

A workshop with

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## Outline

What is functional programming?

- Why is it useful?
- apply family of functions
  - description
  - code demo
- {purrr} package
  - description
  - code demo

## what functional programming?

Programming is an abstraction of reality.

### Different programming paradigms:

### **Functional programming**

- Focus on the evaluation of functions
- Variables and functions
- Declarative programming model

### **Object oriented programming**

- Based on the concept of objects
- Objects and methods
- Imperative programming model

## functional programming in R

R incorporates aspects of both object oriented and functional programming.

### First class functions

• You can do anything with functions that you can do with vectors: you can assign them to variables, store them in lists, pass them as arguments to other functions, create them inside functions, and even return them as the result of a function.

## Functional methods in lieu of loops

- Apply functions
- {purr} package map functions

## why is functional programming useful in R?

Functional programming in R can be:

- more concise
- easier to read and debug
- faster (in certain situations)
- more elegant

# apply functions

- A family of functions that allow the user to *apply* a function to elements of a vector, list, or dataframes.
- Avoids the explicit use of loops.
- Included in base R.
- Now considered **legacy** functionality and should **not** be used for **new** code still useful for maintaining code bases.

# apply functions

function	usage
apply()	apply a given function across a dimension of an array; typically columns or rows of a matrix
lapply()	<i>list apply -</i> apply a given function to every element of a list and obtain a list as a result
sapply()	<b>simplified lapply</b> – apply a given function to every element of a list and returns the simplest possible form of the result

## apply()

```
# general format of apply function
apply(X = dataStructure, MARGIN = margin, FUN = function)
# example usage of apply function
exampleMatrix \leftarrow matrix(1:50, nrow = 10, ncol =5)
# calculate across rows
apply(X = exampleMatrix, MARGIN = 1, FUN = mean)
# returns a vector of length 10 with means of each row
# calculate across columns
apply(X = exampleMatrix, MARGIN = 2, FUN = mean)
# returns a vector of length 5 with means of each column
```

## lapply()

```
# general format of lapply function
lapply (dataList, function)
dataList %>% lapply(function)
# example usage of apply function
exampleList <- list(height = sample(58:72, 10),
                    weight = sample(110:300, 10))
lapply(exampleList, median)
exampleList %>% lapply(median)
# returns a list of length 2 containing median of height in
first element and median of weight in the second element
```

## sapply()

```
# general format of lapply function
sapply(dataList, function)
dataList %>% sapply(function)
# example usage of apply function
exampleList <- list(height = sample(58:72, 10),
                    weight = sample(110:300, 10))
sapply(exampleList, median)
exampleList %>% sapply(median)
# returns a vector of length 2 containing median of height in
first element and median of weight in the second element
```

## Additional apply() functions

function	usage
tapply()	apply a given function to subsets of a data structure
mapply()	<i>multi-variate sapply</i> - passes multiple arguments to a given function
rapply()	recursive lapply - recursively apply a given function to a list
vapply()	apply a given function and set the expected output type

These functions provide an answer to a more niche environments and attempt to solve some shortcomings of the original apply functions.

# live demonstration – apply functions

## {purrr} package

purrr enhances R's functional programming (FP) toolkit by providing a complete and consistent set of tools for working with functions and vectors/lists.

core of the functional programming in {purrr} package is the family of map() functions which replace many instances of for loops.



# improvements in purr::map() vs. apply()

- Always returns a list or lists.
- © Consistent input/output across different functions.
- Easier to write and read code.



## from R for Data Science:

- The goal of using purrr functions instead of for loops is to allow you to break common list manipulation challenges into independent pieces:
  - 1. How can you solve the problem for a single element of the list? Once you've solved that problem, purrr takes care of generalizing your solution to every element in the list.
  - 2. If you're solving a complex problem, how can you break it down into bite-sized pieces that allow you to advance one small step towards a solution? With purrr, you get lots of small pieces that you can compose together with the pipe.

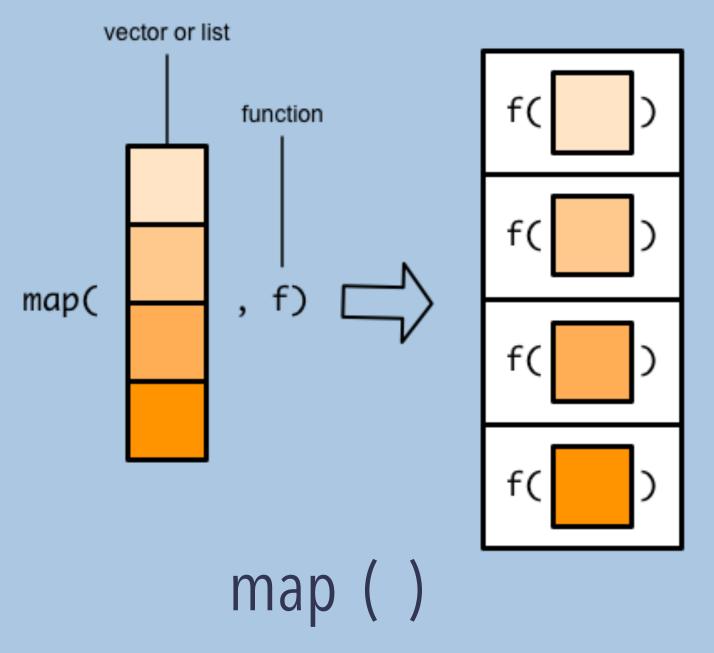


## {purr} map family of functions

function	usage
map()	applies a function to <b>every</b> element of a list/vector
map2()	applies a function to <b>every pair</b> of elements from two lists/vectors
pmap()	parallel map - applies a function to a group of elements in a list/vectors

There are additional, more advanced map functions that allow you to apply a list of functions or apply functions to certain elements of the data structure.







## map()

```
# general format of map function
map(data structure, function)
data structure %>% map(function) # with pipes
# example usage of map function
map(exampleList, mean)
# example usage of map function with pipes
exampleList %>% map(mean)
# returns a list of length 2 mean of each element
 of the input list (each sample).
```

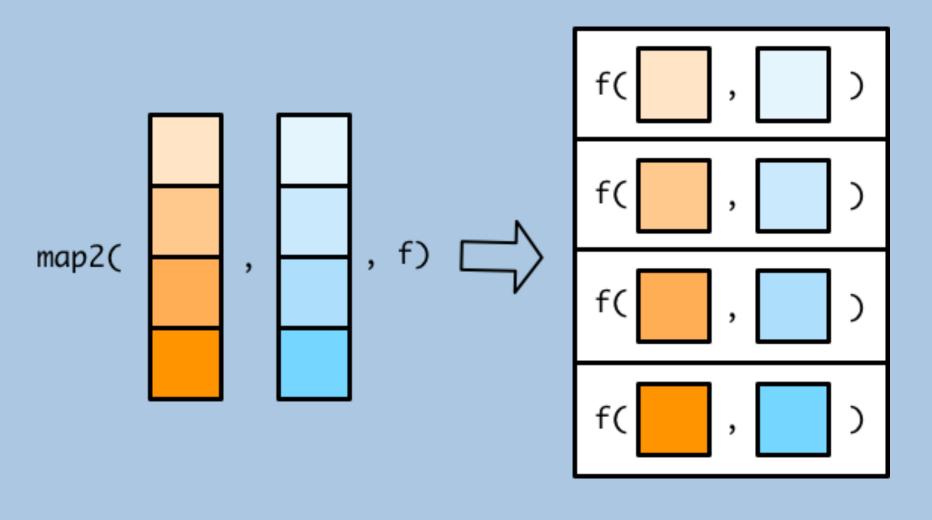


## map()

	input	output
map()	vector/list	vector/list
map_lgl()	vector/list	logical vector/list
map_int()	vector/list	integer vector/list
map_dbl()	vector/list	double vector/list
map_chr()	vector/list	character vector/list

These functions take a vector as input, applies a user-passed function to each element, then returns a new vector of the same length and with the same names as the input vector.

purrr



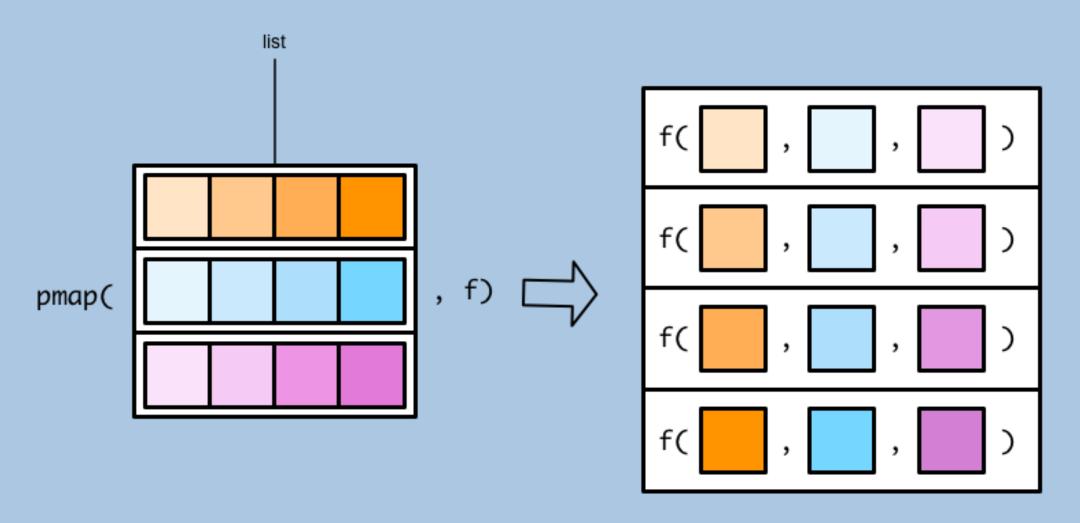




## map2()

```
# general format of map2 function
map2(data_structure1, data_structure2, function)
# example usage of map2 function
exampleVector1 <- sample(1:100, 12)
exampleVector2 <- sample(1:100, 12)
map2_dbl(exampleVector1, exampleVector2, min)</pre>
```

# returns a vector of length 12 with the min of each element pair, i.e. the first element of the vector purrish min (exampleVector1[1], exampleVector2[1])



# pmap()



## pmap ()

```
# general format of pmap function
pmap(data structure, function)
# example usage of pmap function
exampleDF <-data.frame( sample1 = sample(1:100, 10), sample2 = sample(1:100, 10)), sample3 = sample(1:100, 10))
pmap(exampleDF, sum)
# example usage of pmap function with pipes
exampleDF %>% pmap(sum)
# returns a list of length 10 with the sum of each first element,
  each second element, etc.
```

# live demonstration – {purrr} package



# additional resources

## Apply functions with purrr:: CHEAT SHEET

### **Map Functions**

#### **ONE LIST**

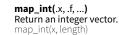
**map(**.x, .f, ...**)** Apply a function to each element of a list or vector, and return a list.

x <- list(a = 1:10, b = 11:20, c = 21:30) l1 <- list(x = c("a", "b"), y = c("c", "d")) map(l1, sort, decreasing = TRUE)





map\_dbl(.x, .f, ...)
Return a double vector.
map\_dbl(x, mean)









map\_dfc(.x, .f, ...)
Return a data frame created by column-binding.
map\_dfc(|1, rep, 3)



map\_dfr(.x, .f, ..., .id = NULL) Return a data frame created by row-binding. map\_dfr(x, summary)



walk(.x, .f, ...) Trigger side effects, return invisibly. walk(x, print)

### TWO LISTS

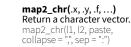
**map2(.**x, .y, .f, ...) Apply a function to pairs of elements from two lists or vectors, return a list. y <- list(1, 2, 3); z <- list(4, 5, 6); |2 <- list(x = "a", y = "z") map2(x, y, ~ x \* .y)





map2\_dbl(.x, .y, .f, ...) Return a double vector. map2\_dbl(y, z,  $\sim$  .x / .y)







map2\_lgl(.x, .y, .f, ...) Return a logical vector. map2\_lgl(l2, l1, `%in%`)



map2\_dfc(.x, .y, .f, ...)
Return a data frame created by column-binding.
map2\_dfc([1, [2, ~ as.data.frame(c(x, .y)))



map2\_dfr(.x, .y, .f, ..., .id = NULL) Return a data frame created by row-binding. map2\_dfr(l1, l2, ~ as.data.frame(c(.x, .y)))



walk2(.x, .y, .f, ...) Trigger side effects, return invisibly. walk2(objs, paths, save)

#### MANY LISTS

**pmap**(.l, .f, ...) Apply a function to groups of elements from a list of lists or vectors, return a list. pmap(list(x, y, z),  $\sim ...1 * (...2 + ...3)$ )





Return a double vector. pmap\_dbl(list(y, z), ~ .x / .y) pmap\_int(.l, .f, ...)

Return an integer vector.

pmap int(list(v, z), `+`)

pmap\_dbl(.l, .f, ...)



pmap\_chr(.l, .f, ...) Return a character vector. pmap\_chr(list(l1, l2), paste, collapse = ",", sep = ":")



pmap\_lgl(.l, .f, ...)
Return a logical vector.
pmap\_lgl(list(l2, l1), `%in%`)



pmap\_dfc(.l, .f, ...) Return a data frame created by column-binding. pmap\_dfc(list(l1, |2), ~ as.data.frame(c(.x, .y)))



pmap\_dfr(.l, .f, ..., .id =
NULL) Return a data frame
created by row-binding.
pmap\_dfr(list(l1, l2),
 as.data.frame(c(x, y)))



**pwalk**(.l, .f, ...) Trigger side effects, return invisibly. pwalk(list(objs, paths), save)

### LISTS AND INDEXES

**imap(**.x, .f, ...**)** Apply .f to each element and its index, return a list.

imap(y, ~ paste0(.y, ": ", .x))





imap\_dbl(.x, .f, ...)
Return a double vector.
imap\_dbl(y, ~ .y)



imap\_int(.x, .f, ...)
Return an integer vector.
imap\_int(y, ~ .y)



imap\_chr(.x, .f, ...)
Return a character vector.
imap\_chr(y, ~ paste0(.y, ": ", .x))



imap\_lgl(.x, .f, ...)
Return a logical vector.
imap\_lgl(l1, ~ is.character(.y))



imap\_dfc(.x, .f, ...)
Return a data frame created by column-binding.
imap\_dfc(l2,
~ as.data.frame(c(.x, .y)))



imap\_dfr(.x, .f, ..., .id = NULL) Return a data frame created by row-binding. imap\_dfr(l2, ~ as.data.frame(c(x, .v)))



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iwalk(.x, .f, ...) Trigger side
effects, return invisibly.
iwalk(z, ~ print(paste0(.y, ": ", .x)))



**posit** 

Use ~ . with functions like **map()** that have single arguments.

map(l, ~ . + 2)
becomes
map(l, function(x) x + 2))

Use ~ .x .y with functions like map2() that have two arguments.

map2(l, p, ~ .x +.y)
becomes
map2(l, p, function(l, p) l + p)

Use ~ ..1 ..2 ..3 etc with functions like **pmap()** that have many arguments.

pmap(list(a, b, c), ~ ..3 + ..1 - ..2)
becomes
pmap(list(a, b, c), function(a, b, c) c + a -b)

Use ~ .x .y with functions like **imap()**. .x will get the list value and .y will get the index, or name if available.

imap(list(a, b, c), ~ paste0(.y, ": ", .x)
outputs "index: value" for each item

Use a string or an integer with any map function to index list elements by name or position. map(l, "name") becomes map(l, function(x) x[["name"]])



### Work with Lists

### Filter



**keep(**.x, .p, ...**)** Select elements that pass a logical test. Conversely, discard(). keep(x, is.numeric)



compact(.x, .p = identity) Drop empty elements. compact(x)



head\_while(.x, .p, ...) Return head elements until one does not pass. Also tail\_while(). head\_while(x, is.character)



**detect(**.x, .f, ..., dir = c("forward", "backward"), .right = NULL, .default = NULL) Find first element to pass. detect(x, is.character)



detect\_index(.x, .f, ..., dir = c("forward", "backward"), .right = NULL) Find index of first element to pass. detect index(x, is.character)



**every(**.x, .p, ...**)** Do all elements pass a test? every(x. is.character)



**some(**.x, .p, ...**)** Do some elements pass a test? some(x, is.character)



**none(**.x, .p, ...**)** Do no elements pass a test? none(x, is, character)



has\_element(.x, .y) Does a list contain an element? has\_element(x, "foo")



pluck\_depth(x) Return depth (number of levels of indexes). pluck\_depth(x)

### Index



pluck(.x, ..., .default=NULL) Select an element by name or index. Also attr\_getter() and chuck(). pluck(x, "b")



assign\_in(x, where, value) Assign a value to a location using pluck selection. assign\_in(x, "b", 5) x |> assign in("b", 5)

x |> pluck("b")



modify\_in(.x, .where, .f) Apply a function to a value at a selected location. modify in(x, "b", abs) x |> modify\_in("b", abs)

### Reshape



flatten(.x) Remove a level of indexes from a list. Also flatten\_chr() etc. flatten(x)



array\_tree(array, margin = NULL) Turn array into list. Also array\_branch(). z <- array(1:12, c(2,2,2))array\_tree(x, margin = 3)



transpose(.l, .names = NULL) Transposes the index order in a multi-level list. transpose(x)



 $set_names(x, nm = x)$ Set the names of a vector/list directly or with a function. set\_names(x, c("p", "q", "r")) set\_names(x, tolower)

### Modify



modify(.x, .f, ...) Apply a function to each element. Also modify2(), and imodify().  $modify(x, \sim .+2)$ 



modify\_at(.x, .at, .f, ...) Apply a function to selected elements. Also map\_at(). modify\_at(x, "b", ~.+ 2)



modify\_if(.x, .p, .f, ...) Apply a function to elements that pass a test. Also map\_if(). modify\_if(x, is.numeric,~.+2)



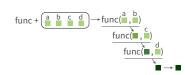
modify\_depth(.x, .depui, .i, ...)

Apply function to each element at a given level of a list. Also map\_depth()

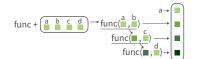
 $modify_depth(x, 1, \sim + 2)$ 

### Reduce

reduce(.x, .f, ..., .init, .dir = c("forward", "backward")) Apply function recursively to each element of a list or vector. Also reduce2(). reduce(x, sum)



accumulate(.x, .f, ..., .init) Reduce a list, but also return intermediate results. Also accumulate2(). accumulate(x, sum)



### List-Columns



max	seq	(
3	<int [3]=""></int>	ä
4	<int [4]=""></int>	١
5	<int [5]=""></int>	

List-columns are columns of a data frame where each element is a list or vector instead of an atomic value. Columns can also be lists of data frames. See tidyr for more about nested data and list columns.

#### **WORK WITH LIST-COLUMNS**

Manipulate list-columns like any other kind of column, using dplyr functions like mutate() and transmute(). Because each element is a list, use map functions within a column function to manipulate each element.

map(), map2(), or pmap() return lists and will create new list-columns.



Suffixed map functions like map int() return an atomic data type and will simplify list-columns into regular columns.





#### Welcome

An evolving book

I Data Structures

1 Introduction

2 Vectors, lists, and tibbles

3 List columns

**II Functions** 

4 Introduction

**5** Vector functions

6 Anonymous functions

**7** Predicate functions

**III Iteration** 

8 Introduction

9 Basic map functions

10 Map with multiple inputs

11 Other purrr functions

**IV Tidy evaluation** 

12 Introduction

13 Tidy evaluation basics

### **Functional Programming**

Sara Altman, Bill Behrman, Hadley Wickham

2021-09-09

### Welcome

This book is a practical introduction to functional programming using the tidyverse.

### An evolving book

This book is not intended to be static. Starting in April 2019, we use this book to teach functional programming in the Stanford Data Challenge Lab (DCL) course. The DCL functions as a testing ground for educational materials, as our students give us routine feedback on what they read and do in the course. We use this feedback to constantly improve our materials, including this book. The source for the book is also available on GitHub where we welcome suggestions for improvements.

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Functional programming text available at https://dcl-prog.stanford.edu

# Stat 8054 Lecture Notes: R as a Functional Programming Language

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### 2 R

- The version of R used to make this document is 4.2.1.
- The version of the rmarkdown package used to make this document is 2.19.
- The version of the magrittr package used to make this document is 2.0.3.
- The version of the CatDataAnalysis package used to make this document is 0.1.5.
- The version of the glmbb package used to make this document is 0.5.1.

### 3 Reading

- A blog post at R Bloggers: Functional programming in R.
- Three chapters in Advanced R:
  - Functional programming
  - Functionals
  - Function Operators
- A question in the R FAQ: What are the differences between R and S?.
- Some sections of my 3701 handout on the basics of R:
  - 4 Functions
  - 6 More on Functions
  - 7 Still More on Functions
  - 7.5 A Long Example (Maximum Likelihood Estimation), especially subsections 7.5.4 and 7.5.5 and 7.5.6
- Some sections of the book *The R Language Definition*, which is one of the R manuals that can be found at CRAN and also in your own R

https://www.stat.umn.edu/geyer/8054/notes/functional.html