Introduction to R

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Areas of interest: miRNA, genetic variation, gene expression data analyses, machine learning,

Collaboratory Website

http://collaboratory.lifesci.ucla.edu/



Workshop 3: Introduction to R

Day I

- Introduction to R
 - Installation, configuration, basic concepts, syntax and usage

Day 2

- Data Analysis
 - Visualization methods
 - Statistical methods

Day 3

- Scripting, batch mode
- Packages, Bioconductor





Workshop 3: Introduction to R

Day I

- What is R? Why use it?
- Getting up and running
 - Installation, configuration issues
 - overview of R interface
 - GUIs for working with R
- Your first R session
- Basic concepts
 - Starting and stopping an R session
 - finding help/documentation
 - Data structures
 - Getting data in and out of R
 - Basic analysis



Dropbox

- In the drop box folder are:
 - course powerpoint
 - sample scripts
 - exercises
- Please DO NOT make changes to these files because they will affect EVERYONE!
- ▶ INSTEAD, copy them to a directory on your computer
- Let me know if you need to be invited to the dropbox folder



What is R?

- A glorified calculator
- Statistical tables
- Visualization tool
- Programming language and environment



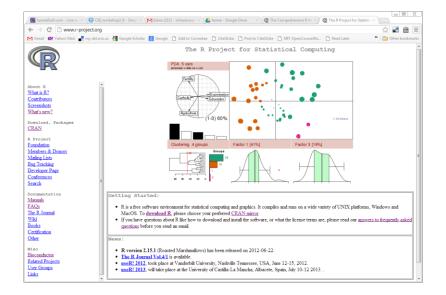
Why R?

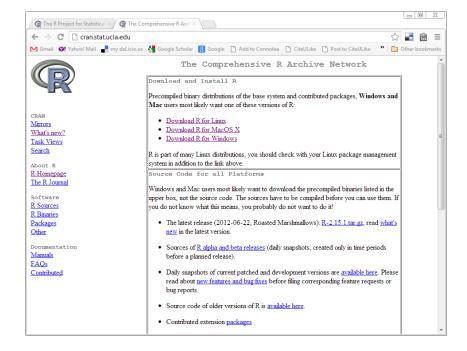
- ▶ Many tools STATA, SPSS, Matlab, Excel...
- Advantages
 - powerful
 - up to date with latest algorithms (packages, Bioconductor)
 - strong community of users
 - Preferred by statistics community
 - it's free
- Disadvantages
 - steep learning curve, but can be useful quickly
 - not pretty
 - can be memory intensive



Where to download R?

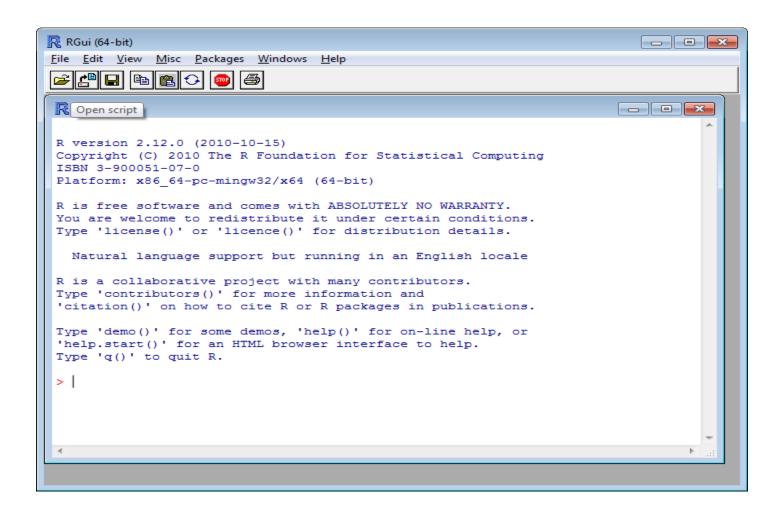
Use http://r-project.org or a CRAN mirror (http://cran.stat.ucla.edu)





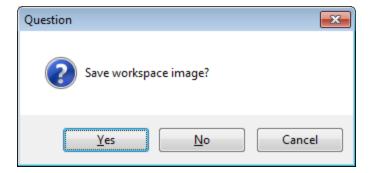


Starting Up...



Stopping

- To end your session
 - Close window or q()
- What does this mean?



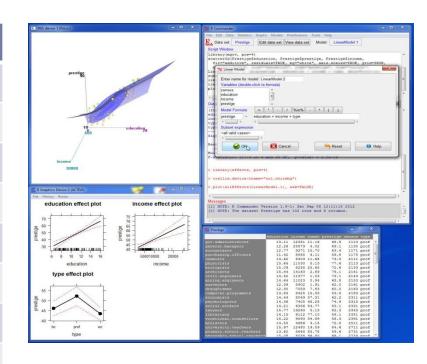
It saves everything you created (variables, etc) into a (hidden) file in the current directory usually named .Rdata

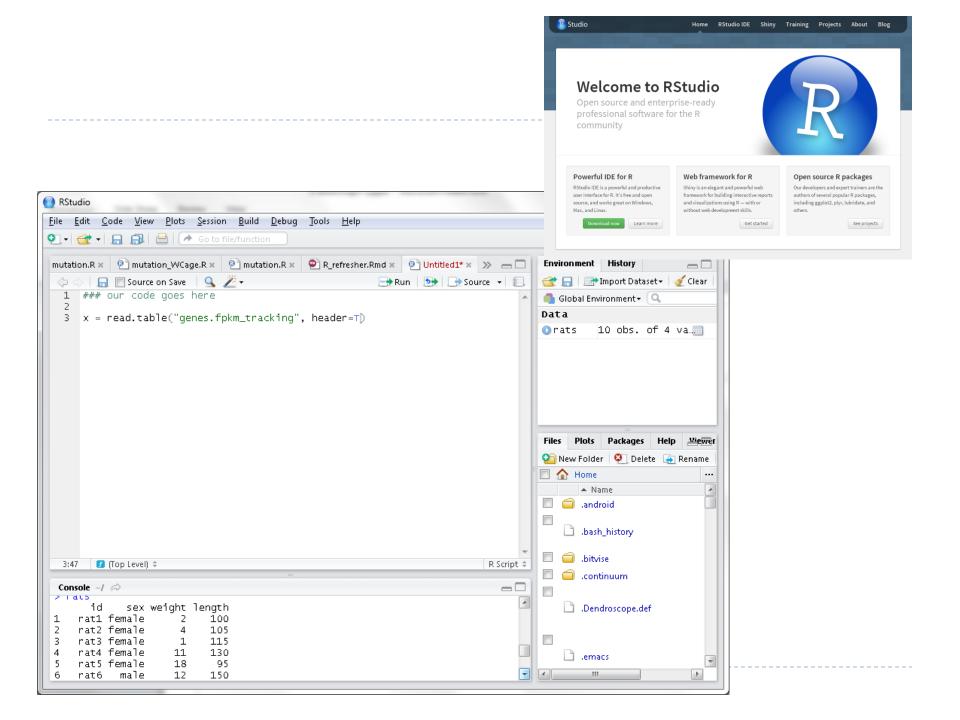


GUIs and IDEs

- ▶ R interface is pretty sparse
- Some alternatives:

	Win	Mac	Linux
RStudio Desktop	X	×	×
RStudio Server			×
Rcommander	X	X	×
Tinn-R	X		
JGR	X	X	×
RKWard	X		×
Emacs/ESS	X	X	×
Vi	X	X	×
notepad	X	X	×





It's like unix

- Many of the commands are similar to command line unix tools (cat, ls, grep, ...)
- but in R, everything is a function so () is important!
 - cat()
 - ► ls()
 - > rm()
- What happens if you forget the parentheses?
 - try it!



R packages

- List installed packages
- > installed.packages()
 > a<- installed.packages()</pre>
- Available packages
- > p<-available.packages()</pre>
- > dim(p)
- Is the package "boot" already installed?
- > is.element("boot", a[,1])
 >[1] TRUE

Finding help

```
> help("data.frame")
> ?data.frame
> help.search("chisquare")
> apropos("mean")
                       "kmeans"
 [1] "colMeans"
                                          "mean"
                                                            "mean.data.frame"
                       "mean.default"
 [5] "mean.Date"
                                          "mean.difftime"
                                                             "mean.POSIXct"
                       "rowMeans"
                                          "weighted.mean"
 [9] "mean.POSIXlt"
> library(help=utils)
> help(BATCH)
```



Advice for learning R

- ▶ There are no easy ways to learn R except by
 - Typing examples in yourself! It makes a difference (muscle memory, syntax)
 - 2. Persisting!
 - 3. Iterating!
- Steep learning curve, but with a few commands it can be useful quickly
- ▶ Try things out you will not break your data!



Getting around

R starts in a particular directory, usually the one where R is located

```
# print the current working directory
getwd()
# set the current working directory
setwd("/home/.../project1")
# show the files in your current directory
dir()
```

- Create a directory for your project
 - good way to organize

```
>dir.create("R CODE")
```



A few things

- comments start with a hash tag (#) and are ignored by R. Great way to make notes!
- ▶ The prompt ">" is where you type commands, statements

```
> a = 1
```

"+" is a continuation character; if your statement is not complete, R prompts you to finish it:

```
> a =
+ 3
> a
[1] 3
```

Typing the variable name gives you its contents

```
> a
[1] 3
```

Case matters! A is not the same as a

```
> A
```

```
Error: object 'A' not found
```

Typing a function name without parens () gives the source code

Try it out

```
# genes.fpkm tracking output of cufflinks
# it is tab delimited with a header line
# what happens if we say header=F?
x = read.table("genes.fpkm tracking", header=T)
ls()
# what do the numbers in dim() mean?
dim(x)
names(x)
head(x)
class(x)
sum(x[,11] > 100)
sum(x$q1 FPKM > 100)
# try changing the axis limits "ylim" and printing character "pch"
plot(x[,11], x[,14], xlim=c(0, 10), ylim=c(0,20), xlab="exp", main="my
experiment", pch=2)
# try chaning the number of breaks
hist(x[,11], col="red", breaks=100)
cor(x[,11], x[,14])
newx = x[x$status=="OK",]
ls()
```

Basic data structures in R

	Homogeneous	Heterogeneous
Id	Vector	List
2d	Matrix	Data frame
nd	Array	

Note: R has no 0-dimensional, or scalar types. Individual numbers or strings, which you might think would be scalars, are actually vectors of length one.



Indexing

let's create a vector

```
a = c(1,2,3,4)

a[2] \# gives 2, indexing in R start with 1 not 0!

a[1:3]
```

create a matrix

```
m = matrix(c(1,2,3,4,5,6), nrow=2, ncol=3)

m[,1]

[1] 1 2

m[,2] = 30
```

logical indexing

```
x = 1:10

x > 5

[1] FALSE FALSE FALSE FALSE TRUE TRUE TRUE TRUE TRUE

x[x > 5]

[1] 6 7 8 9 10
```

Vectors

Basically arrays

```
> v = 100
> v <- 100
> c(1,2,3)
> 1:3
> rep(1:2, 3)
> vector(mode="numeric", length=5)
```

Vector Arithmetic

```
> 1 + 1:3

[1] 2 3 4

> 1:2 + 1:3 #what happens?
```

Vector Arithmetic

Vector Arithmetic

```
> 1:2 + 1:3 #what happens? - ERROR!
> 1:2 + 1:10
[1] 2 4 4 6 6 8 8 10 10 12
```

Vector indexing (1-based)

```
> b = 1:2 + 1:10
> b[1]
```

Vector length

> length(b)

Name

You can name a vector in three ways:

```
> x <- c(a = 1, b = 2, c = 3)
> x <- 1:3; names(x) <- c("a", "b", "c")
> x <- setNames(1:3, c("a", "b", "c"))</pre>
```

Factors: Categorical data

- Categorical variables
- not the same as character class

```
> col = c("red", "green", "red", "yellow", "red")
> class(col)
[1] "character"
> col.f = factor(col)
[1] red green red yellow red
Levels: green red yellow
> levels(col.f) = c("green", "red", "blue")
> col.f
[1] red green red blue red
Levels: green red blue
> table(col.f)
col.f
green red blue
   1 3 1
```



Matrices

Grid of numbers. Must be numeric!

```
> x <- matrix(1:10, nrow=2)
> dim(x)
[1] 2 5
> x
[,1] [,2] [,3] [,4] [,5]
[1,] 1 3 5
[2,]
                                 10
> as.vector(x)
[1] 1 2 3 4 5 6 7 8
                                     9 10
> dim(x)
[1] 2 5
\# hint: x[2,4] means get element from row 2, col 4
```



List

Lists are different from atomic vectors because their elements can be of any type, including lists. You construct lists by using list() instead of c().

```
# define list x
>x <- list(1:3, "a", c(TRUE, FALSE, TRUE), c(2.3, 5.9))
# display structure of x
>str(x)
# vector from a list
> unlist(x, recursive = TRUE, use.names = TRUE)
```



Matrices and arrays

Adding a dim() attribute to an atomic vector allows it to behave like a multi-dimensional array.

```
# Two scalar arguments to specify rows and columns
>a <- matrix(1:6, ncol = 3, nrow = 2)
>rownames(a) <- c("A", "B")
>colnames(a) <- c("a", "b", "c")
# One vector argument to describe all dimensions
>b <- array(1:12, c(2, 3, 2))
> dimnames(b) <- list(c("one", "two"), c("a", "b", "c"), c("A", "B"))</pre>
```



Data frames

- the most useful data structure
- data frames hold data like a spreadsheet
 - observations are rows
 - covariates are columns
- like matrices, they can be subsetted and indexed with two subscripts



Data frames

```
> df = data.frame(type=rep(c("case", "control"), c(2, 3)),
time=rexp(5)
> df
           time
     type
1
    case 1.1745712
    case 1.1691266
3
  control 0.8227643
 control 0.1301390
 control 1.0581316
  df$time
[1] 1.1745712 1.1691266 0.8227643
[4] 0.1301390 1.0581316
> df["time"]
```



Data frames

```
# let's update row names
> names(df)
[1] "type" "time"
> rn <- paste("id", 1:5, sep="")</pre>
> rownames(df) <- rn</pre>
> df[1:2, ]
    type time
id1 case 1.174571
id2 case 1.169127
# update column names
> names(df) = c("var1", "var2")
> df[1:2,]
   var1 var2
id1 case 0.1653743
id2 case 0.4509330
```



Data type conversions

- Many problems are due to unexpected data types
 - eg. using a factor in place of a character string
- to convert from one data type to another, try

```
as.vector()
as.character()
as.data.frame()
as.numeric()
as.matrix()
```

to determine what data type your variable is, try

```
class(variable)
```



Functions

you can define your own functions

R functions take can take named and default parameters

```
say = function(word="hi", repeats=3) {
  cat( rep(word, repeats), "\n")
}
say()
say(word="hola", repeats=1)
say(repeats=2)
```



Reading in data

A very common operation

```
> x = read.table("myfile.txt")
```

- Several variants: read.csv(), read.delim()
- this reads a delimited file into a data.frame object
- every line of the file must have the same number of elements!
 - sometimes you can fix things eg read.table("myfile.txt", sep="\t")
- entire file read into memory, so this can be slow but optimizations possible



Reading the help

```
read.table(file, header = FALSE, sep = "", quote = "\"'",
           dec = ".", row.names, col.names,
           as.is = !stringsAsFactors,
           na.strings = "NA", colClasses = NA, nrows = -1,
           skip = 0, check.names = TRUE, fill = !blank.lines.skip,
           strip.white = FALSE, blank.lines.skip = TRUE,
           comment.char = "#",
           allowEscapes = FALSE, flush = FALSE,
           stringsAsFactors = default.stringsAsFactors(),
           fileEncoding = "", encoding = "unknown")
read.csv(file, header = TRUE, sep = ",", quote="\"", dec=".",
         fill = TRUE, comment.char="", ...)
read.csv2(file, header = TRUE, sep = ";", quote="\"", dec=",",
          fill = TRUE, comment.char="", ...)
read.delim(file, header = TRUE, sep = "\t", quote="\"", dec=".",
           fill = TRUE, comment.char="", ...)
read.delim2(file, header = TRUE, sep = "\t", quote="\"", dec=",",
            fill = TRUE, comment.char="", ...)
```

Writing data

- Writing data to a file is pretty simple
- Some odd defaults need to be commonly overridden

```
write.table(x, file="file_name.txt", quote=F, sep="\t",
row.names=F, col.names=T)
```

Other variants write.csv()

Saving and loading session

- all objects in your workspace can be saved to a file that can be read again on the next startup
- save() or save.image()
- the file is usually called .RData in your current directory
- ▶ load() will load an .RData file
- GUI might have these options



Some useful commands

- show a recent history of commands you typed
 - history()
- customizing your R
 - Rprofile is a file found in your home directory where you can set defaults, define some commonly used functions, etc
- rm() remove an object from workspace
- ▶ Is() list the objects in your workspace
- cbind(array I, array2) combine columns
- rbind(array1, array2) combine rows
- ▶ t() transpose a vector, matrix or data frame
- sessionInfo() info about your settings



Match

match returns a vector of the positions of (first) matches of its first argument in its second.



Merge

Merge two data frames that have a common identifier

```
df1 = read.table("df1.txt", header=T)
df2 = read.table("df2.txt", header=T)
both = merge(df1, df2, by.x=1, by.y=1)
new = merge(df1, df2, by.x=1, by.y=1, all.x=T)
```



which

which helps you find the index of an element

```
> a = 1:10
> which(a < 5)
[1] 1 2 3 4
> a[which(a < 5)]
[1] 1 2 3 4
> ids = which(a < 5)
[1] 1 2 3 4
> a[ids]
```

Special values

- ► NA (Not available)
 - sometimes data shows up as 'NA'
 - usually due to
 - missing data
 - conversion
 - **not** the same as zero (0)!
 - test with is.na(x)
- NaN (not a number)
 - **)** 0/0
 - is.nan()
- Inf (infinite) or -Inf
 - 10/0
 - test with is.finite(x) or is.infinite(x)

NA/NAN/INF

```
#Test if any NaN
 sum(is.nan(df1[,1]))
#Test if any INF
 sum(is.finite(df1$id))
#Test if any NA's
 sum(is.na(df2))
#Find the col/row with the NA
res = is.na(df2)
> colSums(res)
    id age weight
> rowSums(res)
[1] 0 0 0 1 0 0
```



Logical operators

- Useful for chaining operations
 - ▶ | (OR), & (AND), ! (NOT)
 - ▶ for instance, you want values greater than 10 or less than 5

```
df[ df[,1] > 10 | df[,1] < 5]
df[ !is.na(df[,1])]
df[ df[,2] == "gene" & df[,1] > 10]
```

== is the equal operator (not =, which assigns)



Flow control

If statements for branching

```
if (x == 3) {
    doSomething()
} else {
    doSomethingElse()
}
```

loops

▶ slow in R

```
a = seq(1,10)
for(i in 1:length(a)){
cat(a[i], "\n")
}
```

Apply

- R operations are usually vectorized, so loops are avoided.
- Loops are quite slow in R

```
mat = matrix(seq(1,10), 2)
# how to get rowwise or columnwise mean?
# could use a loop, but you have to specify dimension
# margin 1 = row, 2 = cols
apply(mat, 1, mean)
apply(mat, 2, mean)
```



Sorting

Sorting returns an array in sorted order

```
> b = c(1,9,8,2,4,6)
> sort(b, decreasing=F)
```

- Ordering returns a vector of array indicies that puts the vector in a particular order
 - more useful for sorting data frames

```
> b = c(1,9,8,2,4,6)
b.order = order(b, decreasing=F)
> b.order
[1] 1 4 5 6 3 2
> b[b.order]
[1] 1 2 4 6 8 9
```



Reading in a script

- Suppose you've written a R script that processes your data
- source() will read it in as if you typed it and execute all commands listed in the file
- source("R_script_file.R")



Additional resources

Many resources but sometimes difficult to access

- R-bloggers.com
- R-project.org FAQ and manuals
- bioconductor.org
- UCLA ATS (http://www.ats.ucla.edu/stat/r/)
- Quick R http://www.statmethods.net/index.html
- http://onertipaday.blogspot.com/
- http://wiki.stdout.org/rcookbook/
- http://gettinggeneticsdone.blogspot.com/
- manuals.bioinformatics.ucr.edu/workshops

Help sites

- stackoverflow.com
- nabble.com
- google

Good free books

- An Introduction to R (on r-project.org)
- Using R



Additional resources

Courses

- Coursera
 - https://class.coursera.org/rprog-002
- EdX.org
 - https://www.edx.org/course/harvardx/harvardx-ph525x-data-analysis-genomics-1401#.U0xFnvldW_g
- Stanford Class
 - https://class.stanford.edu/courses/HumanitiesScience/StatLearning/Winter2014/about
- Google YouTube
 - https://www.youtube.com/watch?v=iffR3fWv4xw&list=PLOU2XLYxm slK9qQfztXeybpHvru-TrqAP
- Bioconductor
 - http://www.bioconductor.org/help/events/



Shortcuts

To interrupt

- MAC: <CMD> <Option> <.> or <ESC>
- PC/Linux: Ctrl-C

sessionInfo()

this displays details about your environment, what packages are loaded, etc... useful for debugging



Workshop 3: Introduction to R

Day 2

- Statistical methods
 - Probability distributions
 - Random number generation
 - Common tests
 - ☐ Fisher's exact test, chi square, qvalue, etc...
- Visualization
 - Plotting functions
 - Customizing plots
 - Saving plots for publication



Workshop 3: Introduction to R

Day 3

- Automating or scripting R
 - Batch mode
- Packages that extend R
 - Installation
 - Documentation
 - Useful packages
- Bioconductor
 - Extending R for the life sciences
 - Useful bioconductor packages
 - Some basic bioC concepts
 - Bioconductor demo

