

# Introduction to R

Stat 141

1

## R environment, language, ...

stat141

- R is three things
  - virtual machine/interpreter
  - a programming language
  - and a collection of built-in and add-on packages i.e. collections of useful functions.
- The R virtual machine interprets commands in the R language.
- We will not discuss the interpreter very much, except to look at some of the different ways we can run R commands and control the behaviour of the VM.

2

## R language

stat141

- R is a procedural, imperative programming language. executes one command, then another, etc.
- interpreted language
  - commands are not compiled into machine code of the chip/CPU
  - instead, a virtual machine (VM) processes each command
  - the virtual machine is a program/executable that interprets the commands and ends up calling compiled code to perform the actions.

3

## R language

stat141

- Interactive – don't need to run entire scripts or programs, but can give one command at a time and decide what to do next.
- However, one can combine a collection of commands into a file and have R process the entire file  
R CMD batch filename.R or `source('filename.R')`
- And can develop and distribute software in R for others to use.  
Vital to making statistics available to non-statisticians and disseminating new methodology.

4

## A little history

stat141

- R is a dialect of the S language  
So too is the commercial product S-Plus.
- The S language was developed in Bell Labs over the last 30 years.
  - originally in the 70's-80's by John Chambers, then with Rick Becker and Alan Wilks. (S3 - Blue Book)
  - in early 90's by John Chambers & Trevor Hastie and numerous colleagues and visitors to Bell Labs (Statistical Models - White book)
  - in 93-98, by John Chambers (and DTL) (S4 - Green Book)

5

## ACM Award

stat141

- In 1998, John Chambers won the Association of Computing Machinery's Software Award for S
- The citation  
*...forever altered how people analyze, visualize, and manipulate data  
... S is an elegant, widely accepted, and enduring software system,  
with conceptual integrity*
- S is the de facto standard for statistical research and sophisticated data analysis.

6

## Motivation for S

stat141

- FORTRAN - Formula Translation language - was developed to allow scientists to express computations to the computer as if they were regular mathematics.
- S grew out of the desire to express statistical calculations and operations to manipulate data and perform data analyses in a natural, convenient, expressive manner that kept the focus on the analysis and thought process of the statistics rather than on "programming" the computer.
- Before S, most analyses involved a lot of FORTRAN code to do basic analysis.

7

- Standard analysis steps were gathered into reusable libraries of code to avoid having to recreate the code each time or manage the code in different files.
- But in Bell Labs, most analysis were non-standard because they were working on cutting edge problems.
- As a result, there was still a need to write a lot of FORTRAN code and much of it was manipulating data before it could be passed to 3rd party matrix routines.
- S provides an interactive language for managing data, and organizing the order and steps of computations.

stat141

8

### ⌚ The interpreter

- ⌚ reads a command (text)
- ⌚ parses it into a data structure that represents the elements and structure of the command internally
- ⌚ evaluates the command elements and overall command
- ⌚ prints the results (if appropriate)
- ⌚ return to wait for next command.
- ⌚ this is called a REPL - read, eval, print loop.

- ⌚ S is written in C and much of the built-in functionality is also written in C or FORTRAN.
- ⌚ And S provides facilities to call code in other languages such as C, FORTRAN, C++, Objective-C, Java, Perl, Python, the Shell, Matlab, ...
- ⌚ And this allows S to be used as a language to glue low-level computations together in new and different ways specified by each user.
- ⌚ E.g. can add an extra data transformation step before calling an existing algorithm.

- ⌚ In addition to being able to call built-in and externally added code in "foreign" (non-S) language, one can also add new functions to R
- ⌚ and they behave as if they were built-in to the language and behave in exactly the same way as the provided ones.
- ⌚ So R is very extensible by the user at different levels i.e. within R or via foreign code.

## R packages

- ⌚ Large collection of built-in functions
- ⌚ Arranged in packages, with each package providing a collection of somewhat related functions for a given type of computation or task.
- ⌚ R comes with 23 packages and 7 are loaded by default when R starts.
- ⌚ Other packages provide basic statistical functionality.
- ⌚ Different repositories or archives provide many other packages
- ⌚ CRAN, BioConductor, Omegahat, ....
- ⌚ Over 700 packages publically available.

# Graphics

stat141

- One of the things that made S successful early on was that it had excellent graphics facilities for statistics.
- It boasted a very flexible and powerful low-level mechanism for creating very rich displays.
- In addition, it provided functions for a large collection of commonly used statistical displays.
- And the plots were aesthetically pleasing, with sensible defaults (e.g. tick marks, labels, titles, colors and glyphs)

13

- Systems like Matlab caught up with R, even surpassed them in certain types of plots.
- Then trellis was developed by Cleveland, Becker and Shu (Bell Labs) for looking at panels of related plots, based on conditioning and subgroups.
- This is called lattice in R (developed by Deepayan Sarkar) and is available via the library(lattice) package.
- And this is made possible by a very advanced graphics system named 'grid' in R developed by primarily Paul Murrell.

stat141

14

# Basics of R

stat141

- Start R at shell prompt with R (Rgui on Windows or via icon)
- Often run inside a text editor.
- To terminate your R session, call the q() function. answer whether you want to save your session for future use.
- To get help on a function, help("functionName"), or use the help system - help.start() - which displays the pages in your Web browser.

15

# Assignments

stat141

- Store results from a command in "variables"
- `x = log(10)`
- x contains the value of the expression log(10).
- To use the value in a call, just use the name of the variable, e.g.  
    `x = 10`  
    `log(x)`

16

## Variable Names

stat141

- Variables can be arbitrary names, but
  - can't start with digit or `_`
  - Shouldn't have quotes, etc. in the name.
- So `xyz`, `x.y.z`, `.x`, `abc.lm`, `abc_lm`, `x1`, `x2` are okay.
- `1x`, `2x`, `_x`, not okay.
- Avoid names like `c`, `t`, `s`, `.C`, ...
- `#` is the character for a comment.  
All text after this on a line is ignored

17

## Assignments

stat141

- Older versions of R didn't allow `_` in a name.
- Reason: there were three different assignment operators
  - `=`
  - `<-`
  - `_`
- `x = 1; y <- 10; z _ 3`
- `_` is now deprecated.
- Good to know about when reading older scripts.
- Use `=` or `<-`

18

## R session

stat141

- When we start R for the first time, we have nothing in our work space.
- When we make assignments (at top-level), the variables are stored in our work space.
- `objects()` tell us what is there. `rm()` can remove them, etc.
- How do we maintain these across sessions?

19

## rda

stat141

- Use `save()` and `save.image()`
- `save.image()` stores the workspace and puts it into a file named `.RData`
- When R restarts in this directory, it will load the objects from that file if it is present.
- Can avoid this by starting R with  
`R --no-restore`

20

## save() & load()

stat141

- Save.image() saves all objects in the work space.
- May want to save just a few and into a different file name.
- `save(x, y, z, file = "myFile.rda")`
- Can explicitly load the contents of the .rda file into R using  
`load("myFile.rda")`
- Note that the .rda files are platform independent  
I.e. you can save on one OS, and use on another.

21

## Basic Calculator

stat141

- $1 + 2 - (3 * 4) / 8$  - addition, subtraction, ...
- $2^4$  - to the power of
- $10 \% 3$  - modulus
- $\log(10)$ ,  $\log(8, 2)$

22

## Language

stat141

- Essentially, R tries to treat everything as a function call
- `log(10)`, `sum(c(1,2,3))`, `plot(x)`, ....
- Breaks these up into a function call  
with inputs to the call  
and an output from the call
- Can combine calls with one call as an argument to another.

23

## Exceptions

stat141

- What about
  - $1+1$
  - $1:10$
  - $2^8$
- Each of these are actually converted into calls to the associated function.
  - `+(1, 1)`
  - `:(1, 10)`
  - `^(2, 8)`
- So there really is quite a simple structure

24

## Function Calls

- So the big thing in R is knowing how to call functions
- And R tries to make it convenient.
- The syntax is  
    `functionName(arg1, arg2, .....)`  
or  
    `functionName(arg1, name=value, name = value)`

25

## R Function Calls Features

- R allows for
  - Default values
  - “partial name matching”
  - Variable number of arguments, i.e. ... mechanism
  - Lazy evaluation

26

## Default Argument Values

- Function `log()` has 2 arguments  
    `log(x, base = exp(1))`
- This means that we can call  
    `log(3)`  
and it takes the value of base to be the result of `exp(1)`.
- `rnorm` has 3 arguments -  
    `rnorm(n, mean = 0, sd = 1)`
- The call `rnorm(10, sd = 5)` means  
    n is 10 and sd is 5 and the unspecified value of mean is taken to be 0.

27

## Default Values

- Not present in C, Fortran, Java.
- Convenient, because we leave out the ones we don't need to specify.
- R matches the arguments by order or by name. If we skip over an argument, we either use the name for the ones we are specifying, or leave a “blank” argument
- `rnorm(10, sd = 5)` or `rnorm(10, , 5)`

28

## Named Arguments

- We can always use names for arguments
  - `rnorm(10, sd = 5)`
  - `rnorm(10, , 5)`
  - `rnorm(10, mean = 0, sd = 5)`
  - `rnorm(n = 10, mean = 1, sd = 5)`
- Which are more readable? Which is easiest to type?
- Depends on whether we are doing interactive, one-time computations or writing software that we want to be able to read later or have others read.

29

## Lazy Evaluation

- One interesting aspect of R is “lazy evaluation”.
- This means that the actual expression for evaluating the value of an argument does not happen until the variable is needed.

30

## Objects and the Workspace

- `objects()`, `ls()` tell us about what variables we have.
- Where does R find functions like `sum()`, `plot()`, `objects()`
- Like the shell, it uses an “ordered path” in which it looks for *variables, including functions*.
- `search()`
- Shows a collection of libraries/packages that are available to the session.

31

## Search path

- `search()`

```
[1] ".GlobalEnv"      "package:methods" "package:stats"
[4] "package:graphics" "package:grDevices" "package:utils"
[7] "package:datasets" "Autoloads"         "package:base"
```
- Work space, S4 object mechanism, statistics functionality, graphics facilities, utilities, datasets, ..., built-in functions.
- To find where a variable will be found, use `find()`
- `find("lm")`

32



- Can load new packages into search path -  
library(lattice)
- Available libraries can be found with command library()
- More available from CRAN, Omegahat, BioConductor, ...  
via  
install.packages()

## library() & detach()

- search()  
[1] ".GlobalEnv" "package:methods" "package:stats"  
[4] "package:graphics" "package:grDevices" "package:utils"  
[7] "package:datasets" "Autoloads" "package:base"
- library(rpart)  
search()  
[1] ".GlobalEnv" "package:rpart" "package:methods"  
[4] "package:stats" "package:graphics" "package:grDevices"  
[7] "package:utils" "package:datasets" "Autoloads"  
[10] "package:base"
- detach('package:rpart')  
(or detach(2))
- Ps. Note that we can use " or ' as quotes,  
but have to use them in pairs.

## Everything is an Object

- Functions, data, etc. are all variables in R and we can treat them all (even functions) as inputs to functions.
- Every object has a class  
class(x), class(sum)  
This tells us what type it is.
- Most objects have a length.  
length(c(1, 2)).

## Graphics Model

- Devices - different types. Output to
  - a file - postscript(), pdf(), png(), bitmap(), ...
  - interactive screen - window(), x11(), gtk(), ...
- Can explicitly create a device  
or one will be created when a graphics function needs it.
- Plotting/graphics commands write their contents to the screen or to a file.
- For output to a file, need to explicitly close the device to ensure all content is written -  
dev.off()

## Active Graphics Device

- Key point - one and only one device is active at any time.
- All plotting commands work with the currently active device and draw to that device.
- Can have many devices open at any time, but only one is the active device.
- The variable `.Devices` gives a list of the open devices.
- Can switch between devices using `dev.set()`
- Can copy contents of one graphics device to another

37

## Example

- Create a histogram with a line for the mean and median.  

```
x = rnorm(1000)
hist(x)
abline(v = median(x))
abline(v = mean(x), lwd = 2, lty = 2, col = "red")
```
- Now, copy it to a PDF file.  

```
pdf("myHist.pdf") # create new device in a file.
dev.set(dev.prev()) # make previous device active
dev.copy() # copy current to next device
```

38

## Adding to Plots

- When we issue a graphics command, it typically clears the current display on the currently active device.
- Sometimes we want to add to an existing plot.
- Certain graphics functions such as `points()`, `lines()`, `text()`, `matlines()`, `legend()`, add content to an existing plot rather than start a new plot.

39

## Graphics Parameters

- Each graphics device has a LARGE set of "current" settings for things such as color, font, text size, margins, line width and style, ...
- We can explicitly set these using the `par()` function. They take effect in all subsequent graphics commands (for that device).
- We can override individual settings within a single call using the ... mechanism in most graphics functions.

40

## Graphics Parameters

stat141

- ```
par(bg = "gray")  
x = rnorm(100)  
plot(x)  
abline(h = c(-1.96, 1.96), col = "red", lwd = 2)  
title("Sample of 100 from N(0, 1)")  
idx = which(abs(x) > 1.96)  
points(idx, x[idx], pch = "+", col = "red", cex = 2)
```
- This shows setting the background (bg) "globally", col, lwd, pch, cex in individual calls.

41

## Multiple plots

stat141

- One can have multiple devices open, and this allows different plots to be compared on the screen in different devices.
- However, we can also have multiple plots on a single device using "screens" or "frames".
- We can partition the device into separate regions (overlapping or not) and draw on each of these in turn.
- Each new plot is drawn on the next frame.

42

## Creating Frames

stat141

- ```
par(mfrow = c(nr, nc))
```

  
creates a matrix of screens with nr rows, nc columns.
- `layout()` can create much more sophisticated and controlled partitions.

43

## Comparisons

stat141

- R has functionality for most modern statistical methods since it is written by and for statisticians, along with a large user community of contributors of add-on packages.
- Matlab is focused on engineers, and so lacks much of the functionality and convenience for statistics.
- Weaker programming language, but very fast.
- Also, doesn't have the user community of contributors nor a mechanism for creating & distributing add-on packages.

44

## SAS

stat141

- SAS is the most well-known statistics software.
- It deals with large datasets extremely well.
- Focus is on statistics, but increasingly less so and accordingly SAS does not provide all the most modern statistical methods.
- More of a basic language for doing “traditional” statistical analyses.
- The programming language is arcane/awkward.
- But a valuable thing to know and widely used along with S.

45

## Perl, Python

stat141

- Interpreted, fast scripting languages  
Python is often used interactively also.
- These are programming languages with a much more general audience and purpose than S or Matlab.
- As a result, they are more widely used, but have little built-in functionality for statistics.
- We often use Perl to preprocess data.
- Becoming more widely used in the Bioinformatics worlds  
BioPerl & BioPython.
- So too is R via BioConductor.

46

## C, C++, Java

stat141

- These are for software development rather than interactive scripting.
- Low-level compiled languages that require one to declare variables and their types and create entire programs before running commands.

47