)
```

```
[1] 8 10 2 6 4 3
```

```
accumulate(mylist, intersect)
```

```
[[1]]
[1] 9 7 8 10 2 6 4 3 8 4 7 7 10 7
```

```
[[2]]
[1] 9 7 8 10 2 6 4 3
```

```
[[3]]
[1] 8 10 2 6 4 3
```

# reduce() - use of .init 1/2

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

```
[1] 24
```

```
reduce(3, `*`)
```

```
[1] 3
```

```
reduce("purrrrrr", `*`)
```

```
[1] "purrrrrr"
```

So, obviously no check on input type ...

# reduce( ) - use of .init 2/2

```
reduce(integer(), `*`)
```

Error: `.x` is empty, and no `.`.init` supplied

From the help on `init`:

If supplied, will be used as the first value to start the accumulation, rather than using [1]. This is useful if you want to ensure that `reduce` returns a correct value when `.x` is empty. If missing, and `.x` is empty, will throw an error.

```
# prod returns the product of all the values present in its arguments.  
prod(integer())
```

```
[1] 1
```

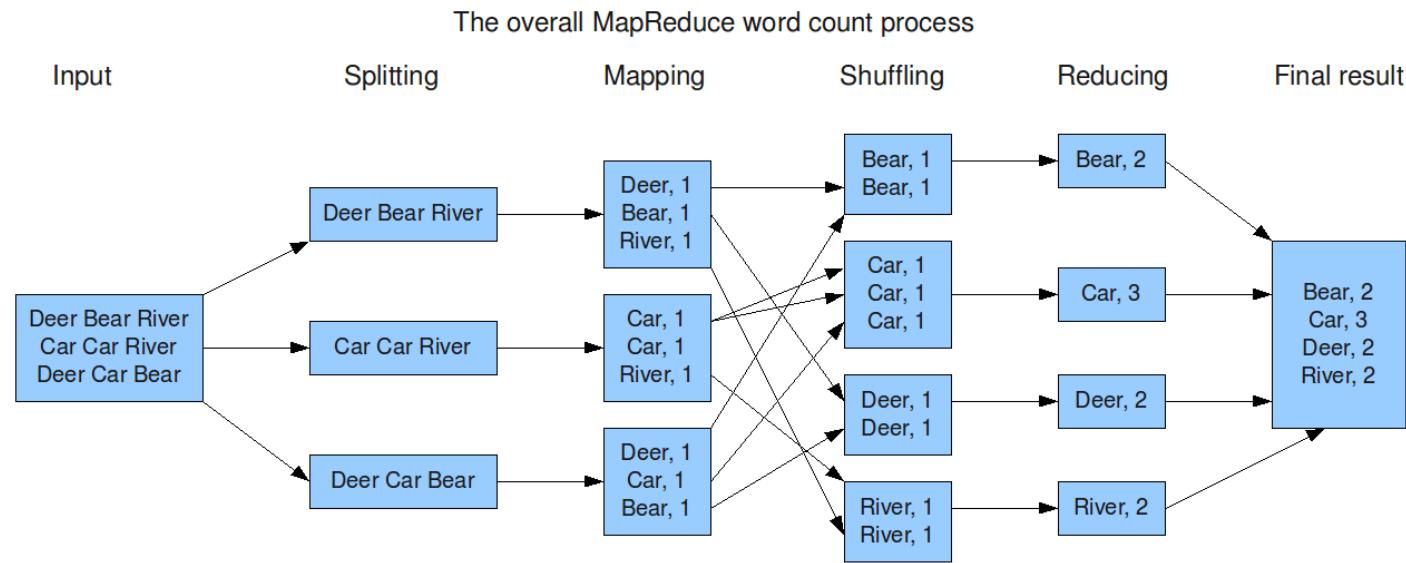
```
reduce(integer(), `*`, .init=1)
```

```
[1] 1
```

```
reduce("purrrrrr", `*`, .init = 1) # and now this DOES throw an error
```

Error in `.x * .y`: non-numeric argument to binary operator

# Map-reduce



Source: [https://whatsbigdata.be/wp-content/uploads/2014/06/MapReduce\\_Work\\_Structure.png](https://whatsbigdata.be/wp-content/uploads/2014/06/MapReduce_Work_Structure.png)

# Predicate functionals 1/4

A **predicate function** is a function that returns TRUE or FALSE.

```
is.numeric(penguins$species)
```

```
[1] FALSE
```

A **predicate functional** applies a predicate function to each element of a vector:

```
# some returns TRUE when first TRUE element is seen  
some(penguins, is.numeric)
```

```
[1] TRUE
```

```
# every returns FALSE when first FALSE element is seen  
every(penguins, is.numeric)
```

```
[1] FALSE
```

```
# none has to check entire vector  
none(mylist, is.numeric)
```

```
[1] FALSE
```

# Predicate functionals 2/4

```
detect(penguins %>% head(), is.factor)      # value of the first match
```

```
[1] Adelie Adelie Adelie Adelie Adelie Adelie  
Levels: Adelie Chinstrap Gentoo
```

```
detect_index(penguins %>% head(), is.factor) # location of the first match
```

```
[1] 1
```

```
keep(penguins %>% head(n = 1), is.numeric)
```

```
# A tibble: 1 x 4  
  bill_length_mm bill_depth_mm flipper_length_mm body_mass_g  
        <dbl>          <dbl>            <int>        <int>  
1         39.1           18.7            181        3750
```

```
discard(penguins %>% head(n = 1), is.numeric)
```

```
# A tibble: 1 x 3  
  species island   sex  
  <fct>    <fct>   <fct>  
1 Adelie   Torgersen male
```

# Predicate functionals 3/4

```
map_dbl(penguins, mean, na.rm = TRUE)
```

species	island	bill_length_mm	bill_depth_mm
NA	NA	43.92193	17.15117
flipper_length_mm	body_mass_g	sex	
200.91520	4201.75439	NA	

```
map_dbl(keep(penguins, is.numeric), mean, na.rm = TRUE)
```

bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
43.92193	17.15117	200.91520	4201.75439

# Predicate functionals 4/4

```
map_if(penguins, is.numeric, mean, na.rm = TRUE) %>% str()
```

```
List of 7
$ species      : Factor w/ 3 levels "Adelie", "Chinstrap", ...: 1 1 1 1 1 1 1 1 ...
$ island       : Factor w/ 3 levels "Biscoe", "Dream", ...: 3 3 3 3 3 3 3 3 3 ...
$ bill_length_mm: num 43.9
$ bill_depth_mm: num 17.2
$ flipper_length_mm: num 201
$ body_mass_g   : num 4202
$ sex          : Factor w/ 2 levels "female", "male": 2 1 1 NA 1 2 1 2 NA NA ...
```

```
modify_if(penguins, is.numeric, mean, na.rm = TRUE) %>% str()
```

```
tibble [344 x 7] (S3:tbl_df/tbl/data.frame)
$ species      : Factor w/ 3 levels "Adelie", "Chinstrap", ...: 1 1 1 1 1 1 1 1 ...
$ island       : Factor w/ 3 levels "Biscoe", "Dream", ...: 3 3 3 3 3 3 3 3 3 ...
$ bill_length_mm: num [1:344] 43.9 43.9 43.9 43.9 43.9 ...
$ bill_depth_mm: num [1:344] 17.2 17.2 17.2 17.2 17.2 ...
$ flipper_length_mm: num [1:344] 201 201 201 201 201 ...
$ body_mass_g   : num [1:344] 4202 4202 4202 4202 4202 ...
$ sex          : Factor w/ 2 levels "female", "male": 2 1 1 NA 1 2 1 2 NA NA ...
```

bye



Let us take a break!

I feel amazing! It's all done.

# Breakout questions

- Did you use these functions before? Please share some details/tips/stories.
- What are the 23 primary variants of `map( )` Hadley Wickham mentiones in 9.4?
- In case you want to try some exercises, try these:
  - 9.2.6.3
  - 9.2.6.6
  - 9.6.3.5
  - or any other exercise :)