R Basics for Math 283

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R is available at:

http://www.r-project.org/

You can find its documentation in HTML and PDF formats at:

http://cran.r-project.org/manuals.html

There's also a shortened list of commands available at:

http://www.personality-project.org/r/r.commands.html

NOTE: These instructions are written based on a Linux platform and may need some slight modifications for Windows/Mac.

Starting up R

In Windows, you can find R in the programs list from the start menu.

In Mac from the Application menu.

In Unix (Mac/Linux), from a terminal command prompt:

Create a directory to work in, and go to that directory:

\$ mkdir work
\$ cd work

Start R in that directory:

\$ F

">" is the R prompt. To guit R, simply type:

> q(

<u>Note:</u> If you are using R remotely with the X Window System, make sure to connect with "ssh -X" or "ssh -Y". However, it is better to install R locally.

Assignment, Vector, Matrices, and Operators

Assignment

Variable are assigned a value using either the **assign()** function, the operator <-, the operator -> or the operator =. The operators <- and = are equivalent.

> assign("var", 1)	> var <- 2	> 3 -> var	> var = 4
> var	> var	> var	> var
[1] 1	[1] 2	[1] 3	[1] 4

Comments

To add comments to the code we simply use the # sign before the statement.

Vectors

Vectors can be defined in R using the function **c()**.

```
> x <- c(1,2,3)
> x
[1] 1 2 3
> x2 <- c(x,0,x)
> x2
[1] 1 2 3 0 1 2 3
```

You can also automatically generate vectors by using the **seq()** and **rep()** functions as illustrated below.

```
> x3 <- seq(1,5, by=.5)
> x3
[1] 1.0 1.5 2.0 2.5 3.0 3.5 4.0 4.5 5.0
> x4 <- rep(x, times=2)
> x4
[1] 1 2 3 1 2 3
> x5 <- rep(x, each=2)
> x5
[1] 1 1 2 2 3 3
```

If the step size in the function seq() is 1 or -1, the colon: can be used.

```
> y <- 10:1
> y
[1] 10 9 8 7 6 5 4 3 2 1
> z <- 1:10
> z
[1] 1 2 3 4 5 6 7 8 9 10
```

Accessing Vector Elements

You can access vector elements using the *square brackets* []. You can specify multiple indexes by using a vector.

```
> v <- 1:10
> v[6]
[1] 6
> v[c(1,5,8)]
[1] 1 5 8
```

Negative indexes specify the values to be excluded instead of included.

```
> v[-(3:8)]
[1] 1 2 9 10
```

Arrays and Matrices

You can create matrices by using the **array()** function, which takes the form **array(***data_vector***,** *dim_vector***)**. If there are less values in the data vector than the dimension of the array calls for, then the values are recycled.

> array(1:4, c(2,2))	[,1] [,2] [1,] 1 3 [2,] 2 4
> array(1:4, c(2,3))	[,1] [,2] [,3] [1,] 1 3 1 [2,] 2 4 2
> array(0, c(2,2))	[,1] [,2] [1,] 0 0 [2,] 0 0

Accessing Array Elements

Array elements can be accessed in the same way as vectors.

A[i] accesses the i-th element in matrix A (going by columns). > A[3]	[1] 3
A[i,] accesses the i-th row in matrix A. > A[1,]	[1] 1 3
A[,j] accesses the j-th column in matrix A. > A[,1]	[1] 1 2
A[i,j] accesses the element in the i-th row, j-th column in A . $A[i,j]$	[1] 3

Also, you can create matrices by using matrix(data_vector, number_rows, number of cols). For example:

Accessing the elements remains the same no matter the function used to create the matrix.

Some Basic Functions on Vectors and Arrays

Type 'ne	eip(<i>tunct</i>	ion_nam	<i>ie)</i> for a	etails on	usage.					
dim	t	max	min	mean	var	sd	sort	sum	prod	

Operators

Arithmetic operators in R

·	iic operaiors in it					
	*	/	+	ı	^	

are *element-by-element operators*, and follow the expected order of operations.

In order to do matrix operations which follow the rules of linear algebra, you must use the following operators (%*% is the dot product (inner product) and %0% is the outer matrix multiplication):

8*8 8O8

Here's what a few operators would return:

> x <- c(1,2,3) > y <- c(4,5,6)	
> x+y	[1] 5 7 9
> x-y	[1] -3 -3 -3
> x*y	[1] 4 10 18
> x/y	[1] 0.25 0.40 0.50
> x^y	[1] 1 32 729
> x%*%y	[,1] [1,] 32
> x%o%y	[,1] [,2] [,3] [1,] 4 5 6 [2,] 8 10 12 [3,] 12 15 18

Note that a vector does not have a "direction", i.e. the function **c()** does not create row vectors or column vectors, just vectors.

More Basic Operators/Functions

diff

exp	log	log10	log2	sqrt

Workspace and File Management

The following functions manage your file system and the variables in your R workspace.

browse.workspace()	creates a window with information about all variables in the	
	workspace	
ls()	list the variables in the workspace	
rm(x)	remove x from the workspace	
rm(list=ls())	remove all the variables from the workspace	
dir()	list the directory/files of the current directory.	
setwd(my_dir)	change current working directory to my_dir	
unlink(filename)	deletes the file <i>filename</i> .	

Reading/Writing Data from/to a File

Data frames

R encourages the use of *data frames* which are matrices in which each column has a title and can be of a different type.

```
> df <- data.frame(name=c("Joe", "Tom"), age=c(21,22))
> df
   name age
1 Joe 21
2 Tom 22
```

Writing data to file

The function write.table() allows to write data to a file. However, all will be converted to a data frame first.

```
> write.table(df, file="example.txt")
    example.txt:
    "name" "age"
    "1" "Joe" 21
    "2" "Tom" 22
```

Notice how R added row labels for you. This can be a problem when trying to write a matrix to a file:

Luckily, if you do not want to deal with data frames, there is a way around it: write.table(array(1:4, c(2,2)), file="example.txt", row.names=FALSE, col.names=FALSE)

```
example.txt:
```

Reading data from file

Data files can be read using the table.read() function.

Let us take the last file we wrote.

```
example.txt:

1 3
2 4
```

Then, we can read its content in the following manner.

```
> A <- read.table("example.txt")
> A
     V1     V2
1     1     3
2     2     4
```

You can use the data frame like a matrix, but if you want to convert it back to a matrix, you can do that in the following way:

```
> array(unlist(A, use.names=FALSE), dim(A))
```

Or the command:

```
> matrix(unlist(A, use.names=FALSE), dim(A))
```

With either command the output would be:

```
[,1] [,2]
[1,] 1 3
[2,] 2 4
```

Control Flow

Conditional Control - if, else, ifelse

Conditional control is achieved using the if/else statements.

```
> if (expr 1) expr 2 else expr 3
```

This first evaluates *expr_1*. If it is true, then *expr_2* is executed, otherwise, *expr_3* is executed.

There is a vectorized version of the if/else construct, the **ifelse()** function. This has the form **ifelse(condition, a, b)** and returns a vector of the length of its longest argument, with elements a[i] if condition[i] is true, otherwise b[i].

```
> x <- 0
> if (x == 0) x = 1 else x = 2
> x
[1] 1
> if (x == 0) x = 1 else x = 2
> x
[1] 2
> ifelse(c(TRUE, FALSE), c(1,1), c(2,2))
[1] 1 2
```

Repetitive Execution - for, while, break

Statements can be repeated a specific number of times using for loops.

```
> for (name in expr 1) expr 2
```

This causes *name* to iterate through a vector *expr_1* and execute *expr_2* for each value of *name*.

For example, we could add the numbers from 1 to 10:

```
> x < -0
> for(i in 1:10) x = x+i
> x
[1] 55
```

We can also repeat a statement (or set of statement) until we fulfill a condition by using **while** loops.

```
> while (expr_1) expr_2
```

In this case, expr 2 is executed until expr 1 becomes false.

We can do the same thing as in the previous example using a while loop. For the use of curly braces, see the section **compound statements**.

```
> i <- 1
> x <- 0
> while (i <= 10) {
+    x = x+i;
+    i = i+1;
+ }
> x
[1] 55
```

The **break** statement allows for an unexpected stop in the loop. Whenever the program reaches a **break** statement, it immediately exits the loop it is in.

Compound Statements

Let's consider the previous example:

```
> while (i <= 10) {
+     x = x+i;
+     i = i+1;
+ }</pre>
```

We can see that the executed expression is actually a series of two statements (x = x+i and i = i + 1) encapsulated by curly brackets { }. This is a compound statement. When a list of statements is surrounded by curly brackets, it is considered to be one expression. In this case, both statements will be executed at each iteration of the while loop.

Simple Plots

Basic Commands

plot(x, y)	If x and y are vectors, plot(x, y) produces a scatterplot of y against x. The same eff ect can be produced by supplying one argument (second form) as either a list containing two elements x and y or a two-column matrix. See next section.	
plot(x)	If x is a time series, this produces a time-series plot. If x is a numeric vector, it produces a plot of the values in the vector against their index in the vector.	
hist(x)	Produces a histogram of the numeric vector x. A sensible number of classes is usually chosen, but a recommendation can be given with the nclass= argument. Alternatively, the breakpoints can be specified exactly with the breaks= argument.	
axis	Set specified x- and y-axis limits. Usage: axis([xmin xmax ymin ymax])	
title(main, sub)	Adds a title main to the top of the current plot in a large font and (optionally) a sub-title sub at the bottom in a smaller font.	

Arguments to High-Level Plotting Commands (plot, hist, ...)

add=TRUE	Forces the function to act as a low-level graphics function, superimposing the plot on the current plot (some functions only).	
axes=FALSE	Suppresses generation of axes useful for adding your own custom axes with the axis() function. The default, axes=TRUE, means include axes.	
log="x" log="y" log="xy"	Causes the x, y or both axes to be logarithmic. This will work for many, but not all, types of plot.	
type="p" type="l" type="o" type="h" type="s" type="S" type="n"	Plot individual points (the default) Plot lines Plot points connected by lines (both) Plot points overlaid by lines Plot vertical lines from points to the zero axis (high-density) Step-function plot. The top of the vertical defines the point. Step-function plot. The bottom of the vertical defines the point. No plotting at all. However axes are still drawn (by default) and the coordinate system is set up according to the data. Ideal for creating plots with subsequent low-level graphics functions.	
xlab=string ylab=string	Axis labels for the x and y axes. Use these arguments to change the default labels.	
main=string	Figure title, placed at the top of the plot in a large font.	
sub=string	Sub-title, placed just below the x-axis in a smaller font.	

Superimposition

You can superimpose one plot into another with the command **par(new=TRUE)** but it is necessary to modify other settings to obtain the desired axes and labels. An example would be:

```
> plot(x, y)
> par(new=TRUE)
> plot(z, t)
```

You can also add a parameter *add=TRUE* to many plot commands.

For scatter plots, you can use **plot()** for the first one and **points()** to superimpose additional plots.

Multiple plots

To have multiple plots or graphs in the same output window, you can use the commands **split.screen()**, **screen()**, **close.screen()**. An example would be:

```
> split.screen(c(1,2))
> screen(1)
> plot(x, y)
> screen(2)
> plot(z, t)
> close.screen()
```

There exist other ways to do this with the functions par() (see parameters mfrow, mfcol, and mfg) and layout(). An example showing how to control the margins around and between multiple plots in one figure is at

http://research.stowers-institute.org/efg/R/Graphics/Basics/mar-oma/index.htm

There are also device-dependent ways to put up multiple plot windows (as opposed to multiple plots in one window). On Mac OS X, use **quartz()**.

Saving plots

You can save your plot using the functions **postcript()**, **pdf()**, **png()** and **jpeg()**. You must first start to record your plot, then write your plot, then you can turn off your "recording" driver by using the **dev.off()** function.

```
> png(file="test.png")
> plot(x, y)
> dev.off() #turns off write-to-file device
```

A second method is to draw the plot in the GUI and output it to a file::

Scripts and Functions

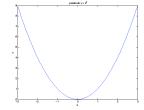
An R-file is a text file that contains R commands and has a .R filename extention. You can edit files from the R prompt using the vi editor (it may need to be installed).

```
> vi (file = "myfilename.R")
```

However, if you are not familiar with vi, I would strongly suggest you use another text editor. The Windows version of R comes with an editor.

The name of the file does not have to match the name of the function. You have to source the file in order to have the functions available in your workspace. Sourcing a file is the equivalent to copying all its content and pasting it into your R command prompt. In the Windows version of R, you can use the *source* item from the menu.

Script R-file. No input or output arguments. Just a series of commands.



```
Example: script.R.
x <- seq(-3,3,by=.1)
y <- x^2
png(file="parabola.png")
plot(x, y, xlab="x", ylab="y",
    main="parabola y = x^2", type="l",
    col="blue")
dev.off() #turns off write-to-file device</pre>
```

The file can then be ran in the R prompt using the **source()** function.

```
> source("script.R")
```

Function R-files. Contain a function definition line and can accept input arguments and return output argument.

Example: stats_wrapper.R

```
stats_wrapper <- function(x) {
    # stats_wrapper.R
    # computes the mean and variance of vector x
    mean_x = mean(x)
    var_x = var(x)
    c(mean_x, var_x)
}</pre>
```

Example usage:

```
> source("stats_wrapper.R")
> stats_wrapper(c(1,2,3))
[1] 2 1
```

Statistics in R

Some Distributions

Functions are provided to evaluate the cumulative distribution function $P(X \le x)$, the probability density function and the quantile function (given q, the smallest x such that $P(X \le x) > q$), and to simulate from the distribution.

Distribution	R name	Additional arguments
binomial	binom	size, prob
chi-squared	chisq	df, ncp
exponential	exp	rate
F	f	df1, df2, ncp
geometric	geom	prob
hypergeometric	hyper	m, n, k
negative binomial	nbinom	size, prob
normal	norm	mean, sd
Poisson	pois	lambda
student's t	t	df, ncp
uniform	unif	min, max
Wilcoxon	wilcox	m, n

Prefix the name given here by 'd' for the density, 'p' for the CDF, 'q' for the quantile function (inverse) and 'r' for simulation (random deviates). The first argument is x for dxxx, q for pxxx, p for qxxx and n for rxxx (except for rhyper and rwilcox, for which it is nn).

For example, let's say we flip a fair coin 3 times and let X be the number of heads. Recall from lecture that the random variable X would follow a binomial distribution with parameters n=3 and $p=\frac{1}{2}$. As a quick reminder, here's how the pdf and the cdf of X look like:

pdf: $P(X = x)$		
x = 0	1/8	
x = 1	3/8	
x = 2	3/8	
x = 3	1/8	
all other x	0	

cdf: $P(X \le x)$	
<i>x</i> < 0	0
$0 \le x < 1$	1/8
$1 \le x < 2$	1/2
$2 \le x < 3$	7/8
$x \ge 3$	1

Let's find the probability that we get exactly 2 heads. We use the pdf of X at x = 2. \Rightarrow dbinom(2, 3, 1/2)

[1] 0.375

Now, let's find the probability that we get 2 heads or less. That's the cdf of X at x = 2. > pbinom(2, 3, 1/2)

[1] 0.875

Now, let's what the smallest value of x such that $P(X \le x) \ge 3/4$. If we look at the cdf table, we can see that we achieve this at x = 2 where $P(X \le x) = 7/8$.

> qbinom(3/4, 3, 1/2)

[1] 2

Finally, let's generate 10 random values of X according to its distribution.

> rbinom(10, 3, 1/2)

[1] 2 2 2 1 0 2 3 2 1 1

(you will get different values since it is random)

Notice how the values 1 and 2 appear more often. That's consistent with the pdf of X.

Some Hypothesis Tests

t.test(x, mu=my_mu)	One sample t-test for mean <i>my_mu</i> .
t.test(x, y)	Welch's two-sample t-test
t.test(x, y, var.equal=TRUE)	Student two-sample t-test (classic)
t.test(x, y, paired=TRUE)	Paired two-sample t-test.
wilcox.test(x, y)	Wilcoxon rank sum test (Mann-Whitney U)
wilcox.test(x, y, paired=TRUE)	Wilcoxon signed rank test.

help (function name)

Best way to know how to correctly use built-in function and understand what it is doing.

Sources:

- An Introduction to R. http://cran.r-project.org/doc/manuals/R-intro.html
- http://www.personality-project.org/r/r.commands.html

Exercises

Generate values from a binomial distribution

Generate 10,000 random values from a binomial distribution and store in row vector *empirical_bino*. Use help to figure out input values for *rbinom*.

```
> empirical bino <- rbinom(10000, 100, 1/2)
```

Check that the mean and variance are close to that of the distribution from which it was generated.

```
> mean(empirical_bino)
> var(empirical_bino)
```

Since binomial with p=1/2 is symmetric, check that median is also near true mean.

```
> median(empirical_bino)
```

Check out the distribution of the values in empirical bino by plotting the histogram.

```
> hist(empirical bino)
```

Draw/trace plot of the distribution over the histogram.

```
> prop <- hist(empirical_bino, xlab = 'bins', ylab='counts')
> lines(prop$mids, prop$counts, col="blue")
> b = prop$breaks
> binsize = b[2] - b[1]
> lines(0:100, 10000*binsize*dbinom(0:100, 100, 1/2), col="red")
```

Working with the cdf

Get the probability that values from B(100,1/2) are less than or equal to the mean of the values in *empirical bino*. Do the same for the median.

```
> pmean = pbinom(mean(empirical_bino), 100, 1/2)
> pmedian = pbinom(median(empirical bino), 100, 1/2)
```

Use the cdf to get back the critical values corresponding to those probability cutoffs.

```
> qbinom(pmean, 100, 1/2)
> qbinom(pmedian, 100, 1/2)
```

Plot the cdf of empirical bino

We need to sort the values in empirical_bino from low to high and get the corresponding probability $Px = P(X \le x)$ for each value in the array according to the theoretical distribution.

```
> Px = pbinom(sort(empirical bino), 100, 1/2)
```

Plot the sorted values of empirical_bino on x-axis and corresponding theoretical probabilities on y-axis.

```
> plot(sort(empirical_bino), Px, xlab="x", ylab="P_x", main="cdf
of empirical bino", type="l", col="blue")
```

Odds and Ends

```
Let us practice file i/o with the first 10 numbers of our empirical distribution > write.table(empirical_bino[1:10], "empirical_bino.txt") > read.table("empirical_bino.txt")  
> unlink("empirical_bino.txt")  
# deletes file
```

Function example

Create simple function called sim_bino that plots the distribution of m values generated from a binomial distribution with parameters n and p, saves the plot to a file and returns the mean and variance of the observed distribution.

The function call should look like the following (n=100, p=1, m=100000, filename=test.png):

```
>>[obs mean, obs var] = sim bino(100,1/2,10000, 'test.png')
```

File: sim bino.R

```
function(n,p,m,plotname) {
    # Simple function that plots distribution of m values from a
    # binomial distribution with parameters n and p and saves the
    # plot to file plotname

empirical_bino <- rbinom(m, n, p)

png(plotname)
prop <- hist(empirical_bino, xlab = "bins", ylab="counts")
lines(prop$mids, prop$counts, col="blue")
binsize = prop$breaks[2] - prop$breaks[1]
lines(0:100, 10000*binsize*dbinom(0:100, 100, 1/2), col="red")
dev.off()

c(mean(empirical_bino), var(empirical_bino))
}</pre>
```