



UNIVERSITÀ
DI TRENTO

Introduction to the R language

Biological Networks – Mod 1
Academic year 2021/2022
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TOPICS

What is R

How to obtain and install R

How to read and export data

How to do basic statistical analyses

Data analysis packages in R

WHAT IS R

- Software for Statistical Data Analysis
- Based on S
- Programming Environment
- Interpreted Language
- Data Storage, Analysis, Graphing
- Free and Open Source Software

STRENGTHS AND WEAKNESSES

- Strengths
 - Free and Open Source
 - Strong User Community
 - Highly extensible, flexible
 - Implementation of high end statistical methods
 - Flexible graphics and intelligent defaults
- Weakness
 - Steep learning curve
 - Slow for large datasets

OBTAINING R

- Current Version: R-4.1.2
- Comprehensive R Archive Network:
<http://cran.r-project.org>
- Binary and source codes
- Windows executables
- Compiled RPMs for Linux
- Can be obtained on a CD

INSTALLING R

- Binary (Windows/Linux): One step process
 - exe, rpm (Red Hat/Mandrake), apt-get (Debian)
- Linux, from sources:

```
$ tar -zxvf "filename.tar.gz"
$ cd filename
$ ./configure
$ make
$ make check
$ make install
```

STARTING R



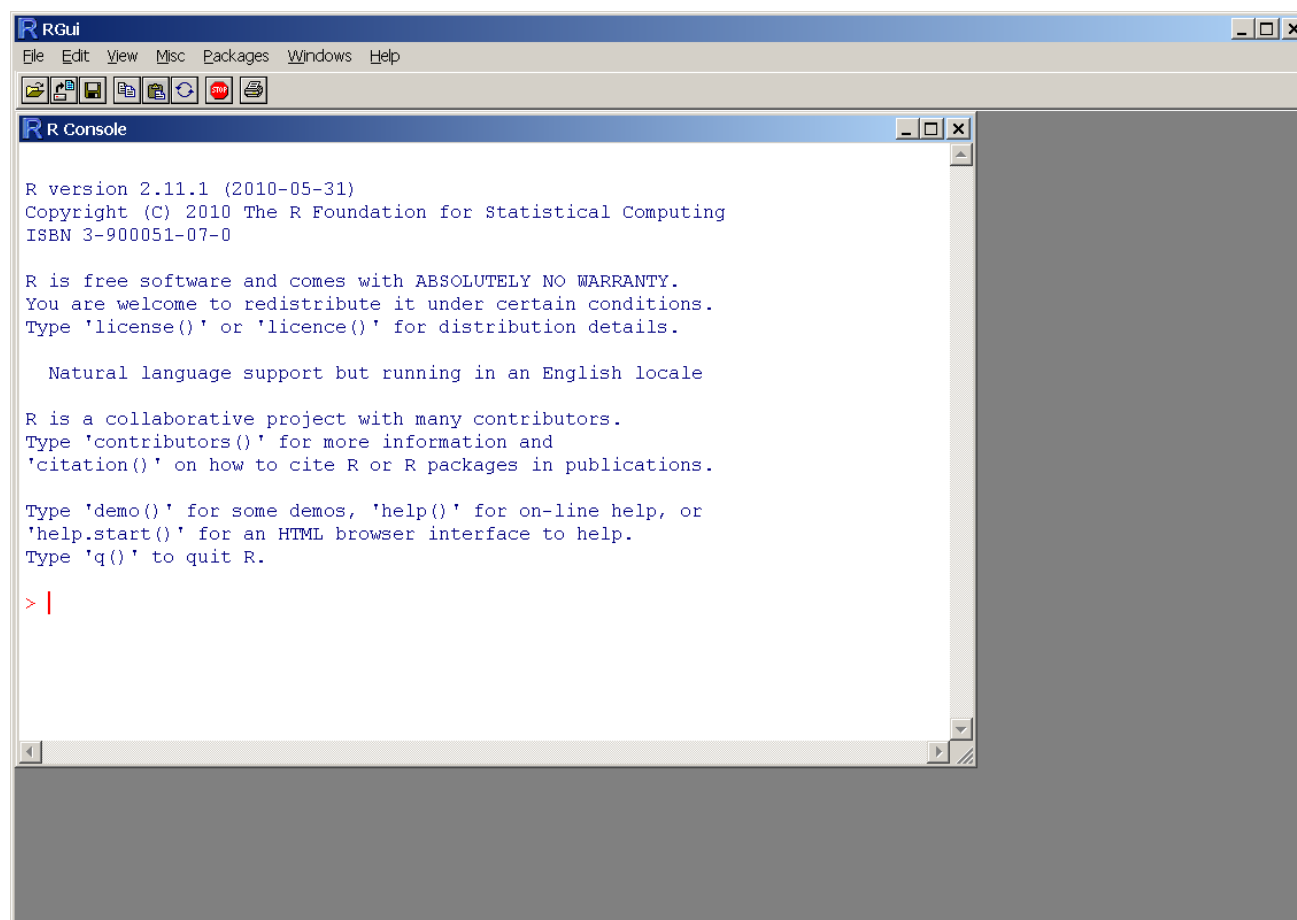
Windows, Double-click on Desktop Icon



Linux, type R at command prompt

RUNNING AND INTERACTING WITH R

Reference implementation of R includes a minimal programming environment – a console and no editor



RUNNING R

- An R program is created using an editor
 - Notepad++, WinEdt, Tinn-R: Windows
 - Xemacs, ESS (Emacs speaks Statistics)
- To run a program in the console:
 - `source("filename.R")`
 - Outputs can be diverted by using `sink("filename.Rout")`

RUNNING R

- What makes R so useful is the ease of installation of additional libraries
 - Libraries are called 'packages' in R parlance
 - Packages add functionality to the basic R installation
 - Hundreds packages are available in open online repositories
- Several IDEs are available for R to facilitate code development
 - a very popular one is Rstudio - available at <http://www.rstudio.com>

```
#
# basic code example
#

# install the XLConnect package
install.packages("XLConnect")
library("XLConnect")

# see what libraries are installed
library()

# change working directory
setwd("C:/Users/Mario/Documents")
getwd()

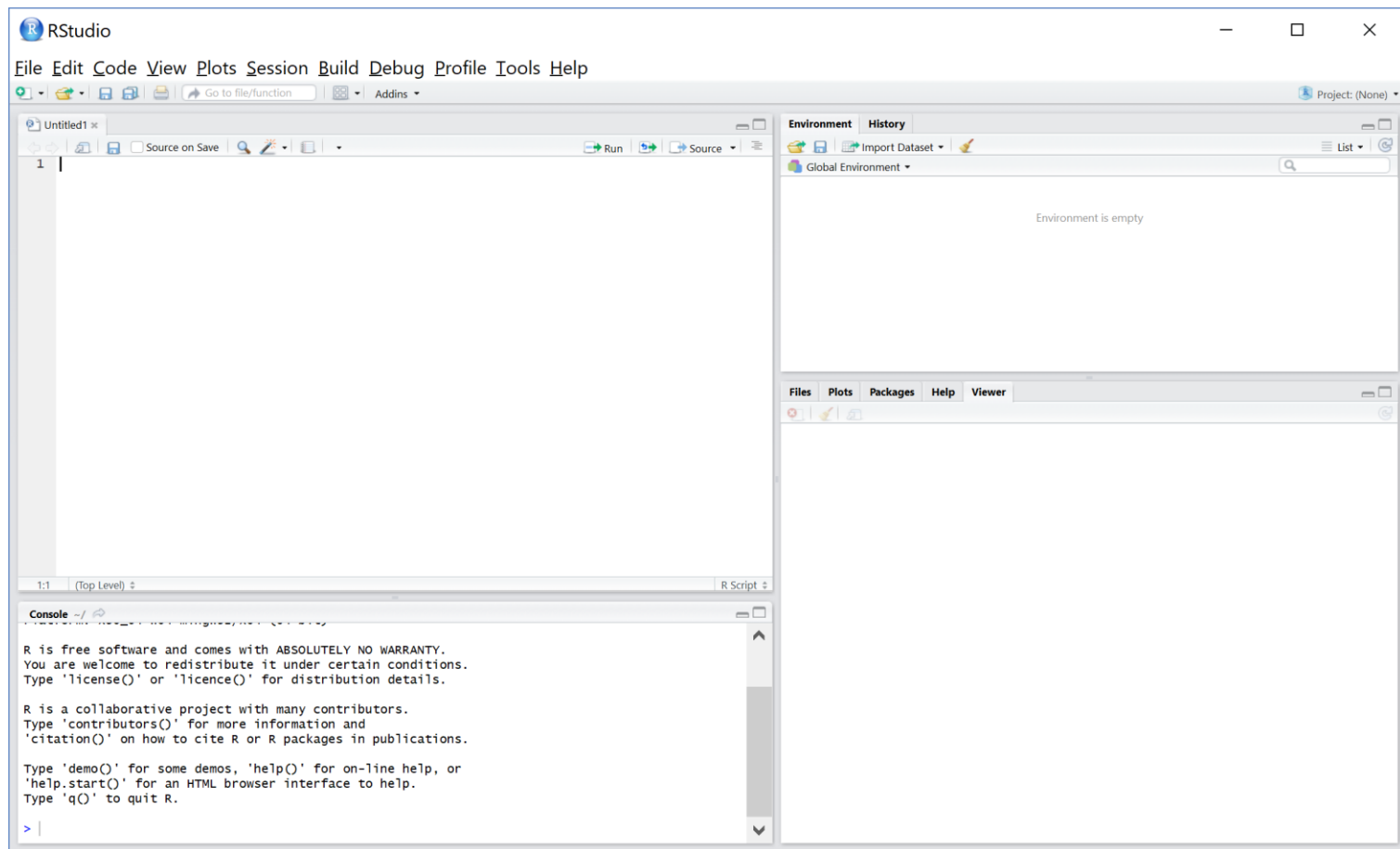
# load internal toy data set
data(iris)

# let's inspect the iris data set
?iris           # same as: help(iris)
class(iris)
dim(iris)
head(iris)
tail(iris)

q()
```

RUNNING R WITH A IDE

- IDE: Integrated Development Environment
- RStudio: an open source IDE for R



GETTING HELP IN R

- From built-in documentation:
 - `?WhatIWantToKnow`
 - `help("WhatIWantToKnow")`
 - `help.search("WhatIWantToKnow")`
 - `help.start()`
 - `getAnywhere("WhatIWantToKnow")`
 - `example("WhatIWantToKnow")`
- Reference manual: "An Introduction to R"
 - <https://cran.r-project.org/doc/manuals/r-release/R-intro.pdf>
- Many active mailing lists and blogs
 - typically found by using the right keywords in Google

DATA STRUCTURES

- Supports virtually any type of data
- Numbers, characters, logicals (TRUE/ FALSE)
- Arrays of virtually unlimited sizes and number of dimensions
- Simplest data structures: Vectors and Matrices
- Lists: Can Contain mixed type variables
- Data Frame: Rectangular Data Set

SYNOPSIS OF OPERATORS

Operator	Usually means	In Formula means
+ or -	add or subtract	add or remove terms
*	multiplication	main effect and interactions
/	division	main effect and nesting
:	sequence	interaction only
^	exponentiation	limiting interaction depths
%in%	no specific	nesting only

EXPRESSIONS IN R

Math:

```
> 1 + 1
```

```
[1] 2
```

```
> 1 + 1 * 7
```

```
[1] 8
```

```
> (1 + 1) * 7
```

```
[1] 14
```

Variables:

```
> x <- 1
```

```
> x
```

```
[1] 1
```

```
> y = 2
```

```
> y
```

```
[1] 2
```

```
> 3 -> z
```

```
> z
```

```
[1] 3
```

```
> (x + y) * z
```

```
[1] 9
```

EXPRESSIONS IN R

Arrays:

```
> x <- c(0,1,2,3,4)
```

```
> x
```

```
[1] 0 1 2 3 4
```

```
> y <- 1:5
```

```
> y
```

```
[1] 1 2 3 4 5
```

```
> z <- 1:50
```

```
> z
```

```
[1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
```

```
[16] 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30
```

```
[31] 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45
```

```
[46] 46 47 48 49 50
```


EXPRESSIONS IN R

Math on arrays:

```
> x <- c(0,1,2,3,4)
> y <- 1:5
> z <- 1:50
> x + y
[1] 1 3 5 7 9
> x * y
[1] 0 2 6 12 20
> x * z
[1] 0 2 6 12 20 0 7 16 27 40 0
[12] 12 26 42 60 0 17 36 57 80 0 22
[23] 46 72 100 0 27 56 87 120 0 32 66
[34] 102 140 0 37 76 117 160 0 42 86 132
[45] 180 0 47 96 147 200
```

DATA STRUCTURES IN R

	Linear	Rectangular
All Same Type	VECTORS	MATRIX
Mixed	LIST	DATA FRAME

VECTORS AND MATRICES

```
# The c() function can be used to create vectors
> x <- c(0,1,2,3,4)
> x
[1] 0 1 2 3 4

# Use the vector() function to create empty vectors
> y <- vector("numeric", length = 10)
> y
[1] 0 0 0 0 0 0 0 0 0 0

> z <- 1:25
> z
 [1]  1  2  3  4  5  6  7  8  9 10 11 12 13 14 15
[16] 16 17 18 19 20 21 22 23 24 25

# There is a matrix() function too
> m <- matrix(0, 2,3)
> m
      [,1] [,2] [,3]
[1,]    0    0    0
[2,]    0    0    0
>
```

ACCESSING MATRIX AND VECTOR ELEMENTS

- Subscripts are an essential tool
 - `x[1]` identifies first element in vector `x`
 - `x[-1]` identifies all elements in vector `x` except the first
 - `x[2:4]` identifies elements from the 2nd to the 4th in vector `x`
 - `x[c(2,4,5)]` identifies specific elements in vector `x`
 - `x[x > 3]` identifies all the elements greater than 3 in vector `x`
 - `y[2,3]` identifies the element in row 2, column 3 in matrix `y`
 - `y[1,]` identifies first row in matrix `y`
 - `y[,1]` identifies first column in matrix `y`
 - `y[1:5,]` gives first 5 rows of data
- To inspect a data object:
 - `edit(<mydataobject>)`

THE INFAMOUS NA

- R uses **NA** to represent data that is "not available"
- The function **is.na()** tests for **NA**
- **NA**'s are often an annoyance, see here an example of how to deal with them

```
> x <- c(1,2,NA,4,NA,5)
> x
[1] 1 2 NA 4 NA 5

> bad <- is.na(x)
> bad
[1] FALSE FALSE TRUE FALSE TRUE FALSE

> !bad
[1] TRUE TRUE FALSE TRUE FALSE TRUE

> x[!bad]
[1] 1 2 4 5
```

DATA FRAMES

- Data frames are R's fundamental data structures
- They are used to store data in a table format
- Unlike matrices, data frames can store different type of data
- Data frames are usually created by calling the `read.table()` or `read.csv()` functions
- They can be converted to a matrix by calling `data.matrix()`

BUILT-IN DATA SETS

- You can practice the use of data frames by importing some of the built-in data sets available in R
- `data()` lists the built-in data sets
- `Iris` is a nice one



Iris virginica



Iris setosa

IRIS DATA SET

RGui (64-bit)

File Windows Edit Help

R Console

```
R version 3.2.2 (2015-12-18)
Copyright (C) 2015 The R Foundation for Statistical Computing
Platform: x86_64-w64-mingw32/x64

R is free software and its use is unrestricted. Redistribution and
modification in source and binary forms are permitted (subject to the
terms of the GNU General Public License, which is included in the
package documentation). You are welcome to read the license at
<http://www.gnu.org/licenses/gpl.html>.
Type 'license()' or 'licence()' for distribution details.

Natural language support but running in an English locale

R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

> data(iris)
> ls()
[1] "iris"
> edit(iris)
```

Data Editor

	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species	var6
1	5.1	3.5	1.4	0.2	setosa	
2	4.9	3	1.4	0.2	setosa	
3	4.7	3.2	1.3	0.2	setosa	
4	4.6	3.1	1.5	0.2	setosa	
5	5	3.6	1.4	0.2	setosa	
6	5.4	3.9	1.7	0.4	setosa	
7	4.6	3.4	1.4	0.3	setosa	
8	5	3.4	1.5	0.2	setosa	
9	4.4	2.9	1.4	0.2	setosa	
10	4.9	3.1	1.5	0.1	setosa	
11	5.4	3.7	1.5	0.2	setosa	
12	4.8	3.4	1.6	0.2	setosa	
13	4.8	3	1.4	0.1	setosa	
14	4.3	3	1.1	0.1	setosa	
15	5.8	4	1.2	0.2	setosa	
16	5.7	4.4	1.5	0.4	setosa	
17	5.4	3.9	1.3	0.4	setosa	
18	5.1	3.5	1.4	0.3	setosa	
19	5.7	3.8	1.7	0.3	setosa	

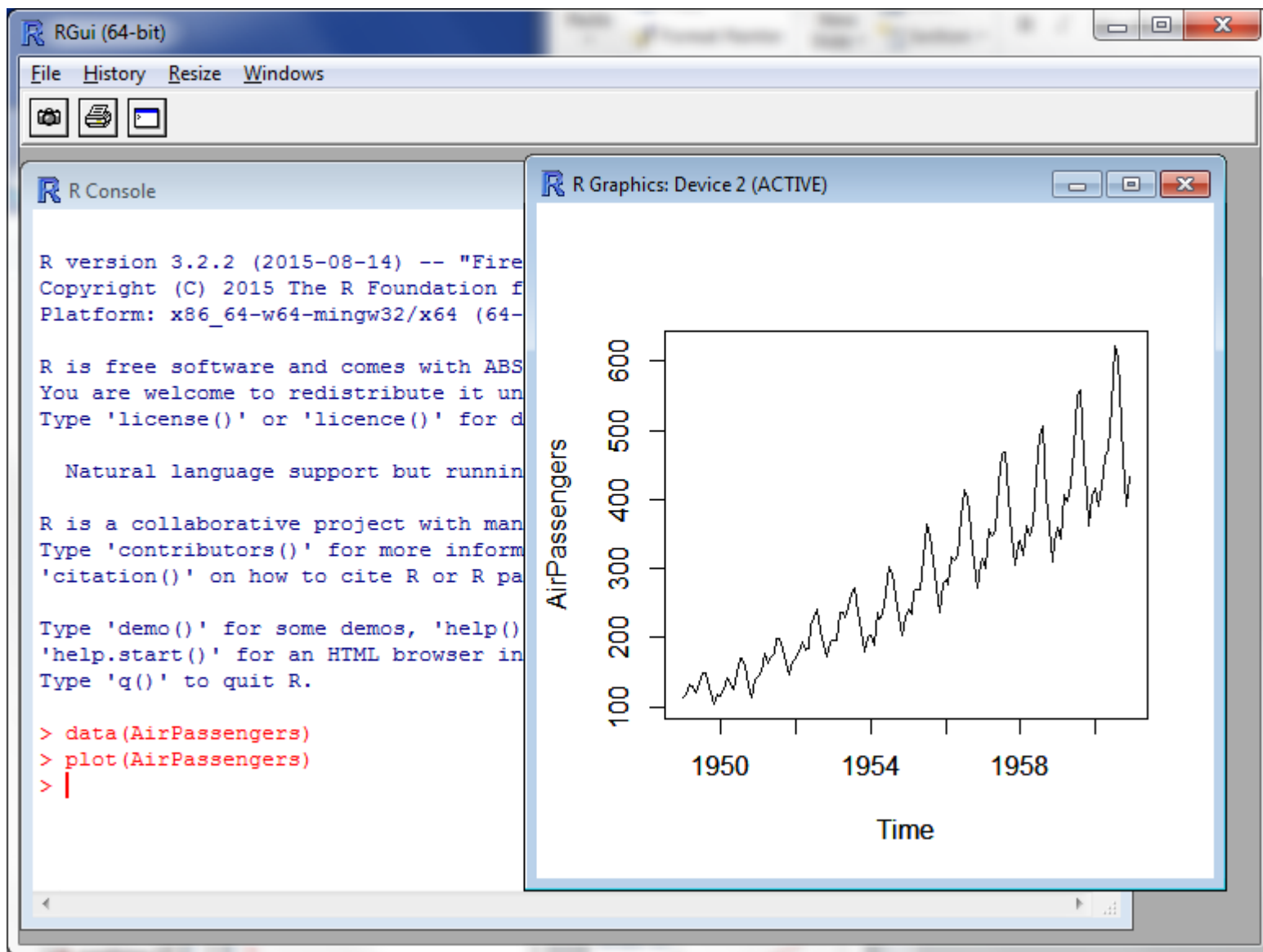
SUBSETTING DATA

- Using subset function
 - `subset()` will subset the dataframe
- Subscripting from data frames
 - `myframe[,1]` gives first column of myframe
- Using logical expressions
 - `myframe[myframe[,1] < 5,]` gets all rows that contain a value less than 5 in the first column
- Use the \$ sign for lists and data frames
 - `myframe$age` gets age variable of myframe
 - `attach(dataframe)` -> make visible in the R environment the row/column names of the dataframe

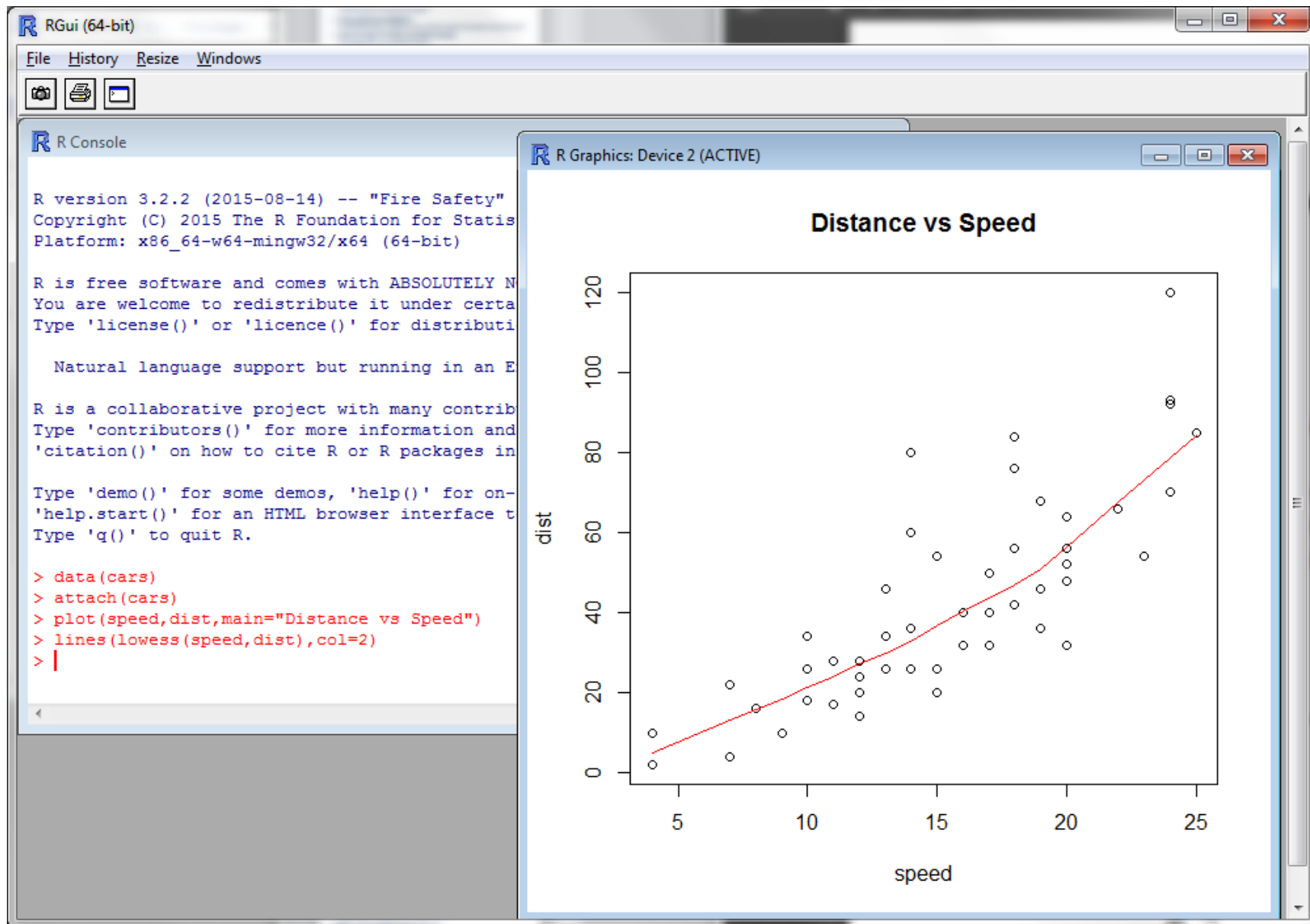
GRAPHICS

- Plot an object, like: `plot(num.vec)`
 - here plots against index numbers
- Plot sends to graphic devices
 - can specify which graphic device you want
 - postscript, gif, jpeg, etc...
- Two types of plotting commands
 - high level: graphs drawn with one call (R does all the work, good results most of the time)
 - Low Level: add additional information to existing graph (i.e. a graph title, a smoothing curve, etc.) to improve quality of the graph

HIGH LEVEL: GENERATED WITH PLOT()



LOW LEVEL: SCATTERGRAM WITH LOWESS



READING DATA INTO R

- R not well suited for data preprocessing
- Preprocess data elsewhere (SPSS, etc...)
- Easiest form of data to input: text file
- Spreadsheet like data:
 - Small/medium size: use `read.table()`
 - Large data: use `scan()`
- Use specialized packages to read from other systems:
 - Use the library “foreign” to read files from other statistical tools: `library(“foreign”)`
 - Can import from SAS, SPSS, Epi Info, can export to STATA format
 - Use the library “XLConnect” to read Excel files: `library(“XLConnect”)`

MORE BASIC R

```
#
# code example showing how to read/write data
#

# load internal toy data set
data(iris)

# create own copy
my.copy.of.iris <- iris

# how to extract columns from a data.frame
my.iris <-
data.frame(SepLen=iris$Sepal.Length,
            SepWid=iris$Sepal.Width,
            Species=iris$Species)

# quicker ways of extracting/removing cols
my.iris <- iris[,c(1,2,5)]
my.iris <- iris[,-c(3,4)]

# how to save/load a data set (R format)
save(my.copy.of.iris, file="mydata.RData")
mydata <- load(file="mydata.RData")

# how to save/load a data set (TAB sep. format)
write.table(iris, "myfile.dat", quote = FALSE,
            sep = "\t")
my.data <- read.table("myfile.dat", header = TRUE,
                    sep = "\t", row.names = 1)

# how to save/load a data set (CSV format)
write.csv(iris, "myfile.dat")
my.data <- read.csv("myfile.dat", row.names = 1)

# sometimes data is available in Excel format
# many packages can be used for this purpose
# install the XLCoonect package
install.packages("XLConnect")
library("XLConnect")

wb1 <- loadWorkbook("TP-Function.xlsx")
f1 <- readWorksheet(wb1, sheet="F1_48")

q()
```

R PROGRAMMING STYLE

- Functions & Operators typically work on entire vectors
 - You are encouraged to write your own functions and use them
- Codes separated by newlines
 - “;” not necessary, sometimes useful to improve code clarity
- You are highly encouraged to use a lot of explicative comments to make your code easier to maintain!
 - Anything following a “#” is a comment and ignored by R

STATISTICAL FUNCTIONS IN R

- Descriptive Statistics
- Statistical Modeling
 - Regressions: Linear and Logistic
 - Probit, Tobit Models
 - Time Series
- Multivariate Functions
- A large and diverse mix of built-in packages and contributed packages

DESCRIPTIVE STATISTICS

- R has functions for all common statistics
- `summary()` gives lowest, mean, median, first, third quartiles, highest for numeric variables
- `stem()` gives stem-leaf plots
- `table()` gives tabulation of categorical variables

STATISTICAL MODELING

- Over 400 functions
 - lm, glm, aov, ts
- Numerous libraries & packages
 - survival, coxph, tree (recursive trees), nls, ...
- Distinction between factors and regressors
 - factors: categorical, regressors: continuous
 - you must specify factors unless they are obvious to R
 - dummy variables for factors created automatically
- Use of data.frame makes life easy

HOW TO MODEL

Specify your model like this:

- $y \sim x_i + c_i$, where
- y = outcome variable, x_i = main explanatory variables, c_i = covariates, + = add terms
- Operators have special meanings
+ = add terms, : = interactions, / = nesting, so on...

Modeling -- object oriented

- each modeling procedure produces objects
- classes and functions for each object

MODELING EXAMPLE: REGRESSION

```
carReg <- lm(speed~dist, data=cars)
# carReg becomes an object
# to get summary of this regression, we type
summary(carReg)

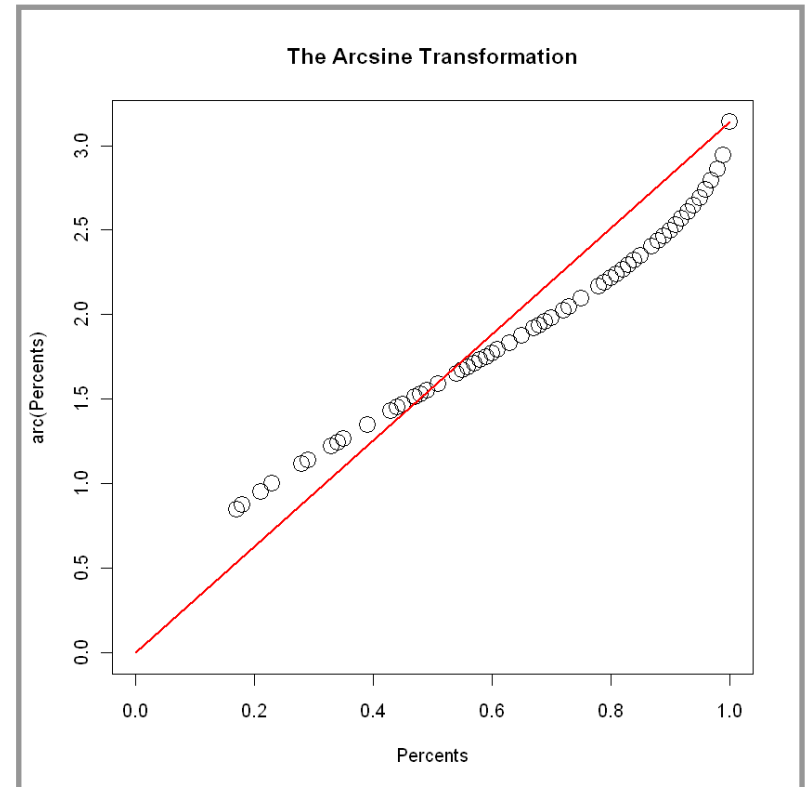
# to get only coefficients, we type
coef(carReg), or carReg$coef
# don't want intercept? add 0, so
carReg <- lm(speed~0+dist, data=cars)
```

INSTALLING, RUNNING, AND INTERACTING WITH R

Functions:

```
> arc <- function(x) 2*asin(sqrt(x))
> arc(0.5)
[1] 1.570796
> x <- c(0,1,2,3,4)
> x <- x / 10
> arc(x)
[1] 0.0000000 0.6435011 0.9272952
[4] 1.1592795 1.3694384
```

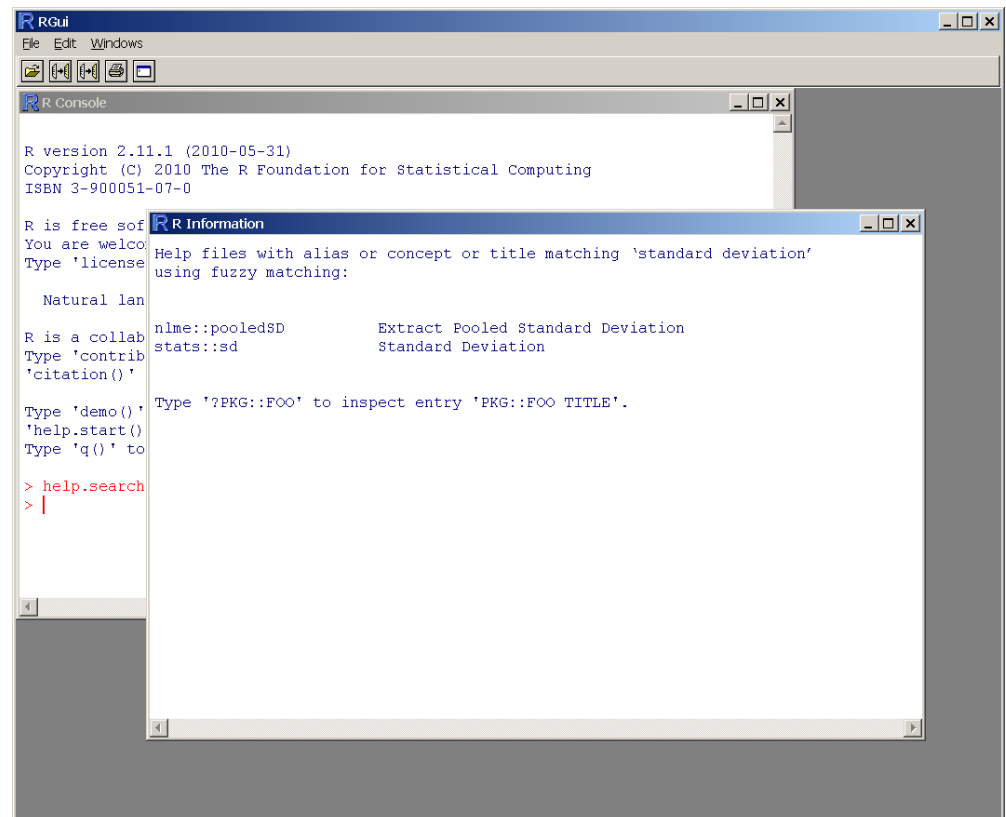
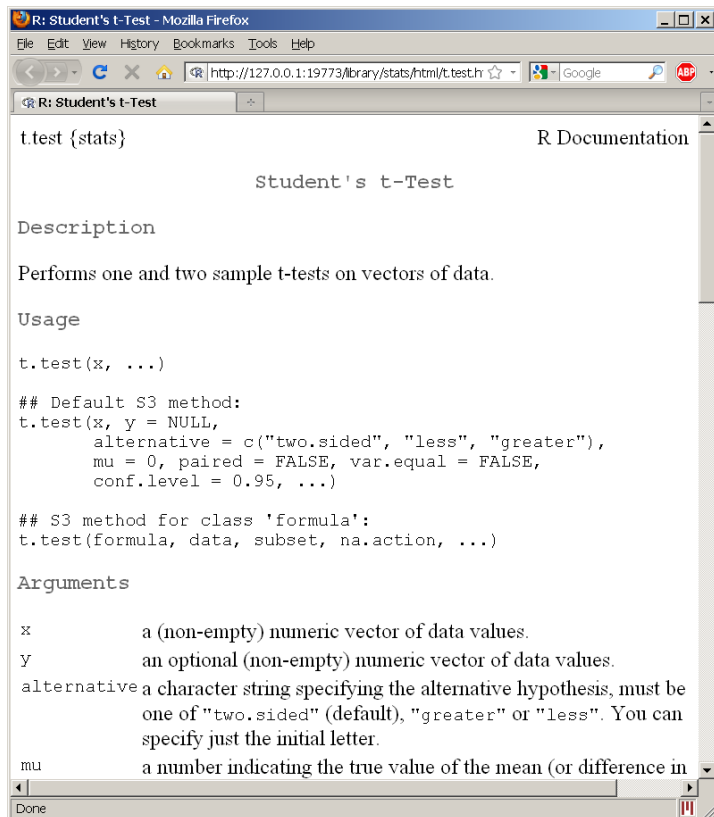
```
> plot(arc(Percents)~Percents,
+ pch=21,cex=2,xlim=c(0,1),ylim=c(0,pi),
+ main="The Arcsine Transformation")
> lines(c(0,1),c(0,pi),col="red",lwd=2)
```



INSTALLING, RUNNING, AND INTERACTING WITH R

Getting help:

```
> help(t.test)
> help.search("standard deviation")
```

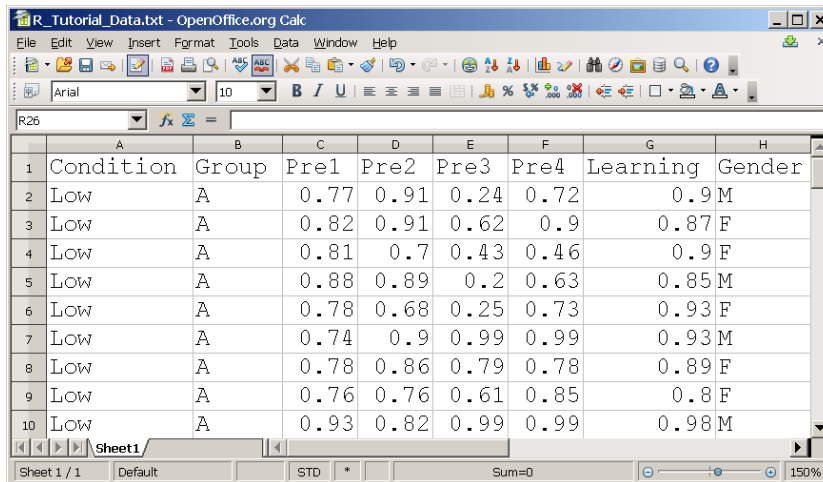


INSTALLING, RUNNING, AND INTERACTING WITH R

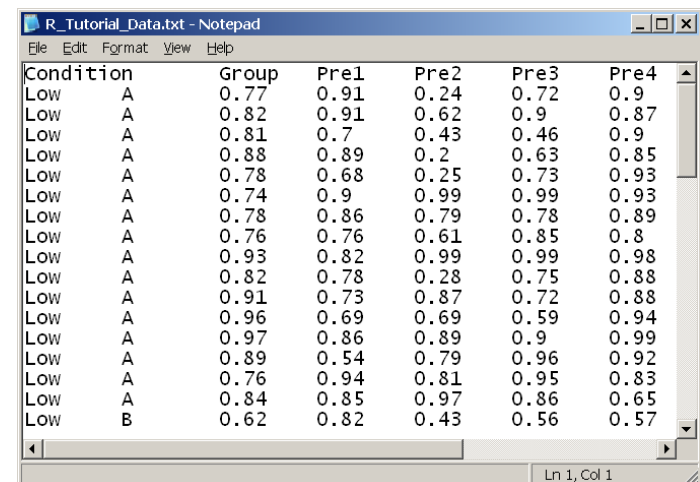
Reading data from files:

```
> myData <- read.table("R_Tutorial_Data.txt",  
+ header=TRUE, sep="\t")  
> myData
```

	Condition	Group	Pre1	Pre2	Pre3	Pre4	Learning
1	Low	A	0.77	0.91	0.24	0.72	0.90
2	Low	A	0.82	0.91	0.62	0.90	0.87
3	Low	A	0.81	0.70	0.43	0.46	0.90
...							
61	High	B	0.44	0.41	0.84	0.82	0.29
62	High	B	0.48	0.56	0.83	0.85	0.48
63	High	B	0.61	0.82	0.88	0.95	0.28



	A	B	C	D	E	F	G	H
1	Condition	Group	Pre1	Pre2	Pre3	Pre4	Learning	Gender
2	Low	A	0.77	0.91	0.24	0.72		0.9 M
3	Low	A	0.82	0.91	0.62	0.9		0.87 F
4	Low	A	0.81	0.7	0.43	0.46		0.9 F
5	Low	A	0.88	0.89	0.2	0.63		0.85 M
6	Low	A	0.78	0.68	0.25	0.73		0.93 F
7	Low	A	0.74	0.9	0.99	0.99		0.93 M
8	Low	A	0.78	0.86	0.79	0.78		0.89 F
9	Low	A	0.76	0.76	0.61	0.85		0.8 F
10	Low	A	0.93	0.82	0.99	0.99		0.98 M

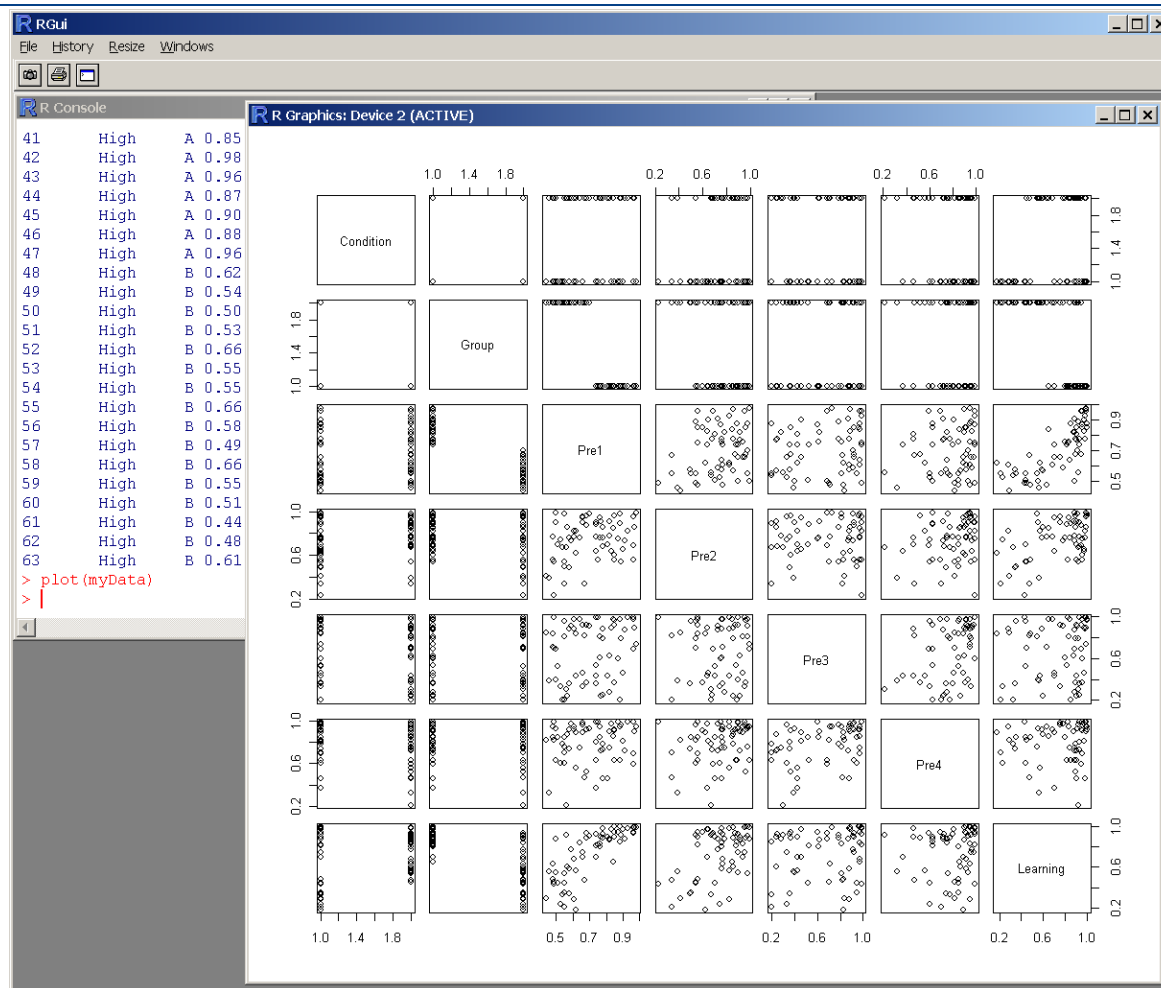


Condition	Group	Pre1	Pre2	Pre3	Pre4
Low	A	0.77	0.91	0.24	0.72
Low	A	0.82	0.91	0.62	0.9
Low	A	0.81	0.7	0.43	0.46
Low	A	0.88	0.89	0.2	0.63
Low	A	0.78	0.68	0.25	0.73
Low	A	0.74	0.9	0.99	0.99
Low	A	0.78	0.86	0.79	0.78
Low	A	0.76	0.76	0.61	0.85
Low	A	0.93	0.82	0.99	0.99
Low	A	0.82	0.78	0.28	0.75
Low	A	0.91	0.73	0.87	0.72
Low	A	0.96	0.69	0.69	0.59
Low	A	0.97	0.86	0.89	0.9
Low	A	0.89	0.54	0.79	0.96
Low	A	0.76	0.94	0.81	0.95
Low	A	0.84	0.85	0.97	0.86
Low	B	0.62	0.82	0.43	0.56

INSTALLING, RUNNING, AND INTERACTING WITH R

Examining datasets:

```
> plot(myData)
```



HOW TO FIND R HELP AND RESOURCES ON THE INTERNET

R wiki:

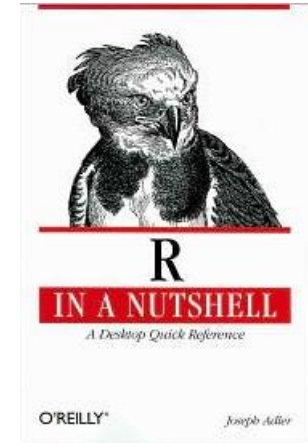
<http://rwiki.sciviews.org/doku.php>

R graph gallery:

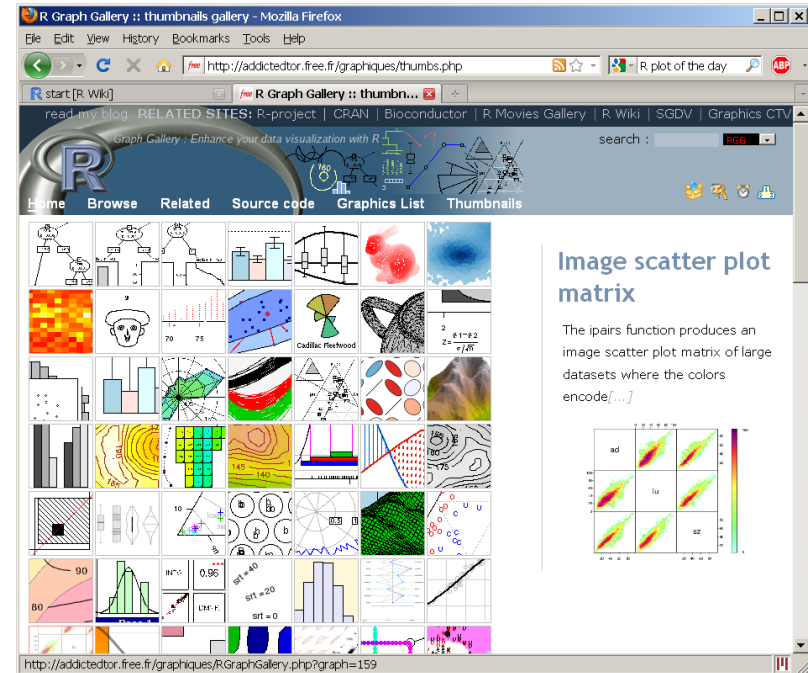
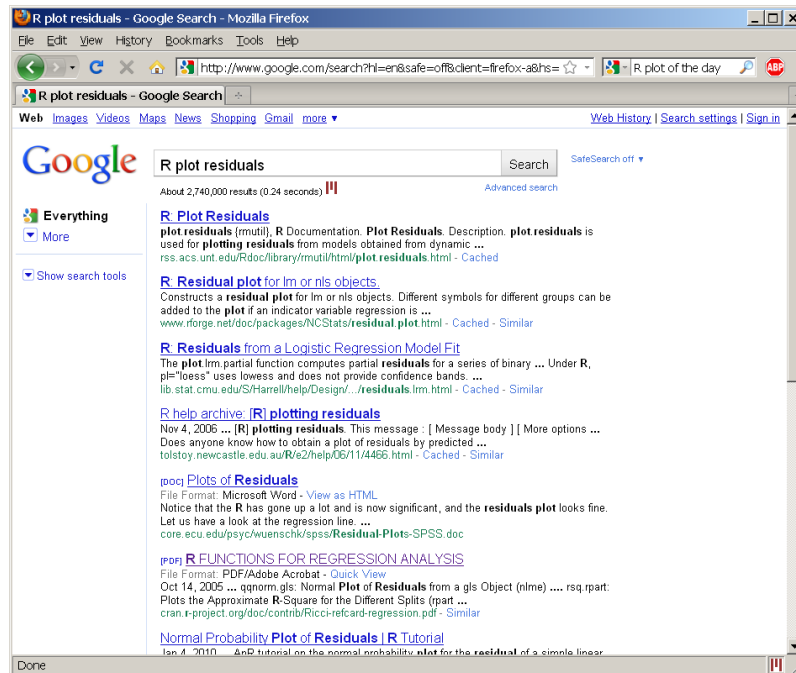
<http://addictedtor.free.fr/graphiques/thumbs.php>

Kickstarting R:

<http://cran.r-project.org/doc/contrib/Lemon-kickstart/>



ISBN: 9780596801700



FOR MORE RESOURCES, CHECK OUT...

- R home page
 - <http://www.r-project.org>
- R discussion group
 - <http://www.stat.math.ethz.ch/mailman/listinfo/r-help>
- Cheat sheets:
 - <https://www.rstudio.com/resources/cheatsheets/>
- Search Google for R and Statistics
 - Typically the most useful hits are those from the www.stackoverflow.com blog