Statistics 360: Advanced R for Data Science Lecture 1

Brad McNeney

Course Objectives

R objects

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Course objectives

- ► Make you R **programmers** rather than just R **users**, by working through most of the book "Advanced R" by Hadley Wickham: https://adv-r.hadley.nz/index.html
- Topics:
 - R objects: names and values
 - Basic data structures and programming.
 - vectors, subsetting, control flow, functions, environments
 - No tidyverse this time
 - R packages (based on the online text by Wickham and Bryan)
 - Object-oriented programming in R
 - Code performance: debugging, profiling, memory, calling Python, or C++ from R
 - Parallelizing R code (if time permits)

Getting started with R, RStudio and git

- ► Follow the "getting started' instructions on the class canvas page to get set up with R, RStudio and git.
 - ▶ R and RStudio will be familiar, but you may not have used git before, so leave some time for that.
- Please try to get R and RStudio installed and create an RStudio project linked to the class GitHub repository (or a forked copy) as soon as possible.
- ► Those still having trouble after the weekend should ask our TA, Pulindu, for help during the first lab sessions in week 2.
 - Note: No lab first week.

Reading

▶ Welcome, Preface and Chapters 1,2 of the text.

R objects

R objects

- ▶ In R, data structures and functions are all referred to as "objects".
- Objects are created with the assignment operator <-; e.g., x <- c(1,2,3).</p>
 - ► The objects a user creates from the R console are contained in the user's workspace, called the global environment.
 - ▶ Use ls() to see a list of all objects in the workspace.
 - Use rm(x) to remove object x from the workspace.

Digging deeper

- The above understanding is an over-simplification that is usually OK, but will sometimes lead to misunderstandings about memory usage and when R makes copies of objects
- Object copying is a major source of computational overhead in R, so it pays to understand what will trigger it.
- Reference: text, chapter 2

Binding names and to objects

► The R code x <- c(1,2,3) does two things: (i) creates an object in computer memory that contains the values 1, 2, 3 and (ii) "binds" that object to the "name" x.

```
# install.packages("lobstr")
library(lobstr)
x <- c(1,2,3)
ls()
## [1] "x"
obj_addr(x) # changes every time this code chunk is run
## [1] "0x1317f45e8"</pre>
```

Binding multiple names to the same object

► The following binds the name y to the same object that x is bound to.

```
y <- x
obj_addr(y)
## [1] "0x1317f45e8"
```

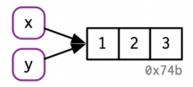


Figure 1: Bind two names to the same object

Aside: Syntactic vs non-syntactic

- ▶ Valid, or "syntactic" names in R can consist of letters, digits, . and _ but should start with a letter.
- Names that start with . are hidden from directory listing with 1s().
- Names that start with _ or a digit are non-syntactic and will cause an error.
- ▶ If you need to create or access a non-syntatic name, use backward single-quotes ("backticks").

```
x <- 1
.x <- 1
`_x` <- 1
ls()
```

```
## [1] "_x" "x" "y"
```

Modifying causes copying

Modifying a variable causes a copy to be made, with the modified variable name bound to the copy.

```
c(obj_addr(x),obj_addr(y))
## [1] "0x105b0cd90" "0x1317f45e8"
x[[2]] \leftarrow 10 \# Note: x[2] \leftarrow 10 \text{ has the same effect}
c(obj_addr(x),obj_addr(y))
  [1] "0x1314a9e88" "0x1317f45e8"
                                           0xcd2
```

Figure 2: Bindings after copying

Tracing copying

► The tracemem() function marks an object so that a message is printed whenever a copy is made.

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

## [1] "<0x10608f868>"

x[[2]] <- 10

## tracemem[0x10608f868 -> 0x130a98558]: eval eval withVisible w
untracemem(x) # remove the trace
x[[1]] <- 10</pre>
```

More on tracmem()

► As the output of tracemem() suggests, the trace is on the object, not the name:

```
x \leftarrow c(1,2,3)
tracemem(x)
## [1] "<0x1042d40b8>"
y <- x
c(obj_addr(x),obj_addr(y))
## [1] "0x1042d40b8" "0x1042d40b8"
y[[2]] \leftarrow 10
## tracemem[0x1042d40b8 -> 0x1308a9f88]: eval eval withVisible w
c(obj addr(x),obj addr(y))
```

[1] "0x1042d40b8" "0x1308a9f88"

Function calls

R has a reputation for passing copies to functions, but in fact the copy-on-modify applies to functions too:

```
f <- function(arg) { return(arg) }
x <- c(1,2,3)
y <- f(x) # no copy made, so x and y bound to same obj
c(obj_addr(x),obj_addr(y))

## [1] "0x10608e0a8" "0x10608e0a8"

f <- function(arg) { arg <- 2*arg; return(arg) }
y <- f(x) # copy made
c(obj_addr(x),obj_addr(y))

## [1] "0x10608e0a8" "0x1047771c8"</pre>
```

Lists

List elements point to objects too:

```
11 <- list(1, 2, 3)
c(obj_addr(11),obj_addr(11[[1]]),obj_addr(11[[2]]),obj_addr(11[[3]]))
## [1] "0x104798aa8" "0x105bb1390" "0x105bb1358" "0x105bb1320"
# Note: ref(l1) will print a nicely formatted version of the above,
# but doesn't work with my slides</pre>
```

Copy-on-modify in lists

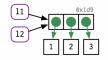


Figure 3: Bindings before

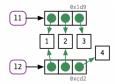


Figure 4: Bindings after modify

Copies of lists are said to be "shallow"

- As shown above, we copy the list itself and any list **elements** that are modified. This is called a "shallow" copy.
- By contrast, a "deep" copy would be a copy of all elements.

```
11 <- list(1,2,3)
c(obj_addr(l1),obj_addr(l1[[1]]),obj_addr(l1[[2]]),obj_addr(l1[[3]]))
## [1] "0x10608c378" "0x1059e1358" "0x1059e1320" "0x1059e12e8"
11[[3]] <- 4
c(obj_addr(l1),obj_addr(l1[[1]]),obj_addr(l1[[2]]),obj_addr(l1[[3]]))
## [1] "0x130a98d28" "0x1059e1358" "0x1059e1320" "0x1059e1160"</pre>
```

Data frames are lists with columns as list items

```
dd \leftarrow data.frame(x=1:3,y=4:6)
c(obj_addr(dd[[1]]),obj_addr(dd[[2]]))
## [1] "0x1045bc820" "0x1045bc740"
dd[,2] <- 7:9
c(obj_addr(dd[[1]]),obj_addr(dd[[2]])) # only changes second element
## [1] "0x1045bc820" "0x105846dd8"
dd[1,] \leftarrow c(11,22)
c(obj_addr(dd[[1]]),obj_addr(dd[[2]])) # changes to both elements
## [1] "0x1315cb898" "0x1315cb848"
dd[1,1] <- 111
c(obj_addr(dd[[1]]),obj_addr(dd[[2]])) # only changes first element
## [1] "0x131184228" "0x1315cb848"
```

Beware of data frame overhead

- Data frames are convenient, but the convenience comes at a cost.
 - ► For example, coercion to/from lists

```
dd <- data.frame(x=rnorm(100)) # try yourself with rnorm(1e7)
tracemem(dd); tracemem(dd[[1]])
## [1] "<0x1061614f0>"
## [1] "<0x127f6fcf0>"
dmed <- lapply(dd, median) # makes a list copy of dd
## tracemem[0x1061614f0 -> 0x1061dd870]: as.list.data.frame as.list lap
## tracemem[0x127f6fcf0 -> 0x127f71280]: sort.int sort.default sort mea
dd[[1]] <- dd[[1]] - dmed[[1]] #
## tracemem[0x1061614f0 -> 0x105b04f18]: eval eval withVisible withCall
## tracemem[0x105b04f18 -> 0x105b07a78]: [[<-.data.frame [[<- eval eval
```

Fewer copies if we do the same with a list.

```
11 <- list(x=rnorm(100))
tracemem(ll); tracemem(ll[[1]])

## [1] "<0x105bd35f8>"

## [1] "<0x127f9e940>"
lmed <- lapply(ll,median) # no need for a list copy

## tracemem[0x127f9e940 -> 0x127f9f6b0]: sort.int sort.default sort mea
ll[[1]] <- ll[[1]] - dmed[[1]]

## tracemem[0x105bd35f8 -> 0x1042d24a0]: eval eval withVisible withCall
```

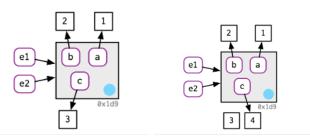
Modify-in-place

- ▶ The text says there are two exceptions to the copy-on-modify:
 - 1. Modify an element of an object with one binding, or
 - 2. Modify an environment. but in my experiments, only the second applies.

```
v \leftarrow c(1,2,3) # creates object (1,2,3) and binds v to it tracemem(v)
```

```
## [1] "<0x105aee008>"
v[[3]] <- 4 # for me, this triggers a copy
```

tracemem[0x105aee008 -> 0x106037518]: eval eval withVisible withCall



```
e1 <- rlang::env(a = 1, b = 2, c = 3)
e2 <- e1
# note: can't use tracemem() on an environment
e1$c <- 4
e2$c</pre>
```

[1] 4

Object size

Use lobstr::obj_size() to find the size of objects.
obj_size(dd)
1,528 B
obj_size(11)
1,128 B

```
## 840 B
obj_size(e2)
```

840 B

obj_size(e1)