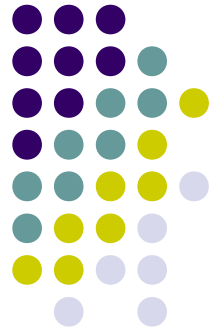


Lecture #2

The R language and environment





Outline

- What is R?
- Basic syntax rules
- Math operators
- Vectors, Matrices, and Data frames
- Logical operators
- Casting data types
- Read/Write
- Iteration
- Packages
- Libraries/Functions
- Plotting commands
- Par command
- Data sets in R



What is R?

- R is both a language and environment for statistical and computational analysis
- Where does the name come from?
 - Authors: Robert Gentleman and Ross Ihaka and S language play-on words
- The language is similar to that of Splus, which gets its roots from S
 - Developed by Bell Labs
- R was developed as a collaborative project from multiple contributors (CRAN)
 - You can readily download and contribute “packages” to R
- R software is available free, however there is no support
 - Splus (on the other hand) is not free, so support is provided



Basic syntax rules

- Procedures in R are implemented as functions
 - “()” syntax
 - i.e. `c(4,5,6)` means to concatenate the three numbers into a vector
- R processes are best handled as objects
 - If you run any function, assign it to a variable (object) so you can manipulate the attributes in the object
- Numerical objects are interpreted best in formats of matrices, vectors, data frames, or lists
 - Statistical language vs. computer language
- **!To get help with any function in R, use the help function!**
 - **help(mean)**
 - **?mean**
- To quit out of R, use the quit function
 - `q()`
- Variable assignment is “=” or “<-”



Basic syntax rules (cont.)

- Annotation syntax is “#” (R will not read comments that start with this symbol)
- To view the objects in the database
 - `ls()`
- Remove an object in the database
 - `rm()`
- Concatenate/combine multiple numbers/objects together
 - `c()`
- Underscores (“_”) are not handled well in variable names
 - The underscore is synonymous with the “=” and “<-” assignment operators
 - Try to avoid underscores in scripts
- Calculate a mean and variance
 - `mean()`
 - `var()`
- View current memory size and set memory limits
 - `memory.size(T)`
 - `memory.limit`



Mathematical Calculations

- General math operators
 - $> 5 + 5$
 - $> 10 - 2$
 - $> 10 * 10$
 - $> 25 / 5$
 - $> 3 ^ 2$
 - $> \text{exp}(2)$
 - $> \text{log}(10)$
 - $> \text{logb}(10,2)$



Vectors

- Vector syntax
 - Create a vector
 - `> x <- c(2,4,6,3,4,6)`
 - `> y <- c(5,6,7,8,8,0)`
 - Specify certain elements of the vector
 - `> x[1:3]`
[1] 2 4 6
 - Remove an element from the vector
 - `> x[-2]`
[1] 2 6 3 4 6
 - Bind 2 vectors together (must be same length)
 - `> x.y.bound <- cbind(x,y)` # cbind means column bind



Misc. vector operations

- Intersect elements of 2 vectors
 - `> intersect(x,y)`
[1] 6
- Diff 2 vectors
 - `> setdiff(x,y)` # what is in x that is not in y?
[1] 2 4 3
- Find length of x vector
 - `> length(x)`
[1] 6



Matrices

- Matrix syntax
 - Create a matrix
 - `> x.matrix <- matrix(data=x,nrow=3,ncol=2)`
 - `> x.matrix`

```
      [,1] [,2]  
[1,]  2  3  
[2,]  4  4  
[3,]  6  6
```
 - Look at matrix dimensions
 - `> dim(x.matrix)`

```
[1]  3  2
```
 - Specify elements in the matrix
 - `> x.matrix[1:2,2]` # rows 1 and 2 of column 2

```
[1]  3  4
```
 - `> x.matrix[1,]` # rows 1, all columns

```
[1]  2  3
```
 - Transpose the matrix
 - `> x.matrix.t <- t(x.matrix)`
 - `> x.matrix.t`

```
      [,1] [,2] [,3]  
[1,]  2  4  6  
[2,]  3  4  6
```



Data frames

- Similar syntax to matrices, but have row and column names
 - Create a data frame
 - `> x.df <- data.frame(x,y)`
 - `> x.df`

```
x y
1 2 5
2 4 6
3 6 7
4 3 8
5 4 8
6 6 0
```

has row names (1-6) and column names (x and y)
 - Look at row names
 - `> dimnames(x.df)[[1]]`

```
[1] "1" "2" "3" "4" "5" "6"
```
 - Look at column names
 - `> dimnames(x.df)[[2]]`

```
[1] "x" "y"
```



Logical Operators

- R works very well with boolean logic
 - Look for values greater than 4 in x.matrix
 - `> x.matrix.g4 <- x.matrix>4`
 - `> x.matrix.g4`

```
      [,1] [,2]  
[1,] FALSE FALSE  
[2,] FALSE FALSE  
[3,] TRUE TRUE
```
 - Print only those values out
 - `> x.matrix[x.matrix.g4]`

```
[1] 6 6
```



Casting

- Casting is basically changing the data-type from one type to another
 - i.e. data is in matrix format and you wish to convert it to a vector:
 - use the “as.” syntax followed by the desired data-type
 - `> x.vector <- as.vector(x.matrix) # change to vector`
 - `> x.vector`
`[1] 2 4 6 3 4 6`
 - `> x.df <- as.data.frame(x.matrix) # change to dataframe`
 - `> x.df`
 `V1 V2`
`1 2 3`
`2 4 4`
`3 6 6`
 - `> x.char <- as.character(x.vector) # change to character`
 - `> x.char`
`[1] "2" "4" "6" "3" "4" "6"`
- Can also check type of data object with “mode” and “class” functions



Read in/Write out

- Multiple ways to read a data file in
 - `> read.table(file="C:\\Class\\data.txt",header=T)`
 - must use `"\\"` instead of `"\"` for path
 - `> scan(file="C:\\Class\\data.txt")`
 - must use `"\\"` instead of `"\"` for path
 - best for vectors or lists (not best for 2D data)
- Writing data out to a file
 - `> write.table(x.matrix,file="dataFile.txt",sep="\t")`
 - `> write(x.matrix,file="dataFile.txt",ncolumns=2)`



Iteration & If Statements

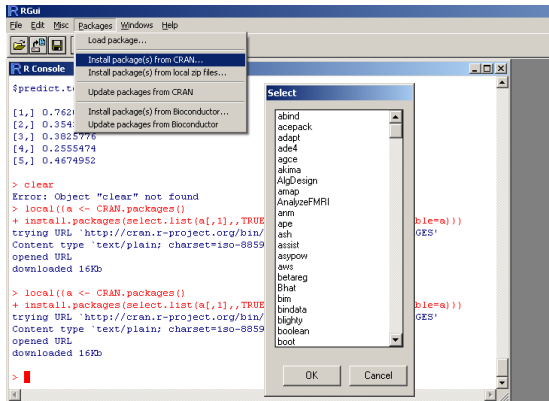
- Loops are not optimized in R
 - They are possible, but not memory favorable
- Loop syntax
 - `> for(i in 1:6) {}` # for loop
 - `> while(i < 7) {}` # while loop
- apply statements
 - Inherent looping mechanism
 - Much more memory efficient
 - `> apply(x.matrix,1,mean)`
 - Will calculate a mean on each row in x.matrix
 - `> apply(x.matrix,2,mean)`
 - Will calculate a mean on each column in x.matrix
- if statements are fairly well implemented
 - `if(x[1]>5) {x[1]=6}` # simple if command
 - `ifelse(x[1]>5,x[1]=6,x[1]=9)` # combine if and else into one statement

Packages

- Binary files that build library access structure
 - Allow specific functions to be accessible
 - If you know the package name, you can simply type: `install.packages("tree")`
 - "tree" is the example package in this case

Can install packages from CRAN or Bioconductor

Can install packages from local drive (if saved zip file is on local drive)



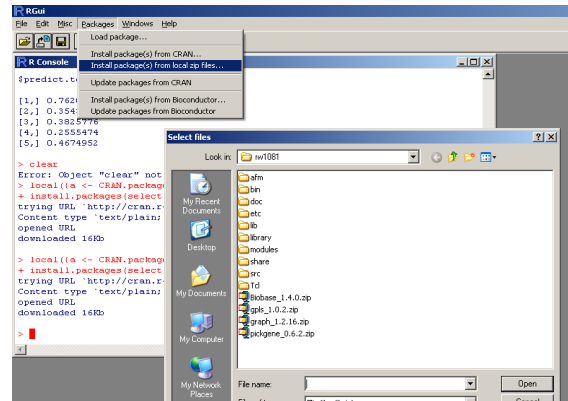
The screenshot shows the R GUI interface. The R Console window displays the following code and output:

```
> predict.t
[1.] 0.7621  Install package(s) from CRAN...
[2.] 0.354   Update packages from Bioconductor
[3.] 0.3825776
[4.] 0.2555474
[5.] 0.4674952

> clear
Error: Object "clear" not found
> local((a <- CRAN.packages())
+ install.packages(select.list(a[,1]), TRUE
trying URL 'http://cran.r-project.org/bin/
Content type 'text/plain'; charset=iso-8859
opened URL
downloaded 16Kb

> local((a <- CRAN.packages())
+ install.packages(select.list(a[,1]), TRUE
trying URL 'http://cran.r-project.org/bin/
Content type 'text/plain'; charset=iso-8859
opened URL
downloaded 16Kb
```

A "Select" dialog box is open, showing a list of packages to be installed. The list includes: abind, acoPACK, adapt, ade4, agge, akima, AlgDesign, amap, AnalyseFMRI, arm, ape, ash, assist, asyppow, aws, betareg, BHat, bin, biada, bihigh, boolean, and boot. The "OK" button is highlighted.



The screenshot shows the R GUI interface. The R Console window displays the following code and output:

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[1.] 0.7621  Install package(s) from Bioconductor...
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opened URL
downloaded 16Kb

> local((a <- CRAN.packages)
+ install.packages(select
trying URL "http://cran.r
Content type 'text/plain;
opened URL
downloaded 16Kb

>
```

A "Select files" dialog box is open, showing a file explorer view of the local drive. The file list includes: afm, bin, cfile, etc, fb, library, modules, share, src, Id, package_1.4.0.zip, gds_1.0.2.zip, graph_1.2.16.zip, and package_0.6.2.zip. The "File name" field is empty, and the "Files of type" dropdown is set to "Zip files (*.zip)". The "Open" button is highlighted.





Libraries & Functions

- Once package has been installed, library can be accessed
- Libraries include many functions
 - `> library(base)` # implements base library
 - Allows access to functions inside the library
- Data sets internal to R can also be accessed and utilized
 - `> data(iris)` # allows access to famous iris flower data set
 - MORE EXPANATION ON THIS FURTHER IN LECTURE
- Functions
 - Handled well in R
 - Similar to a subroutine in other languages (packaged operation that is called)
 - ```
> square <- function(x=num1,y=num2) {
 r1 <- x^2
 r2 <- y^2
 output <- c(r1,r2)
 return(output)
}
```
  - ```
> square(x=4,y=2)  
[1] 16 4
```

 # call square function



Packages & Libraries

- You can view the complete list of installed packages
 - `> .packages(all = TRUE)`
- Also view the list of current attached packages
 - `> (.packages())`
- Can set the path for R to look for the libraries
 - `> .libPaths("C:/PROGRA~1/R/RW2000~1/library")`
- Remove package from session
 - `> detach("package:stats")`
- View package contents (e.g. functions, author, date, version, description, etc.)
 - `> library(help=odesolve)`



Plotting commands

- There are many optional commands available in R for plotting data
 - Only a few of the main commands are necessary
- Scatter plot
 - `> plot(x,y)` # x and y must be same length and numeric
 - Labels
 - `xlab='x vector'` # x-axis label
 - `ylab='y vector'` # y-axis label
 - `main='Scatter plot'` # title for plot
 - Points, size, and colors
 - `pch='*'` # symbol to plot
 - `cex=1.5` # size of symbol
 - `col='red'` # color of points

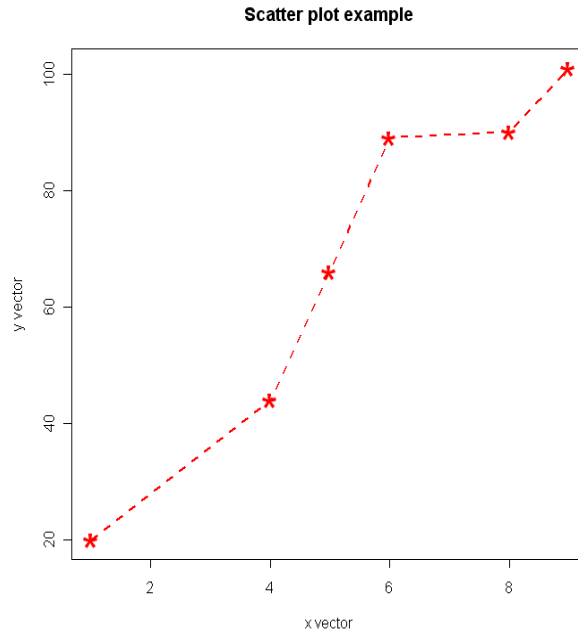


Plotting commands (steps)

- Plotting function can be given in subsequent commands for modifications to the graph
- Scatter plot options
 - Plot without points/lines
 - `type='n'` # no points or lines are not drawn
 - `type='b'` # lines and points drawn together
 - `type='p'` # only points are draw
 - `type='l'` # only lines are drawn
 - Points only
 - `points(x,y,col='blue')`
 - Lines only
 - `lines(x,y,col='yellow')`
 - `lwd(2)` # line width
 - `lty(3)` # line pattern

Plot example

```
> x <- c(1,4,5,6,8,9)
> y <- c(20,44,66,89,90,101)
> plot(x,y,type='n',xlab='x vector',ylab='y vector',main='Scatter plot example')
> points(x,y,col='red',pch='*',cex=3)
> lines(x,y,col='red',lwd=2,lty=2)
```

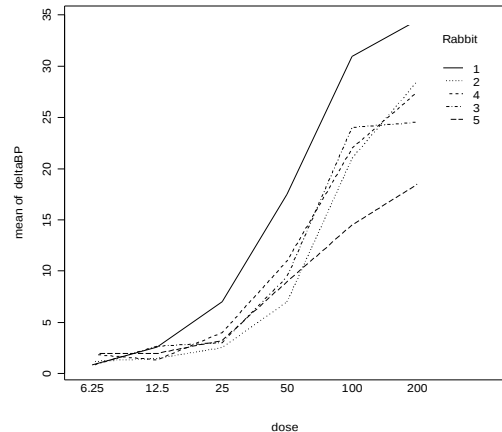
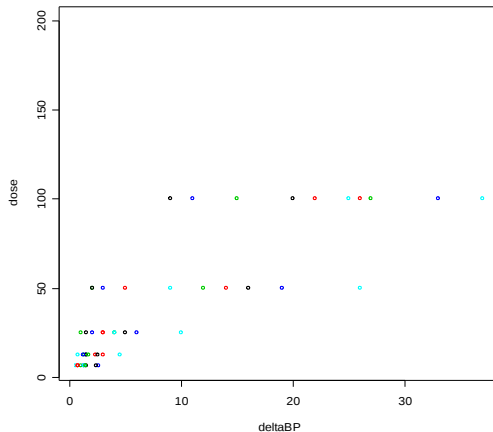


Plotting commands (options)



- Data can be plot using grouping variable commands

```
> library(nlme)
> data(PBG)
> attach(PBG)
> plot(deltaBP dose)
> plot(deltaBP dose, col=as.integer(Rabbit))
> interaction.plot(dose, Rabbit, deltaBP)
```





Par commands

- Page formatting to add more than 1 plot per page
 - `> par(mfrow=c(2,3))` # 6 total plots (2 rows & 3 columns)
- Page formatting to add margin sizes
 - `> par(oma=c(2,4,2,4))` #order is bottom, left, top, right
- Multiple other page format options
 - See help section for “par”



Data sets in R

Using packages in Bioconductor, multiple microarray data sets are available

```
library(Biobase);  
library(annotate);  
library(golubEsets);  
library(multtest);
```

Golub et al. AML/ALL Affy data set (training set and test set are separate)

```
data(golubTrain);  
data(golubTest)  
dat.train <- exprs(golubTrain)           (7129 genes x 38 samples)  
data(golubTest)  
dat.test <- exprs(golubTest)            (7129 genes x 34 samples)  
class <- pData(phenoData(golubTrain))   (class labels for training set samples)
```

or

```
data(golub)  
smallgd<-golub           (7129 genes x 38 samples)  
classlabel<-golub.cl    # class labels for training set samples
```

Alon et al. colon cancer Affy data set

```
library(colonCA)  
data(colonCA)  
dat <- exprs(colonCA)       (2000 genes x 62 samples)  
classlabel <- colonCA$class # class labels for samples
```

Some unidentified Affy data set

```
data(geneData)  
dat <- geneData           (500 genes x 26 samples)
```


Getting data sets from CRAN (cont.)



Once the zip file has been downloaded to a local directory, go into R and use package instructions to install zip file from local directory

Then,

access data from data object as follows:

```
# load package from local drive
> library(Biobase)
> library(annotate)
> library(colonCA)
> data(colonCA)
> dat <- exprs(colonCA)           # expression data
> ann.dat <- colonCA$class       # annotations for samples
```



References

- R Primer
 - <http://www.r-project.org/>
 - <Documentation>
 - <Manuals>
 - <An Introduction to R> or
 - <The R Reference Index>
- Introductory Statistics with R, Peter Dalgaard