

○○ programming

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Warmup

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?

Attributes add arbitrary metadata to any object

```
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 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
```

Your turn

What is a **generic function**?

What makes a generic function generic?

How is it different from a regular function?

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

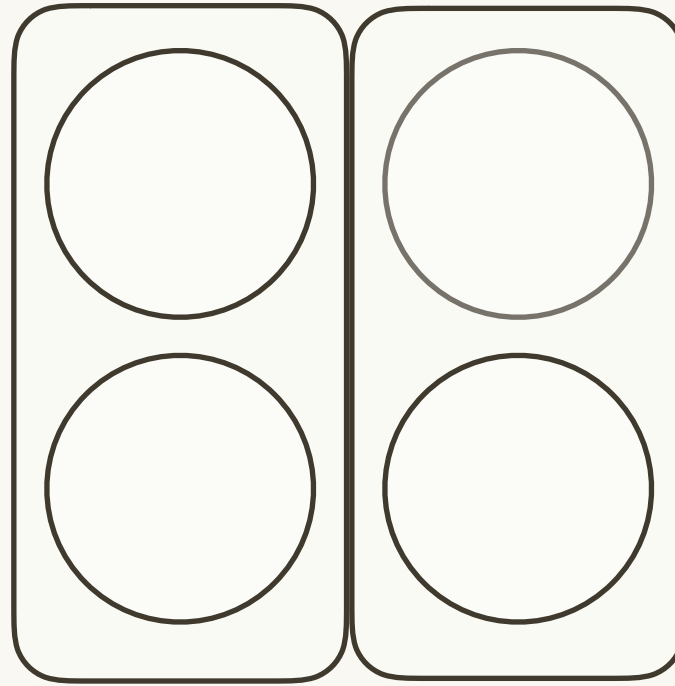
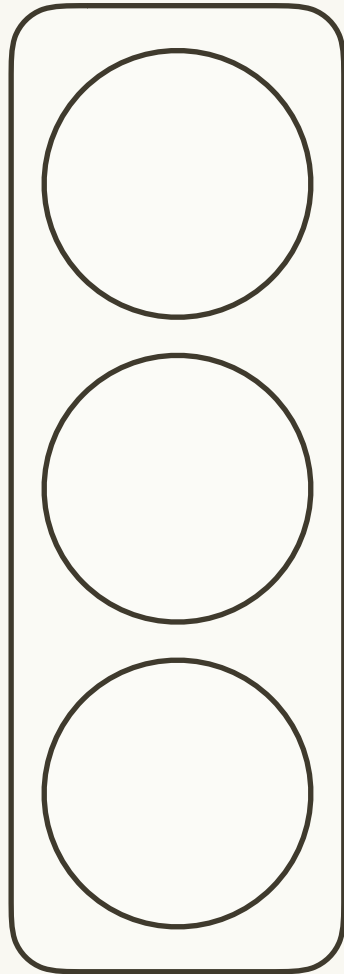
```
mean <- function(x, ...) {  
  fun <- paste("mean", class(x), sep = ".")  
  match.fun(fun)(x, ...)  
}
```

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

Motivation

Why should you care about **S3**?

Complex functions need to return multiple things



S3 lets you control how your objects are printed

	carat	cut	color	clarity	depth	table	price	x	y	z
1	0.23	Ideal	E	SI2	61.5	55.0	326	3.95	3.98	2.43
2	0.21	Premium	E	SI1	59.8	61.0	326	3.89	3.84	2.31
3	0.23	Good	E	VS1	56.9	65.0	327	4.05	4.07	2.31
4	0.29	Premium	I	VS2	62.4	58.0	334	4.20	4.23	2.63
5	0.31	Good	J	SI2	63.3	58.0	335	4.34	4.35	2.75
6	0.24	Very Good	J	VVS2	62.8	57.0	336	3.94	3.96	2.48
7	0.24	Very Good	I	VVS1	62.3	57.0	336	3.95	3.98	2.47
8	0.26	Very Good	H	SI1	61.9	55.0	337	4.07	4.11	2.53
9	0.22	Fair	E	VS2	65.1	61.0	337	3.87	3.78	2.49
10	0.23	Very Good	H	VS1	59.4	61.0	338	4.00	4.05	2.39
11	0.30	Good	J	SI1	64.0	55.0	339	4.25	4.28	2.73
12	0.23	Ideal	J	VS1	62.8	56.0	340	3.93	3.90	2.46
13	0.22	Premium	F	SI1	60.4	61.0	342	3.88	3.84	2.33
14	0.31	Ideal	J	SI2	62.2	54.0	344	4.35	4.37	2.71
15	0.20	Premium	E	SI2	60.2	62.0	345	3.79	3.75	2.27
16	0.32	Premium	E	I1	60.9	58.0	345	4.38	4.42	2.68
17	0.30	Ideal	I	SI2	62.0	54.0	348	4.31	4.34	2.68
18	0.30	Good	J	SI1	63.4	54.0	351	4.23	4.29	2.70
19	0.30	Good	J	SI1	63.8	56.0	351	4.23	4.26	2.71
20	0.30	Very Good	J	SI1	62.7	59.0	351	4.21	4.27	2.66
21	0.30	Good	I	SI2	63.3	56.0	351	4.26	4.30	2.71
22	0.23	Very Good	E	VS2	63.8	55.0	352	3.85	3.92	2.48
23	0.23	Very Good	H	VS1	61.0	57.0	353	3.94	3.96	2.41
24	0.31	Very Good	J	SI1	59.4	62.0	353	4.39	4.43	2.62
25	0.31	Very Good	J	SI1	58.1	62.0	353	4.44	4.47	2.59
26	0.23	Very Good	G	VVS2	60.4	58.0	354	3.97	4.01	2.41
27	0.24	Premium	I	VS1	62.5	57.0	355	3.97	3.94	2.47
28	0.30	Very Good	J	VS2	62.2	57.0	357	4.28	4.30	2.67
29	0.23	Very Good	D	VS2	60.5	61.0	357	3.96	3.97	2.40
30	0.23	Very Good	F	VS1	60.9	57.0	357	3.96	3.99	2.42

Plus 970 more rows

A good print method is very valuable

Total size

Source: local data frame [53,940 x 10]

	carat (dbl)	cut (fctr)	color (fctr)	clarity (fctr)	depth (dbl)	table (dbl)	price (int)	x (dbl)	y (dbl)	z (dbl)
1	0.23	Ideal	E	SI2	61.5	55	326	3.95	3.98	2.43
2	0.21	Premium	E	SI1	59.8	61	326	3.93	3.97	2.42
3	0.23	Good	E	VS1	56.9	65	327	4.05	4.07	2.51
4	0.29	Premium	I	VS2	62.4	58	334	4.20	4.23	2.63
5	0.31	Good	J	SI2	63.3	58	335	4.34	4.35	2.75
6	0.24	Very Good	J	VVS2	62.8	57	336	3.94	3.96	2.48
7	0.24	Very Good	I	VVS1	62.3	57	336	3.95	3.98	2.47
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9	0.22	Fair	E	VS2	65.1	61	337	3.87	3.78	2.49
10	0.23	Very Good	H	VS1	59.4	61	338	4.00	4.05	2.39
..

Variable type

Only shows first 10 rows

Over 100 lines of code!

A constructor ensures that you always return the same thing

```
doctor <- function(name, messages) {  
  structure(  
    length(messages) == 0,  
    doctor = paste0("DR_", toupper(name)),  
    messages = messages,  
    class = "doctor"  
  )  
}
```

Powers `dr_github()`
and `dr_devtools()`

```
print.doctor <- function(x, ...) {  
  if (x) {  
    message(attr(x, "doctor"), " SAYS YOU LOOK HEALTHY")  
    return()  
  }  
}
```

```
warning(attr(x, "doctor"), " FOUND PROBLEMS", call. = FALSE, immediate. = TRUE)  
messages <- strwrap(attr(x, "messages"), exdent = 2)  
message(paste(messages, collapse = "\n"))  
}
```

S3 makes packages extensible

New methods

Lets you extend other packages

New generics

Write packages in way that others can easily extend.

Complex
return values

Your turn

How is a date represented? How is a date time represented? (what do the values *mean*? `unclass(x)`)

How is a linear model represented?

```
x1 <- as.Date("1970-01-02")
```

```
x2 <- as.POSIXct(x1)
```

```
x3 <- lm(mpg ~ wt, data = mtcars)
```

Dates and date times are built on top of vectors

```
x <- as.Date("1970-01-02")
typeof(x)
attributes(x)
unclass(x)
# Number of days since Jan 1, 1970

y <- as.POSIXct(x)
typeof(y)
attributes(y)
unclass(y)
# Number of seconds since Jan 1, 1970
```

Linear models are built on top of lists

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

```
typeof(mod)
```

```
attributes(mod)
```

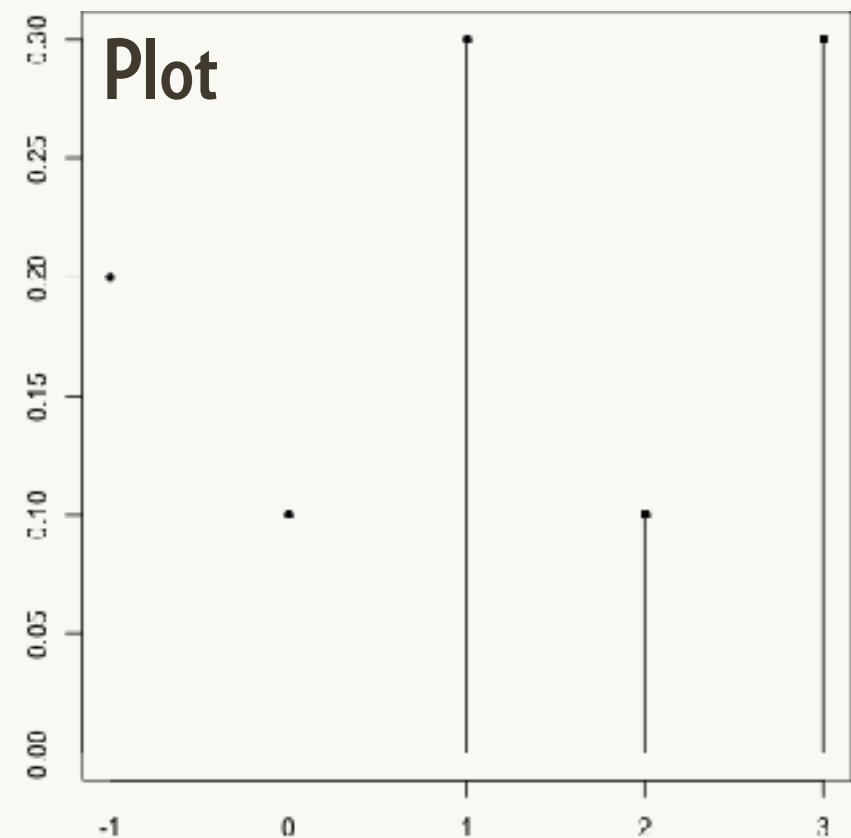
```
names(mod)
```

```
unclass(mod)
```

Goal: model random variables in R

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

1.9
Mean



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

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

```
mean(dice)
```

```
min(dice)
```

```
max(dice)
```

```
range(dice)
```

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

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

discreteRV explores these ideas in more depth

discreteRV	Casella and Berger
$E(X)$	$E(X)$
$P(X == x)$	$P(X = x)$
$P(X >= x)$	$P(X \geq x)$
$P((X < x_1) \%AND\% (X > x_2))$	$P(X < x_1 \cap X > x_2)$
$P((X < x_1) \%OR\% (X > x_2))$	$P(X < x_1 \cup X > x_2)$
$P((X == x_1) \mid (X > x_2))$	$P(X < x_1 \mid X > x_2)$
$\text{probs}(X)$	$f(x)$
$V(X)$	$Var(X)$

<http://journal.r-project.org/archive/2015-1/hare-buja-hofmann.pdf>

Representing a random variable

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)
```

How might we represent this random variable
as a single R object?

(There are at least five 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)
```

```
# Ways to store
```

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

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

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

```
data.frame(x, p)
```

```
# Use whichever makes your life easier
```

S3 is just a convention

Simplest OO system that might possibly work

No formal definition!

```
structure(x, prob = p, class = "rv")
```

```
# Or
```

```
rv <- structure(x, prob = p)
```

```
class(rv) <- "rv"
```

S3 is easy to abuse!

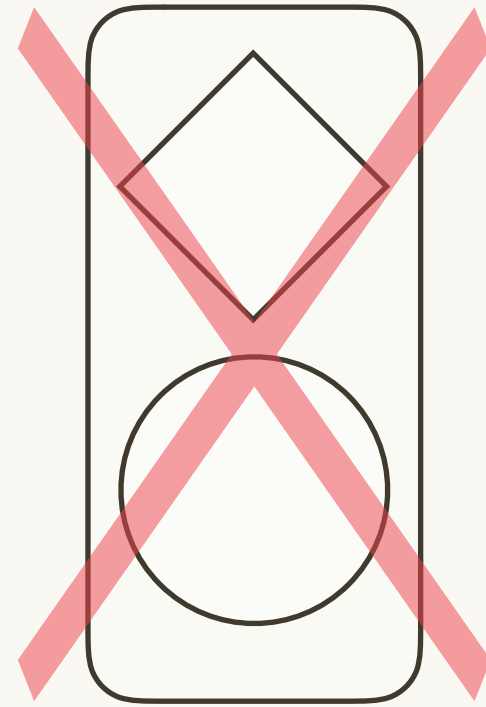
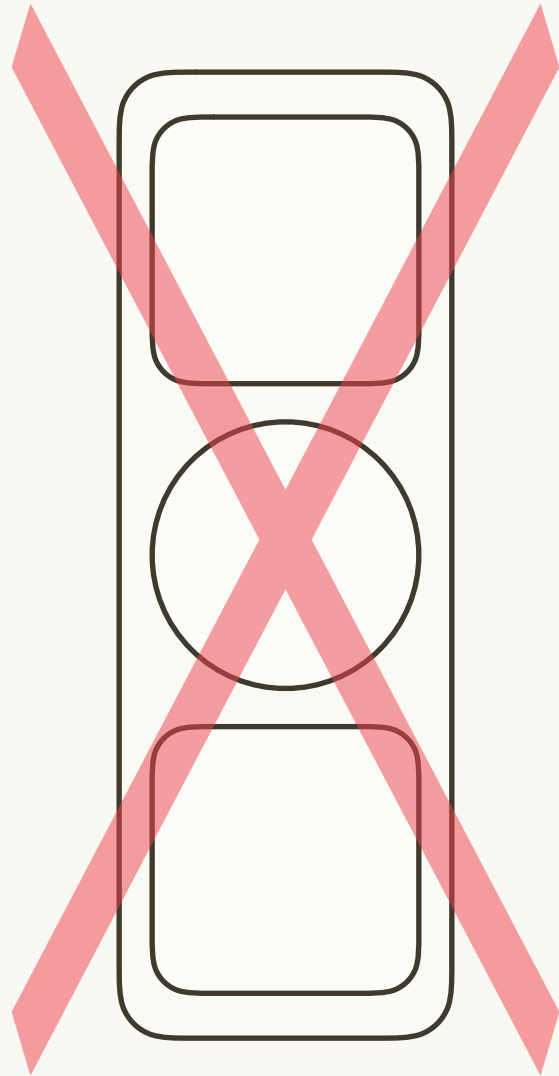
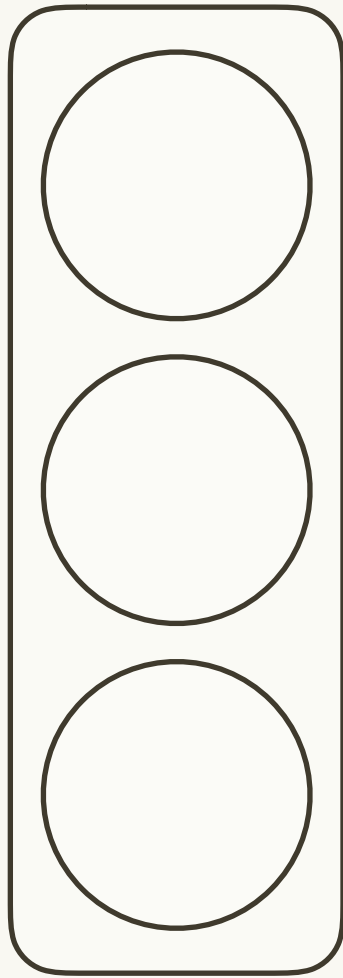
```
mod <- lm(log(mpg) ~ log(displacement), data = mtcars)
class(mod)
mod
```

```
class(mod) <- "data.frame"
mod
```

```
# Surprisingly, this doesn't cause many problems!
```

Constructors

A constructor ensures that you're consistent



Constructor functions can be helpful

```
rv <- function(x, probs = NULL) {  
  if (is.null(probs)) {  
    probs <- rep(1, length(x)) / length(x)  
  }  
  structure(x, probs = probs, class = "rv")  
}
```

```
# We'll also write a helper to extract  
# the probabilities  
probs <- function(x) attr(x, "probs")
```

Also useful to add an `is.*()` function

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

What's wrong with each of these rvs?

1

x	-1	0	1
P(x)	0.5	0.5	0.5

2

x	-1	0	1
P(x)	"a"	FALSE	😊

3

x	-1	0	1
P(x)	-0.25	1	0.25

4

x	-1	-1	-1
P(x)	0.5	0.4	0.1

5

x	-1	0	1
P(x)	0.33	0.67	

6

x	-1	0	1
P(x)	NA	0.5	0.5

How could you write code to detect the problem?

A constructor should also check its inputs

```
check_probs <- function(x) {  
  if (!is.numeric(x)) {  
    stop("`prob` must be numeric.")  
  }  
  if (any(is.na(x))) {  
    stop("`prob` must not contain any NA")  
  }  
  if (any(x < 0)) {  
    stop("All `prob` must be >= 0")  
  }  
  if (sum(x) != 1) {  
    stop("`sum(prob)` must equal 1")  
  }  
}  
x <- rep(1/49, 49)  
check_probs(x)
```

Beware the perils of floating point!

```
check_probs <- function(x) {  
  if (!is.numeric(x)) {  
    stop("`prob` must be numeric.")  
  }  
  if (any(is.na(x))) {  
    stop("`prob` must not contain any NA")  
  }  
  if (any(x < 0)) {  
    stop("All `prob` must be >= 0")  
  }  
  if (abs(sum(x) - 1) > 1e-6) {  
    stop("`sum(prob)` must equal 1")  
  }  
}  
  
x <- rep(1/49, 49)  
check_probs(x)
```

Need to balance strictness and helpfulness

```
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 (length(x) != length(probs)) stop("Values and probability...")  
    check_probs(probs)  
  }  
}
```

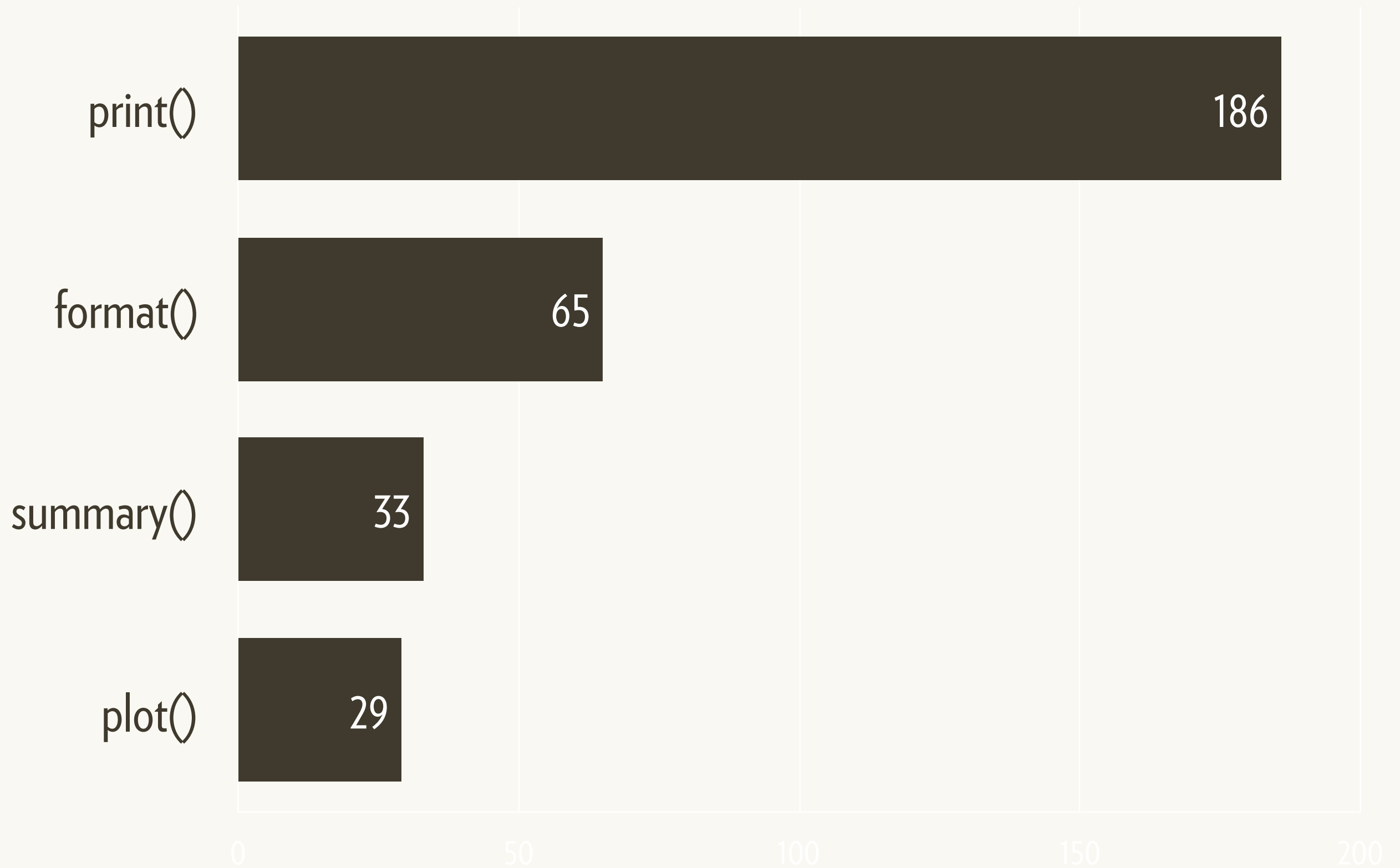
Strict

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

Helpful

Methods

Most common methods are to do with output



Use `methods()` to find methods!

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

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

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

Methods belong to
functions, not classes

Method calls look different in other languages

S3 and S4

```
method(object, 1, 2, 3)
```

Many other languages

```
object.method(1, 2, 3)
```

RC/R6

```
object$method(1, 2, 3)
```


	factor	Date	data frame
relevel	✓		
mean		✓	
rep	✓	✓	
print	✓	✓	✓

Message passing

	factor	Date	data frame
relevel	✓		
mean		✓	Generic function
rep	✓	✓	
print	✓	✓	✓

Methods belong to **functions**, not classes

This is called “generic function” style OO.
It’s also what Julia uses.

Print method is most important

```
# To write method, first identify generic and find  
# its arguments  
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 on the next slide to create a print method for rv objects.

Good print methods are really hard, so aim to get the important data out, even if it doesn't look great.

Fill in this template:

```
print.rv <- function(x, ...) {  
  cat("THIS IS MY METHOD\n")  
}
```

Prints text directly to console

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

`print()` is usually called automatically

This is my method:

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

What name should your function have? What arguments?

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

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

Inheritance

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

Extensibility

Write your own generics so others can extend

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

```
# Now anyone can extend your framework by  
# implementing methods for their objects.
```

```
# Documentation extremely important. Few  
# packages ever get popular enough to garner  
# significant contributions
```

Learning more

S4

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

`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.

Do not copy-on-modify. Has mutable state, or reference semantics.

Recommendations

Use **S3** unless:

You have complex network of classes and methods —
use **S4**.

You have objects with changing state — use **R6**.

Other resources

John Chambers, “*Extending R*”,

<https://amzn.com/1498775713>

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

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

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