

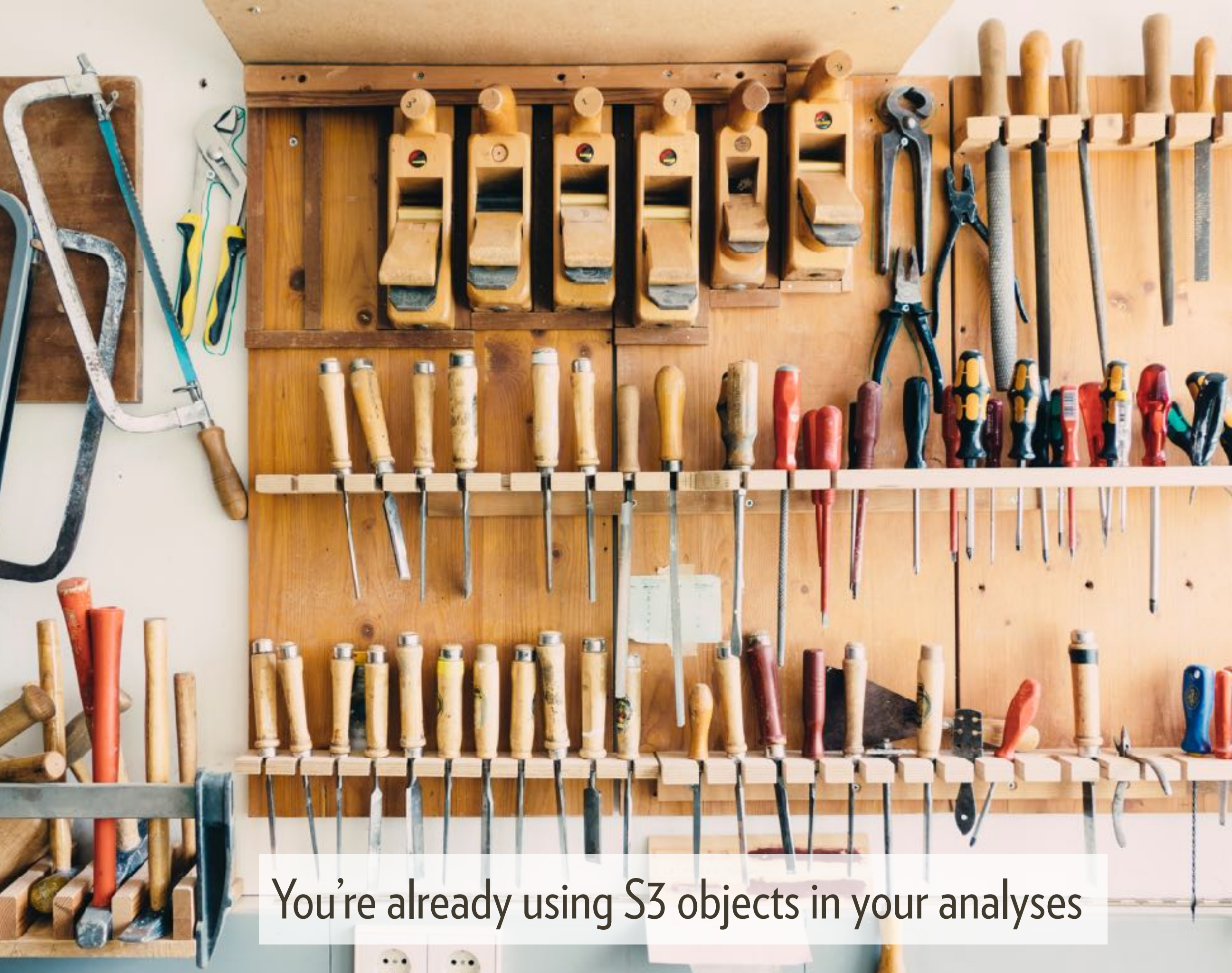
○○ programming

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Motivation

Why should you care about **S3**?



You're already using S3 objects in your analyses

Important S3 objects in base R

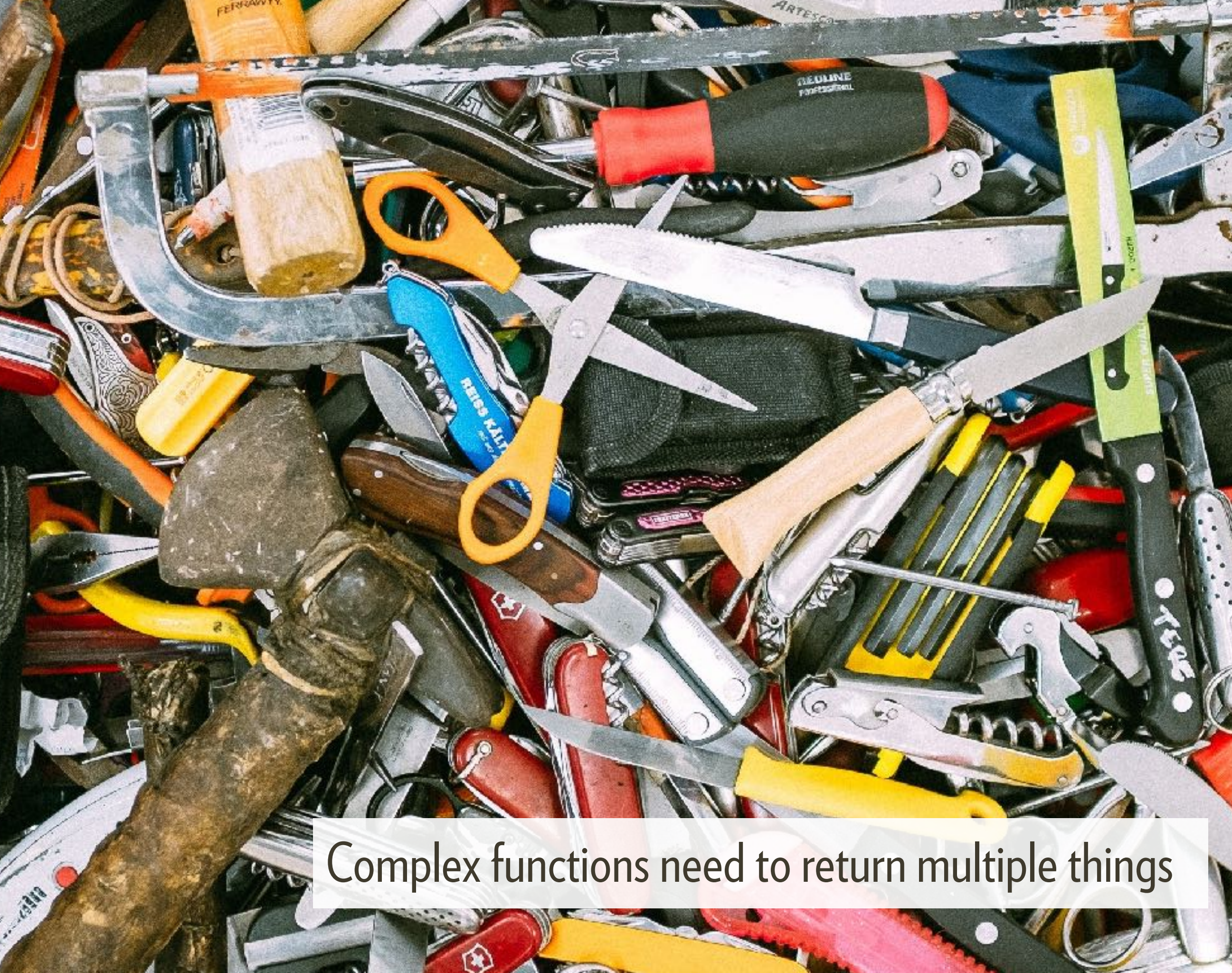
`data.frame()`

`factor()`

`Sys.Date()`

`Sys.time()`

`table()`



Complex functions need to return multiple things

This is obviously important for linear models

```
mod <- lm(mpg ~ wt, data = mtcars)
str(mod)
```

```
# But also their summaries
sum <- summary(mod)
str(sum)
```

Form follows function



One example is linear models

```
sum
#> Call:
#> lm(formula = mpg ~ wt, data = mtcars)
#>
#> Residuals:
#>      Min       1Q   Median       3Q      Max
#> -4.5432 -2.3647 -0.1252  1.4096  6.8727
#>
#> Coefficients:
#>              Estimate Std. Error t value Pr(>|t|)
#> (Intercept)  37.2851     1.8776   19.858  < 2e-16 ***
#> wt          -5.3445     0.5591   -9.559 1.29e-10 ***
#> ---
#> Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#>
#> Residual standard error: 3.046 on 30 degrees of freedom
#> Multiple R-squared:  0.7528, Adjusted R-squared:  0.7446
#> F-statistic: 91.38 on 1 and 30 DF, p-value: 1.294e-10
```


Another example is tibbles

Total size

A tibble: 53,940 x 10

	carat	cut	color	clarity	depth	table	price	x	y	z
	<dbl>	<ord>	<ord>	<ord>	<dbl>	<dbl>	<int>	<dbl>	<dbl>	<dbl>
1	0.230	Ideal	E	SI2	61.5	55.0	326	3.95	3.98	2.43
2	0.210	Premium	E	SI1	59.8	61.0	326	3.8		
3	0.230	Good	E	VS1	56.9	65.0	327	4.05	4.07	2.51
4	0.290	Premium	I	VS2	62.4	58.0	334	4.20	4.23	2.63
5	0.310	Good	J	SI2	63.3	58.0	335	4.34	4.35	2.75
6	0.240	"Very Good"	J	VVS2	62.8	57.0	336	3.94	3.96	2.48
7	0.240	"Very Good"	I	VVS1	62.3	57.0	336	3.95	3.98	2.47
8	0.260	"Very Good"	H	SI1	61.9	55.0	337	4.07	4.11	2.53
9	0.220	Fair	E	VS2	65.1	61.0	337	3.87	3.78	2.49
10	0.230	"Very Good"	H	VS1	59.4	61.0	338	4.00	4.05	2.39

Variable type

... with 53,930 more rows

Only shows first 10 rows



Intermediate objects can allow you to partition API complexity

e.g. string manipulation in base R vs stringr

```
str_replace(x, fixed("y"), "")  
str_replace(x, regex(".", ignore_case = TRUE), "")
```

vs.

```
gsub("y", "", x, fixed = TRUE)  
gsub("y", "", x, fixed = FALSE, ignore.case = TRUE)
```

S3 makes packages extensible

New methods

Lets you extend other packages

New generics

Write packages in way that others can easily extend.

You could use a nested if statement

```
mean <- function(x, ...) {  
  if (is.Date(x)) {  
    ...  
  } else if (is.difftime(x)) {  
    ...  
  } else if (is.POSIXct(x)) {  
  
  } else if (is.POSIXlt(x)) {  
    ...  
  } else {  
    ...  
  }  
}
```

But a generic function lets anyone extend

```
mean <- function(x, ...) {  
  UseMethod("mean")  
}
```

```
mean.Date <- function(x, ...) ...  
mean.diffftime <- function(x, ...) ...  
mean.POSIXct <- function(x, ...) ...  
mean.POSIXlt <- function(x, ...) ...  
  
mean.default <- function(x, ...) ...
```


Vector classes

<https://adv-r.hadley.nz/S3.html>

Your turn

What is an **attribute**? What types of objects can have attributes?

How do you *get* the value of an attribute?

How do you *set* the value an attribute?

What's *the* most important attribute?

Attributes add arbitrary metadata to any object

```
x <- 1:6  
attr(x, "max") <- 5  
attr(x, "max")  
attributes(x)
```

```
# structure returns a modified object  
structure(1:10, min = 1, max = 10)
```

```
# Most important attribute is names()
```

Your turn

Every S3 class is built on a base type (e.g. a vector). The two most important S3 classes are factor and data frame.

What are **factors** built on top of?

What attributes do they use?

What are **data frames** built on top of?

What attributes do they use?

```
f <- factor(c("a", "b", "c"))
typeof(f)      # Built on top of integer
attributes(f)  # Use levels and class attributes

d <- data.frame(f)
typeof(d)      # Built on top of list
attributes(d)  # names, row.names and class
```


“Scalar” classes

Principle:

Provide consistent structure
and print method for
complex return values

Change working directory/project to:

[safely]

Challenge: how can improve the output of safely?

```
library(purrr)
safe_log <- safely(log)
```

```
safe_log("a")
#> result
#> NULL
#>
#> $error
#> <simpleError in log(...):
#>   non-numeric argument to
#>   mathematical function>
```

```
safe_log(10)
#> $result
#> [1] 2.302585
#>
#> $error
#> NULL
```

1. Figure out name
2. Define properties of the class
3. Write the constructor
4. Write methods

Your turn

What are the constraints on the results of safety?

Now, write the constructor

```
new_safely <- function(result = NULL, error = NULL) {  
  if (!xor(is.null(result), is.null(error))) {  
    stop(  
      "One of `result` and `error` must be NULL",  
      call. = FALSE  
    )  
  }  
  
  structure(  
    list(  
      result = result,  
      error = error  
    ),  
    class = "safely"  
  )  
}
```

Most S3 classes will
have this form

Then use the constructor

```
safely <- function(.f) {  
  stopifnot(is.function(.f))  
  
  function(...) {  
    tryCatch({  
      new_safely(result = .f(...))  
    }, error = function(e) {  
      new_safely(error = e)  
    })  
  }  
}
```

How could we test the constructor?

Abbreviation	Test
<code>expect_null()</code>	Checks if a literal NULL
<code>expect_type()</code> <code>expect_s3_class()</code> <code>expect_s4_class()</code>	Check that inherits from a given base type, S3 class, or S4 class.
<code>expect_true()</code> <code>expect_false()</code>	Catch all expectations for anything not otherwise covered

Your turn

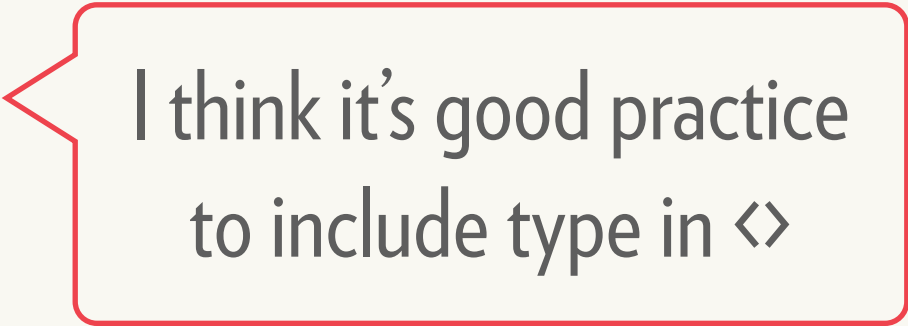
Write tests to ensure that our new `safely()` function returns the correct type of output regardless of whether or not an error occurs.

Now we can improve the output with a print method

```
safe_log(10)
```

```
#> <safely: ok>
```

```
#> [1] 2.302585
```



I think it's good practice
to include type in <>

```
safe_log("a")
```

```
#> <safely: error>
```

```
#> Error: non-numeric argument to
```

```
#> mathematical function
```

S3 methods all have the same basic structure

generic.class

Same arguments as generic

```
print.safely <- function(x, ...) {
```

```
}
```

Methods belong to
functions, not classes

Your turn: fill in the blanks

```
print.safely <- function(x, ...) {
```

```
}
```

```
# Useful helper found in utils.R
```

```
cat_line <- function(...) {
```

```
  cat(..., "\n", sep = "")
```

```
}
```

```
# See https://github.com/r-lib/cli for
```

```
# many more helpers.
```

My print method

```
print.safely <- function(x, ...) {  
  if (is.null(x$result)) {  
    cat_line("<safely: error>")  
    cat_line("Error: ", x$error$message)  
  } else {  
    cat_line("<safely: ok>")  
    print(x$result)  
  }  
}
```

```
invisible(x)  Called primarily for side-effects  
}
```

How do we test printing code?

```
expect_output(  
  new_safely(result = 1:10),  
  "<safely: ok>"  
)
```

How do we test printing code?

```
expect_output(  
  print(new_safely(result = 1:10)),  
  "<safely: ok>"  
)
```


But this is tedious & error prone

```
# It's hard to define precisely what the output
# should be (but we know it when we see it)

# So instead we can use a regression test.
# This is a weaker guarantee than a unit test:
# we'll just get alerted when it changes

expect_known_output(
    new_safely(result = 1:10),
    test_path("safely-ok.txt"),
    print = TRUE
)
```

A little colour can be transformative

```
print.safely <- function(x, ...) {  
  if (is.null(x$result)) {  
    cat_line("<safely: ", crayon::bold(crayon::red("error")), ">")  
    cat_line(crayon::red("Error: "), x$error$message)  
  } else {  
    cat_line("<safely: ", crayon::green("ok"), ">")  
    print(x$result)  
  }  
  
  invisible(x)  
}
```

Parameter objects

<https://refactoring.guru/introduce-parameter-object>

Principle:

Extract repeated groups of arguments into an object

<https://www.refactoring.com/catalog/introduceParameterObject.html>

Your turn: brainstorm problems

strsplit(x, split, fixed = FALSE, perl = FALSE, useBytes = FALSE)

grep(pattern, x, ignore.case = FALSE, perl = FALSE, value = FALSE, fixed = FALSE, useBytes = FALSE, invert = FALSE)

grepl(pattern, x, ignore.case = FALSE, perl = FALSE, fixed = FALSE, useBytes = FALSE)

sub(pattern, replacement, x, ignore.case = FALSE, perl = FALSE, fixed = FALSE, useBytes = FALSE)

gsub(pattern, replacement, x, ignore.case = FALSE, perl = FALSE, fixed = FALSE, useBytes = FALSE)

regexpr(pattern, text, ignore.case = FALSE, perl = FALSE, fixed = FALSE, useBytes = FALSE)

gregexpr(pattern, text, ignore.case = FALSE, perl = FALSE, fixed = FALSE, useBytes = FALSE)

regexec(pattern, text, ignore.case = FALSE, perl = FALSE, fixed = FALSE, useBytes = FALSE)

The arguments of `grepl()` interact in complex ways

	perl = FALSE	perl = TRUE
fixed = FALSE	POSIX 1003.2	perl
fixed = TRUE	exact matching	exact matching (with warning)

And `ignore.case` does not apply to exact matching

stringr takes a different approach

```
str_replace(x, fixed("y"))
```

```
str_replace(x, regex("."))
```

```
str_replace(x, boundary("word"))
```

```
# Allows different modes to have different
```

```
# arguments and hides details until you need
```

```
# to know about them
```

```
fixed <- function(pattern, ignore_case = FALSE) {  
  structure(  
    pattern,  
    ignore_case = ignore_case,  
    class = c("fixed", "pattern", "character")  
  )  
}
```



```
type <- function(x) UseMethod("type")

type.boundary <- function(x) "bound"

type.regex <- function(x) "regex"

type.coll <- function(x) "coll"

type.fixed <- function(x) "fixed"

type.character <- function(x) {
  if (identical(x, "")) "empty" else "regex"
}
```

```
# In str_replace
str_detect <- function(string, pattern) {
  switch(type(pattern),
    empty = ,
    bound = str_count(string, pattern) > 0,
    fixed = stri_detect_fixed(...),
    coll = stri_detect_coll(...),
    regex = stri_detect_regex(...)
  )
}
```

Learning more

Advanced R (2nd ed) has four chapters

S3: <https://adv-r.hadley.nz/s3.html>

S4: <https://adv-r.hadley.nz/s4.html>

R6: <https://adv-r.hadley.nz/r6.html>

Trade-offs: <https://adv-r.hadley.nz/oo-tradeoffs.html>

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