## Functional programing

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### Motivation

#### DRY principle: Don't Repeat Yourself

Every piece of knowledge must have a single, unambiguous, authoritative representation within a system

Popularised by the "Pragmatic Programmers"

```
# Fix missing values
dfa\Gamma dfa == -99\Gamma <- NA
df$b\Gamma df$b == -997 <- NA
df$c\[df$c == -99\] <- NA
df$d\Gamma df$d == -997 <- NA
df$e\( \text{d} f$e == -99\( \text{ - NA}\)
df$f[df$f == -99] <- NA
df g[df = -98] < - NA
df h[df = -99] < - NA
df$i\Gammadf$i == -99\Gamma <- NA
df i [df j == -99] <- NA
df$k\Gamma df$k == -99\Gamma <- NA
```

```
fix_missing <- function(x) {</pre>
  x[x == -99] \leftarrow NA
  X
df$a <- fix_missing(df$a)</pre>
df$b <- fix_missing(df$b)</pre>
df$c <- fix_missing(df$c)</pre>
df$d <- fix_missing(df$d)</pre>
df$e <- fix_missing(df$e)</pre>
df$f <- fix_missing(df$f)</pre>
df$g <- fix_missing(df$g)</pre>
df$h <- fix_missing(df$h)
df$h <- fix_missing(df$i)</pre>
df$j <- fix_missing(df$j)</pre>
df$k <- fix_missing(df$k)</pre>
```

## DRY principle prevents inconsistency

## More powerful abstractions lead to less repetition

```
fix_missing <- function(x) {
   x[x == -99] <- NA
   x
}

df[] <- lapply(df, fix_missing)</pre>
```

#### And easier generalisation

```
fix_missing <- function(x) {
    x[x == -99] <- NA
    x
}
numeric <- vapply(df, is.numeric, logical(1))
df[numeric] <- lapply(df[numeric], fix_missing)</pre>
```

#### And easier generalisation

```
missing_fixer <- function(missing) {
   function(x) {
     x[x == missing] <- NA
     x
   }
}
numeric <- vapply(df, is.numeric, logical(1))
df[numeric] <- lapply(df[numeric], missing_fixer(-99))</pre>
```

## Mainus

```
x < -5
f <- function() {
  x < -20
  y <- 10
  c(x = x, y = y)
```

What does this function return?

```
x < -5
f <- function() {
  x < -20
  y <- 10
  c(x = x, y = y)
```

```
What does this
x < -5
                    function return?
 <- function() {
  y <- 10
  c(x = x, y = y)
g()
```

```
x < -5
g <- function() {
  y <- 10
  c(x = x, y = y)
g()
```

```
What does this
x < -5
                           function return?
h <- function() {</pre>
  y <- 10
  i <- function() {</pre>
    z < -20
     c(x = x, y = y, z = z)
  i()
```

```
x < -5
h <- function() {</pre>
  y <- 10
  i <- function() {</pre>
    z < -20
    c(x = x, y = y, z = z)
  i()
                        5 10 20
```

```
j <- function() {</pre>
  if (!exists("a")) {
    a <- 5
  } else {
    a < -a + 1
  print(a)
j()
```

What does this function return the first time you run it? The second time?

```
j <- function() {</pre>
  if (!exists("a")) {
    a <- 5
  } else {
    a < -a + 1
  print(a)
j()
```

What does this function return the first time you run it? The second time?

#### Functionals

#### Functionals

- A functional is a function that takes a function as input and returns a vector.
- Functionals are used to abstract over common patterns of looping.
- Common functions are lapply(), apply(), tapply(), ...
- Reduce bugs by better communicating intent.

```
set.seed(1014)
# Create some random output:
# 20 random vectors with random lengths
1 <- replicate(20, runif(sample(1:10, 1)),</pre>
  simplify = FALSE)
str(1)
```

```
# Extract length of each element
lengths <- vector("list", length(l))
for (i in seq_along(l)) {
  lengths[[i]] <- length(l[[i]])
}
lengths</pre>
```

```
Preallocating space for output saves a lot of time # Extract length or each element lengths <- vector("list", length(l)) for (i in seq_along(l)) { lengths[[i]] Safe shortcut for 1:length(l) } lengths
```

How would you change this to compute the mean of each element?

```
# Extract length of each element
lengths <- vector("list", length(l))
for (i in seq_along(l)) {
  lengths[[i]] <- length(l[[i]])
}
lengths</pre>
```

```
# Compute mean of each element
means <- vector("list", length(l))
for (i in seq_along(l)) {
  means[[i]] <- mean(l[[i]])
}
means</pre>
```

#### Your turn

What are the common parts of this pattern? How could you extract them into a function?

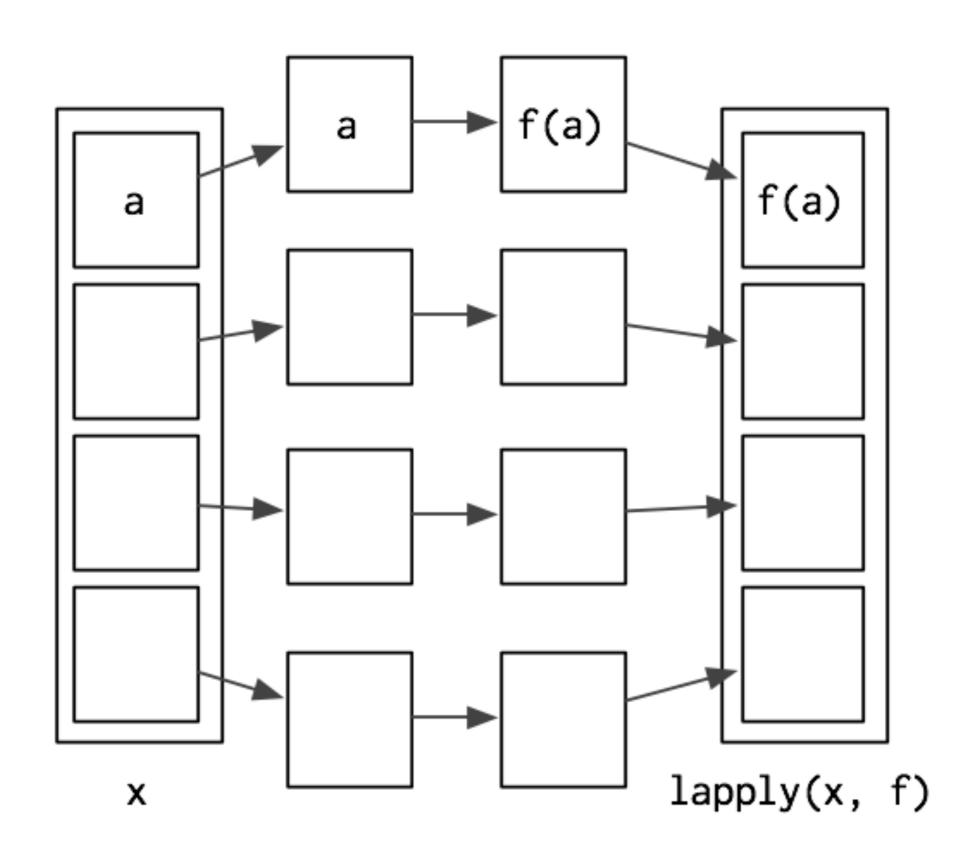
```
do_each <- function(?, ?) {</pre>
do_each(1, length)
do_each(1, mean)
```

#### No peeking until you've made an attempt!

#### Functions can be arguments!

```
do_each <- function(x, f) {</pre>
 out <- vector("list", length(x))</pre>
  for(i in seq_along(x)) {
    out[[i]] <- f(x[[i]])
  out
do_each(1, length)
do_each(1, mean)
do_each(1, min)
```

```
# BUT WAIT...
lapply(1, length)
lapply(1, mean)
lapply(1, min)
# We've just reinvented lapply :)
# Two differences:
# * lapply() uses some C tricks to be faster
# * lapply() passes ... on to f
```



Placeholder for "any other" arguments

```
do_each <- function(x, f, ...) {</pre>
  out <- vector("list", length(x))</pre>
  for(i in seq_along(x)) {
    out[[i]] <- f(x[[i]], ...)
  out
do_{each}(1, mean, trim = 0.5)
```

#### Your turn

The function below scales a vector so it falls in the range [0, 1]. How would you apply it to every column of a data frame?

```
scale01 <- function(x) {
    rng <- range(x, na.rm = TRUE)
    (x - rng[1]) / (rng[2] - rng[1])
}</pre>
```

```
mtcars <- lapply(mtcars, scale01)</pre>
mtcars # a list :(
rm(mtcars)
mtcars[] <- lapply(mtcars, scale01)</pre>
mtcars # a data frame :)
rm(mtcars)
for(i in seq_along(mtcars)) {
  mtcars[[i]] <- scale01(mtcars[[i]])</pre>
```

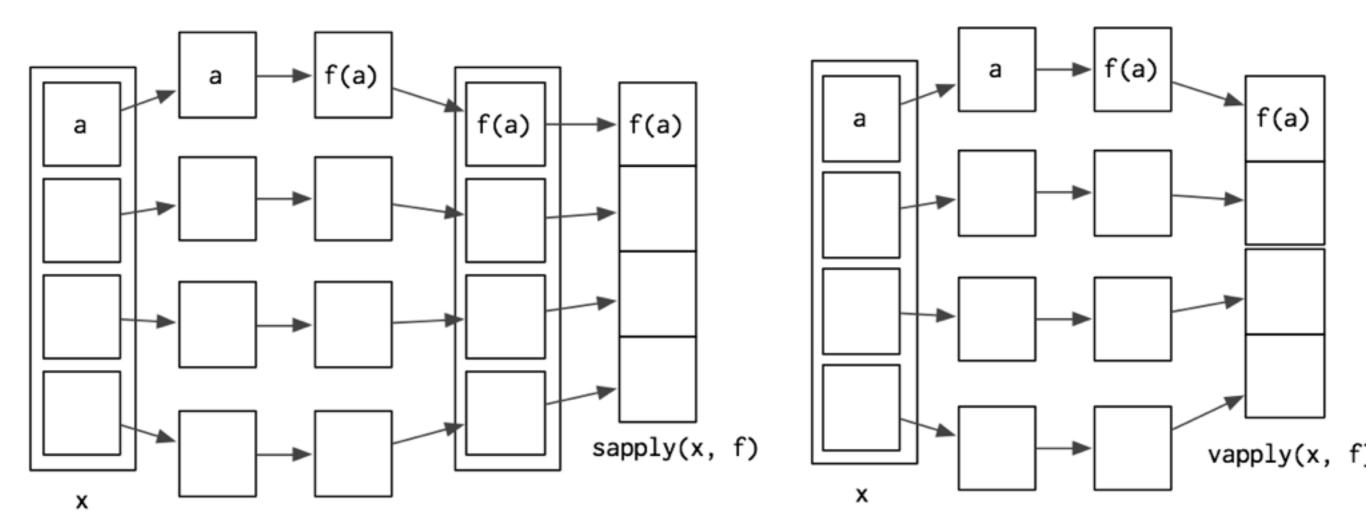
# Friends of lapply()

Variation	Function	
I want an atomic vector, not a list	<pre>sapply(), vapply()</pre>	
I have more than one input	Map(), mapply()	•
I have lots of computation to do	mclapply()	

```
# Output is annoyingly long!
lapply(mtcars, is.numeric)
# We'd prefer a vector instead of a list
sapply(mtcars, is.numeric)
vapply(mtcars, is.numeric, logical(1))
# sapply() is useful interactively, but
# dangerous in a function because it has
# to guess the output type.
```

sapply(mtcars, class)

mtcars\$x <- Sys.now()
sapply(mtcars, class)</pre>



```
sapply2 <- function(x, f, ...) {
  res \leftarrow lapply2(x, f, ...)
  simplify2array(res)
}
vapply2 <- function(x, f, f.value, ...) {</pre>
  out <- matrix(rep(f.value, length(x)), nrow = length(x))
  for (i in seq_along(x)) {
    res <- f(x[[i]], ...)
    stopifnot(
      length(res) == length(f.value),
      typeof(res) == typeof(f.value)
    out[i, ] <- res
  }
  out
```



Never use inside a function!

```
col_means <- function(df) {
  numeric <- vapply(df, is.numeric, logical(1))
  numeric_cols <- df[, numeric, drop = FALSE]

  data.frame(lapply(numeric_cols, mean))
}</pre>
```

## Anonymous functions

```
# Don't have to name functions to use with
# apply etc. Can create an inline anonymous
# function:
lapply(mtcars, function(x) length(unique(x)))
integrate(function(x) sin(x)^2, 0, pi)
```

```
# Creating an anonymous function
function(x) 3
# Calling an anonymous function
(function(x) 3)()
# Not:
function(x) 3 ()
# Anonymous functions work just like ordinary
# functions
formals(function(x = 4) g(x) + h(x))
body(function(x = 4) g(x) + h(x))
environment(function(x = 4) g(x) + h(x))
```

```
# Functions that input and output
# functions are called function operators.
# They abstract away common uses of anonymous
# functions
library(pryr)
sapply(mtcars, function(x) length(unique(x)))
sapply(mtcars, compose(length, unique))
sapply(mtcars, function(x) mean(x, trim = 0.2))
sapply(mtcars, partial(mean, trim = 0.2))
# Allows very expressive code, where the
# focus is on the operations, not the arguments
```

	Vector	Function	
Vector	Regular function	Function factory	
Function	Functional	Function operators	

# Function factories

```
x <- 0
y <- 10
f <- function() {</pre>
  x < -1
  function() {
    y <- 2
    x + y
# What does f() return?
# What does f()() mean? What does it do?
# How does it work?
```

```
f <- function() {</pre>
  x <- sample(1000, 1)
  function() {
    x + 2
f1 <- f()
f2 <- f()
f1()
f1()
f2()
```

### Scoping

R uses lexical scoping: variable lookup is based on where functions were created.

If a variable isn't found in the current environment, R looks in the parent: the environment where the function was created.

Anonymous functions remember their parent environment, even if it has since "disappeared".

```
# Closures are useful when you want a function
# that can create a whole class of functions:
# a function factory
power <- function(exponent) {</pre>
  function(x) {
    x ^ exponent
square <- power(2)</pre>
square(2)
square(4)
cube <- power(3)</pre>
cube(2)
cube(4)
```

#### square

```
# We can find the environment and its parent
environment(square)
as.list(environment(square))
as.list(environment(cube))
```

pryr::unenclose(square)

pryr::unenclose(cube)

```
missing_fixer <- function(missing) {</pre>
  force(missing)
  function(x) {
    x[x == missing] <- NA
    X
missing_vals <-c(-99, -99, -9000, -90)
fixers <- lapply(missing_vals, missing_fixer)</pre>
invoke <- function(f, ...) f(...)</pre>
Map(invoke, fixers, df)
->
invoke(fixers[[1]], df[[1]]) -> fixers[[1]](df[[1]])
invoke(fixers[[2]], df[[2]]) -> fixers[[2]](df[[2]])
```

# Learning more

### Advanced R

http://adv-r.had.co.nz/Functions.html

http://adv-r.had.co.nz/Functionalprogramming.html

http://adv-r.had.co.nz/Functionals.html

http://adv-r.had.co.nz/Functionoperators.html

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