

OO programming

Jan 2019

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sorry, I forgot this one

```
install.packages("sloop")
```

Adapted from *Tidy Tools* by Hadley Wickham



What is S3?

What does **S3** do?

S3 powers context specific behaviour

```
x <- 1:5  
y <- factor(letters[1:5])
```

A 6-number
summary

```
summary(x)
```

```
#      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   
#           1         2         3         3         4         5
```

```
summary(y)
```

```
# a b c d e   
# 1 1 1 1 1
```

A table of
categories

summary() is an S3 generic

```
sloop::ftype(summary)
```

```
# [1] "S3"          "generic"
```

```
# summary() will look for methods based on an  
# objects class
```

```
sloop::s3_class(y)
```

```
# [1] "factor"
```

```
sloop::s3_dispatch(summary(x))
```

```
# => summary.factor      => this method gets called
```

```
# * summary.default      * this method exists but wasn't called
```

Your Turn

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

What is the **class** of mod?

Which **method** gets dispatched by summary()?

Can you find the code for the method?

```
library(sloop)
mod <- lm(mpg ~ wt, data = mtcars)
summary(mod)
```

```
s3_class(mod)
# [1] "lm"
```

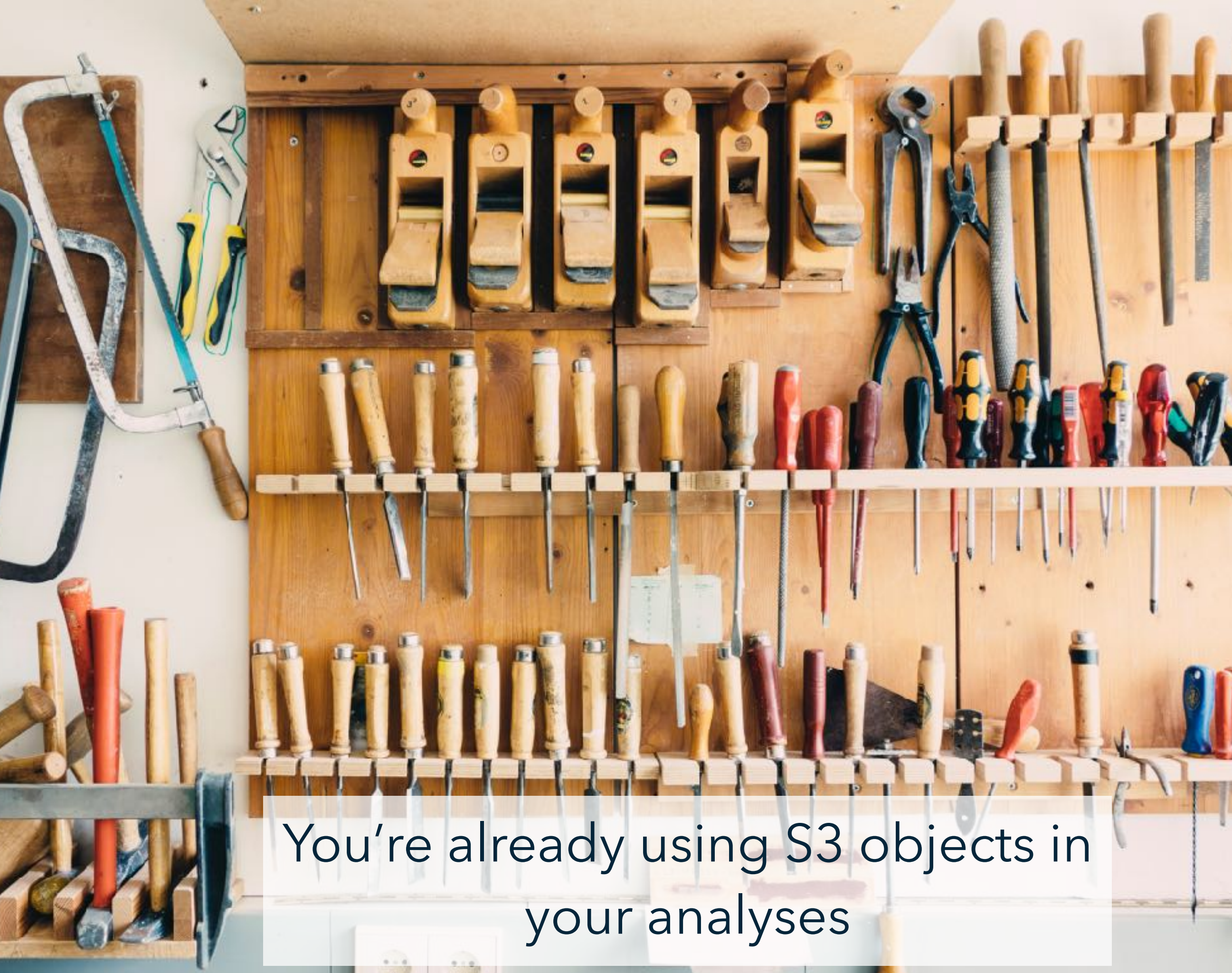
```
s3_dispatch(summary(mod))
# => summary.lm
# * summary.default
```

```
summary.lm
```

```
# won't always work
# use `s3_get_method()` to find non-exported methods
s3_get_method(summary.lm)
```

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()`

A dense, chaotic pile of various tools including saws, pliers, screwdrivers, and knives, illustrating the concept of complex functions.

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

Variable
type

```
# 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   3.89   3.84   2.31
3  0.230 Good       E      VS1     56.9  65.0   327   4.05   4.07   2.31
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
# ... with 53,930 more rows
```

Only shows first 10 rows

S3 makes packages extensible

New methods

Lets you extend other packages

New generics

Write packages in way that others
can easily extend.

“Scalar” classes

a single complex object

Principle:

Provide consistent structure
and print method for
complex return values

Change project to:

[safely]

Challenge: how can the output of safely be improved?

```
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
```

Creating a new S3 class

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

Your turn

What are the invariants of the results of safely?

```
safe_log <- purrr::safely(log)  
# what do we know to be always true  
# about the result of safe_log?
```

Invariants

Returns a list

- two components: result and error
- result should always come first
- one is always NULL

Now, write the constructor

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

Check
inputs

```
structure(  
  list(  
    result = result,  
    error = error  
  ),  
  class = "safely"  
)  
}
```

Enforce
structure and
apply class

Definition of safely

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

Then use the constructor

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

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. (i.e. express the invariants as unit tests)

```
# In tests/testthat/test-safely.R
context("test-safely.R")
```

```
test_that("can only supply error or result", {
  expect_error(new_safely(1, 2), "must be NULL")
})
```

```
test_that("it's ok for both to be null", {
  expect_error(new_safely(NULL, NULL), NA)
})
```

Expect no error

```
test_that("result and error are captured", {
  s1 <- new_safely(result = 1)
  s2 <- new_safely(error = 1)

  expect_s3_class(s1, "safely")
  expect_equal(s1$result, 1)
  expect_equal(s1$error, NULL)

  expect_s3_class(s2, "safely")
  expect_equal(s2$result, NULL)
  expect_equal(s2$error, 1)
})
```


Now we can improve the output with a print method

```
safe_log(10)
#> <safely: ok>
#> [1] 2.302585
```

```
safe_log("a")
#> <safely: error>
#> Error: non-numeric argument to
#> mathematical function
```



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

S3 methods all have the same basic structure

generic

Same arguments as generic

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

class

```
}
```

Methods belong to
generics, not
classes

	Date	POSIXct	integer
print			
mean			
sum			

	Date	POSIXct	integer
print			
mean			
sum			

Your turn: fill in the blanks

```
# In R/safely.R
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.
```

Some test cases

```
f <- function() stop("message")
```

```
g <- function() 1
```

```
safe_f <- safely(f)
```

```
safe_g <- safely(g)
```

```
safe_f()
```

```
safe_g()
```

My print method

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

```
invisible(x)  
}
```

Called primarily for
side-effects

A little colour can be transformative

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

New generic

Change project to:

[bizarro]

Goal: create a bizarro function

```
bizarro("abc")
```

```
#> [1] "cba"
```

```
bizarro(1)
```

```
#> [1] -1
```

```
bizarro(c(TRUE, FALSE))
```

```
#> [1] FALSE TRUE
```

We could use if + else

```
str_reverse <- function(x) {  
  purrr::map_chr(stringr::str_split(x, ""),  
    ~ stringr::str_flatten(rev(.x))  
  )  
}
```

```
bizarro <- function(x) {  
  if (is.character(x)) {  
    str_reverse(x)  
  } else if (is.numeric(x)) {  
    -x  
  } else if (is.logical(x)) {  
    !x  
  } else {  
    stop(  
      "Don't know how to make bizzaro <", class(x)[[1]], ">",  
      call. = FALSE)  
    }  
}
```

But instead we'll create a new S3 generic

```
bizarro <- function(x) {  
  UseMethod("bizarro")  
}
```

Magically passes all
arguments to correct
method

generic.class

```
bizarro.character <- function(x) {  
  str_reverse(x)  
}
```

```
bizarro("abc")
```

```
#> [1] cba
```

Allows anyone to extend

Your turn

Implement:

1. a numeric method that multiplies by -1
2. a logical method which inverts TRUE/FALSE
3. a data frame method that bizzarros the column names, as well as each column.

(i.e. get tests passing)

```
bizarro.numeric <- function(x) {  
  -x  
}
```

```
bizarro.logical <- function(x) {  
  !x  
}
```

```
bizarro.data.frame <- function(x) {  
  names(x) <- bizarro(names(x))  
  x[] <- purrr::map(x, bizarro)  
  x  
}
```

Useful technique Method for complex object: apply generic to components.

What happens when a method isn't available?

```
bizarro(factor(letters))
```

```
#> Error in UseMethod("bizarro") :
```

```
#> no applicable method for 'bizarro'
```

```
#> applied to an object of class "factor"
```

```
# How can we do better?
```

```
# We need to provide a catch-all default  
method
```

```
bizarro.default <- function(x) {  
  stop(  
    "Don't know how to make bizzaro <",  
    class(x)[[1]], ">",  
    call. = FALSE  
  )  
}
```

```
bizarro(factor(letters))
```

```
#> Error: Don't know how to make
```

```
#> bizzaro <factor>
```

Your turn

What should `bizzaro(factor("abc"))` return?

Decide, encode your decisions in tests, then implement `bizarro.factor()`.

One idea: reverse letters in factor levels

```
# In tests/testthat/test-bizarro.R
```

```
test_that("bizarro factors have levels reversed", {  
  f1 <- factor(c("abc", "def", "abc"))  
  f2 <- factor(c("cba", "fed", "cba"))  
  
  expect_equal(bizarro(f1), f2)  
  expect_equal(bizarro(f2), f1)  
})
```

```
# In R/bizarro.R
```

```
bizarro.factor <- function(x) {  
  levels(x) <- bizarro(levels(x))  
  x  
}
```

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>

Vector classes

Classes built on top of vector types



"vctrs will typically be used by other packages, making it easy for them to provide new classes of S3 vectors that are supported throughout the tidyverse (and beyond)."

--<https://vctrs.r-lib.org/>

Adapted from *Tidy Tools* by Hadley Wickham

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