

# OO programming

**Hadley Wickham**

Chief Scientist

RStudio

**January 2015**

1. Motivation

2. Warmup

3. Classes

4. Inheritance

5. Methods

6. Generic functions

7. Other OO systems

# Motivation

# Motivation

- Understanding more code
- Extensibility
- Programming “in the large”
- Focus on S3 because it is most commonly used, and solves most problems.

```
print(1:10)
print(mtcars)
print
# How does print work?
```

```
mean
# How does mean work?
```

**Warmup**

# Special names

How do you see the source code of the `%in%` function? What about `diag<-`?

If you have a column named “Income (\$1000)” in your data frame called `df`, how can you access it with `$`?

`%in%`

`diag<-`

df\$`Income (\$1000)`



# 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 are the three most important attributes?  
(Hint: what makes a matrix different from a vector?)

**Brainstorm with your neighbours for 2 minute.**

```
# Attributes add arbitrary additional  
# metadata to any base type
```

```
x <- 1:6
```

```
attr(x, "max") <- 5
```

```
attr(x, "max")
```

```
attributes(x)
```

```
# structure returns a modified object with attrs  
structure(1:10, max = 5)
```

```
# Most important attributes are dim, class and  
# names. Should always use dim(), class() and  
# names() respectively to get and set their  
# values.
```

# Your turn

Every S3 class is built on a basic type (like character, integer, function, list, ...). 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
```

Class	Base type	class()
Data frame	list	data.frame
Factor	integer	factor
Date	integer	Date
Time	double	POSIXct, POSIXt
Linear model	list	lm

# Classes

# Goal

Make a class that allows us to easily work with discrete random variables.

A discrete rv connects probabilities to numbers. Probabilities all greater than 0 and add to one; finite number of numbers.

Want to be able to plot, sample, take expectations, compute probabilities, combine etc.

# Mean & variance

The **mean** summarises the “middle” of the distribution. Mean =  $E(X)$  = “Sum” of all outcomes, weighted by their probability.

x	-1	0	1	2	3
P(x)	0.2	0.1	0.3	0.1	0.3



```
source("rv.r")
```

```
dice <- rv(1:6)
```

```
mean(dice)
```

```
min(dice)
```

```
max(dice)
```

```
range(dice)
```

```
P(dice > 3)
```

```
plot(dice + dice + dice)
```

# S3

No formal definition of the attributes that make up a class.

Instead, just set `class()` attribute of base type.

The simplest OO system that might possibly work. Adequate for 95% of R programming.

```
# No checks for object correctness, so easy to abuse
mod <- lm(log(mpg) ~ log(displacement), data = mtcars)
class(mod)
mod
```

```
class(mod) <- "table"
mod
```

```
# But surprisingly, this doesn't cause that
# many problems - instead of the language enforcing
# certain properties you need to do it yourself
```

# Random variables

What two things do we need to store to model a discrete random variable? What constraints are there on them?

How might you store them? (There are at least three ways)

x	-1	0	1	2	3
P(x)	0.2	0.1	0.3	0.1	0.3

```
x <- c(-1, 0, 1, 2, 3)
```

```
p <- c(0.2, 0.1, 0.3, 0.1, 0.3)
```

```
# Constraints
```

```
all(p > 0)
```

```
sum(p) == 1
```

```
abs(sum(p) - 1) < 1e-6
```

```
# Ways to store
```

```
structure(x, prob = p)
```

```
structure(p, val = x)
```

```
list(x = x, p = p)
```

```
# Start by defining a constructor function. It
# uses structure to set the class attribute.
rv <- function(x, probs = NULL) {
  if (is.null(probs)) {
    probs <- rep(1, length(x)) / length(x)
  }
  structure(x, probs = probs, class = "rv")
}
```

```
# Also customary to create function to test if  
# an object is of that class:
```

```
is.rv <- function(x) {  
  inherits(x, "rv")  
  # equivalent to "rv" %in% class(x)  
}
```

```
# And we'll also write a helper to extract  
# the probabilities
```

```
probs <- function(x) attr(x, "probs")
```

# Your turn

What's wrong with the following objects?

```
rv(1:3, c(-1, 2))
```

```
rv(c(1, 1), c(0.5, 0.5))
```

Can you fix `rv()` to prevent these problems from occurring?



```
rv <- function(x, probs = NULL) {  
  if (is.rv(x)) x <- as.numeric(x)  
  if (is.null(probs)) {  
    probs <- rep(1, length(x)) / length(x)  
  } else {  
    if (any(probs < 0)) stop("Probabilities must be positive")  
    if (abs(sum(probs) - 1) > 1e-6) stop("Probabilities must sum to 1")  
  }  
}
```

Strict

```
# Simplify by summing probabilities with equal x's. Need to use  
# addNA since otherwise tapply silently drops groups with missing values  
grp <- addNA(x, ifany = TRUE)  
x_new <- as.vector(tapply(x, grp, "[", 1))  
probs <- as.vector(tapply(probs, grp, sum))
```

Helpful

```
# Set probs and class attributes  
structure(x_new, probs = probs, class = "rv")  
}
```

# Methods

# Methods

- To make a class act differently, need to supply **methods** for **generic functions**.
- Most commonly provided methods are for: `print()` (202!), `format()` (63), `summary()` (32), `as.data.frame()`, `plot()`

```
# See what methods are defined for data.frame  
# and factor  
methods(class = "data.frame")  
methods(class = "factor")
```

```
`[.factor`  
print.factor  
getS3method("[", "factor")
```

```
# See what methods are defined for print and summary  
methods("print")  
methods("summary")
```

Methods belong  
to **functions**, not  
classes

```
# First method is usually a print method. Always  
# look at the generic first so that you can match  
# the arguments correctly.
```

```
print
```

```
# Can tell it's a generic function because it uses  
# UseMethod()
```

```
# Methods follow simple naming scheme
```

```
print.rv <- function(x, ...) {  
  ...  
}
```

# Your turn

Fill in the template to create a print method for rv objects.

```
print.rv <- function(x, ...) {  
  X <- format(x, digits = 3)  
  P <- format(probs(x), digits = 3)  
  out <- cbind(X = X, "P(X)" = P)  
  rownames(out) <- rep("", nrow(out))  
  print(out, quote = FALSE)  
}
```

```
dice <- rv(1:6)  
plot(dice)
```



```
# Another common method is plot
plot.rv <- function(x, ...) {
  name <- deparse(substitute(x))
  ylim <- range(0, probs(x))

  plot(as.numeric(x), probs(x), type = "h", ylim = ylim,
       xlab = name, ylab = paste0("P(", name, ")"), ...)
  points(as.numeric(x), probs(x), pch = 20)
  abline(h = 0, col = "gray")
}
```

# Your turn

Implement a mean method.

```
mean.rv <- function(x, ...) {  
  sum(x * probs(x))  
}
```

# Inheritance

# S3 inheritance

- Multiple elements in class attribute.
- First looks method for first class, then second, and so on.
- Then looks for method for *implicit* class.
- Then looks for default method.

Class	class()	Implicit class
Time	POSIXct, POSIXt	double
Generalised linear model	glm, lm	list
Data frame	data.frame	list

```
iclass <- function(x) {
  c(
    if (is.matrix(x)) "matrix",
    if (is.array(x) && !is.matrix(x)) "array",
    if (is.double(x) || is.integer(x)) "numeric",
    typeof(x)
  )
}

method_names <- function(generic, x) {
  paste0(generic, ".", c(class(x), iclass(x), "default"))
}

s3_dispatch <- function(call) {
  call <- substitute(call)
  generic <- as.character(call[[1]])
  object <- eval(call[[2]], parent.frame())
  methods <- method_names(generic, object)
  exists <- vapply(methods, exists, logical(1))
  cat(paste0(ifelse(exists, "*", " "), " ", methods,
    collapse = "\n"), "\n", sep = "")
}
```

```
x <- Sys.time()
s3_dispatch(print(x))
s3_dispatch(is.numeric(x))
s3_dispatch(as.Date(x))
```

```
x <- 1:10
class(x) <- c("b", "a")
s3_dispatch(print(x))
```

```
print.b <- function(x) cat("B!")
s3_dispatch(print(x))
```



```
dice <- rv(1:6)
```

```
# Why do these work?
```

```
min(dice)
```

```
range(dice)
```

```
# What's wrong with these?
```

```
dice * 2
```

```
abs(dice)
```

```
abs(dice - 2)
```

```
dice[1:3]
```

# Inheritance

Want to use the default behaviour for `abs`, `[]`, etc.

`NextMethod()` call the next method in the sequence.

You never supply arguments - it uses non-standard evaluation magic to figure them out.

```
x <- 1:10  
class(x) <- c("c", "b", "a")
```

```
print.c <- function(x) {  
  cat("c\n")  
  NextMethod()  
}  
s3_dispatch(print(x))
```

```
rm(print.b)
```

```
print.a <- function(x) {  
  cat("a\n")  
  NextMethod()  
}
```

```
s3_dispatch(print(x))  
print(x)
```

```
`[.rv` <- function(x, i, ...) {  
  rv(NextMethod(), prop.table(probs(x)[i]))  
}
```

```
abs.rv <- function(x) {  
  rv(NextMethod(), probs(x))  
}
```

```
# What would methods for sqrt, log and exp  
# look like?
```

```
abs.rv <- function(x)  rv(NextMethod(), probs(x))  
log.rv <- function(x)  rv(NextMethod(), probs(x))  
exp.rv <- function(x)  rv(NextMethod(), probs(x))  
sqrt.rv <- function(x) rv(NextMethod(), probs(x))
```

```
# Can use shortcut with a "group" generic
```

```
Arith.rv <- function(x) rv(NextMethod(), probs(x))
```

```
# Relatively advanced technique and unnecessary
```

```
# but does save some typing
```

# Your turn

Look at `rv.R`. What other methods are implemented? What do they do?

# **Generic functions**

# S3 generics

- As well as creating methods for existing generics, you can also create your own.
- Creating a generic is very very simple!
- Just call `UseMethod("generic name")`.
- Other arguments figured out by NSE magic.



```
# Creating your own generics
mean2 <- function (x, ...) {
  UseMethod("mean2")
}
```

```
mean2.numeric <- function(x, ...) sum(x) / length(x)
mean2.data.frame <- function(x, ...)
  sapply(x, mean2, ...)
mean2.matrix <- function(x, ...) apply(x, 2, mean)

mean2.default <- function(x, ...) {
  stop("mean2 not implemented for objects of type ",
    class(x))
}
```

# Namespacing

In Java/C#/Ruby/Python etc., often have many small methods, even if only used by one class.

This is not useful in R – only useful to define methods that are used by multiple classes.

Use namespaces (tomorrow) for the equivalent encapsulation.

**Learning  
more**

# S4

Same basic style as S3, but formal and rigorous (and verbose).

`setClass()` defines classes.

`setGeneric()` defines generic functions.

`setMethod()` defines methods.

# R6

Package inspired by ReferenceClasses, but much faster & fixes major problem.

Class-based (message passing) OO.  
Much closer to Java/C#/Python/Ruby etc.

Have mutable state.

All methods/fields are public.

# Recommendations

- Use **S3**, unless:
- You have complex network of classes and methods - use **S4**.
- You have objects with changing state - use **R6**.

<http://adv-r.had.co.nz/00-essentials.html>

<https://www.google.com/search?q=bioconductor+s4>

library(pryr)

?otype

?ftype





This work is licensed under the Creative Commons Attribution-Noncommercial 3.0 United States License. To view a copy of this license, visit <http://creativecommons.org/licenses/by-nc/3.0/us/> or send a letter to Creative Commons, 171 Second Street, Suite 300, San Francisco, California, 94105, USA.