

Recap From Last Time:

- Substitution matrices: Where our alignment match and mis-match scores typically come from
- Comparing methods: The trade-off between *sensitivity*, *selectivity* and *performance*
- Sequence motifs and patterns: Finding functional cues from conservation patterns
- Sequence profiles and position specific scoring matrices (PSSMs), Building and searching with profiles, Their advantages and limitations
- PSI-BLAST algorithm: Application of iterative PSSM searching to improve BLAST sensitivity
- Hidden Markov models (HMMs): More versatile probabilistic model for detection of remote similarities

Today's Learning Goals

- Familiarity with R's basic syntax.
- Familiarity with major R data structures.
- Understand the basics of using functions.
- Be able to use R to read and parse comma-separated (.csv) formatted files ready for subsequent analysis.
- Appreciate how you can use R scripts to aid with reproducibility.



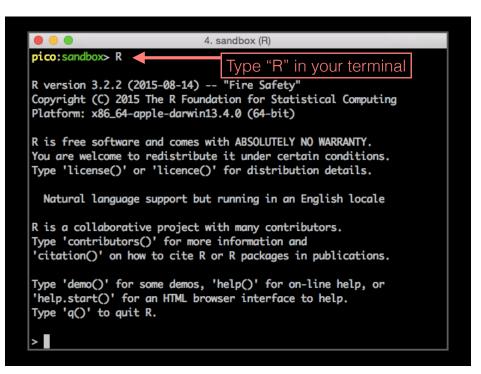
What is R?

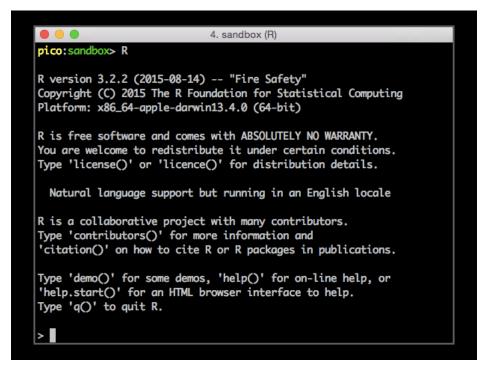
R is a freely distributed and widely used programing language and environment for statistical computing, data analysis and graphics.

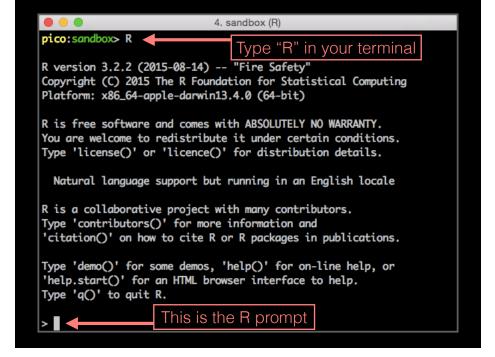


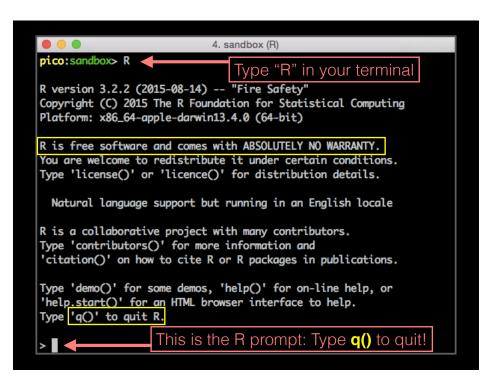
R provides an unparalleled interactive environment for data analysis.

It is script-based (*i.e.* driven by computer code) and not GUI-based (point and click with menus).









What R is **NOT**

A performance optimized software library for incorporation into your own C/C++ etc. programs.

A molecular graphics program with a slick GUI.

Backed by a commercial guarantee or license.

Microsoft Excel!

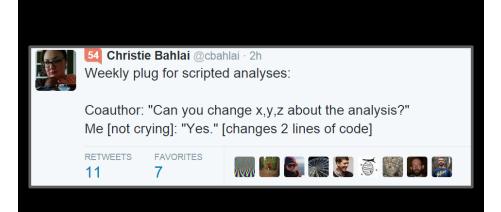
What about Excel?

- Data manipulation is easy
- · Can see what is happening
- But: graphics are poor
- Looping is hard
- Limited statistical capabilities
- Inflexible and irreproducible



Use the right tool!

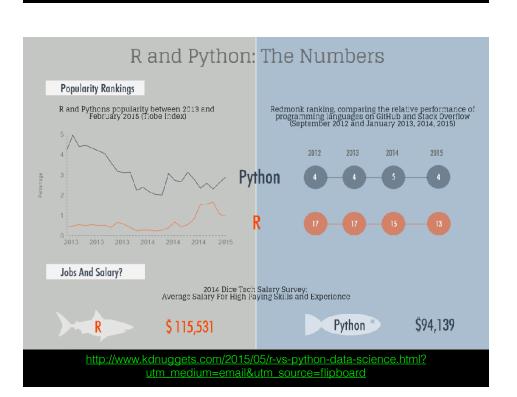
• There are many many things Excel just cannot do!



Rule of thumb: Every analysis you do on a dataset will have to be redone 10–15 times before publication. Plan accordingly!

Why use R?

Productivity
Flexibility
Designed for data analysis



IEEE 2016 Top Programming Languages Language Rank Types Spectrum Ranking 1. C] T 100.0 2. Java $\bigoplus \square \neg$ 98.1 3. Python 98.0 4. C++ 95.9 5. R 87.9 \bigoplus \square \neg 6. C# 86.7 **7.** PHP 82.8 8. JavaScript ₩ 🗆 82.2 9. Ruby 74.5 71.9 10. Go http://spectrum.ieee.org/computing/software/the-2016-top-programming-languages

- R is the "lingua franca" of data science in industry and academia.
- Large user and developer community.
 - As of Aug 1st 2016 there are 8811 add on R packages on <u>CRAN</u> and 1211 on Bioconductor - more on these later!
- Virtually every statistical technique is either already built into R, or available as a free package.
- Unparalleled exploratory data analysis environment.

Modularity	Core R functions are modular and work well with others
Interactivity	R offers an unparalleled exploratory data analysis environment
Infrastructure	Access to existing tools and cutting- edge statistical and graphical methods
Support	Extensive documentation and tutorials available online for R
R Philosophy	Encourages open standards and reproducibility

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Modularity

R was designed to allow users to interactively build complex workflows by interfacing smaller 'modular' functions together.



An alternative approach is to write a **single complex program** that takes raw data as input, and after hours of data processing, outputs publication figures and a final table of results.

All-in-one custom 'Monster' program



Another common approach to bioinformatics data analysis is to write individual scripts in Perl/ Python/Awk/C etc. to carry out each subsequent step of an analysis





This can offer many advantages but can be challenging to make robustly modular and interactive.

Interactivity & exploratory data analysis

Learning R will give you the freedom to explore and experiment with your data.

"Data analysis, like experimentation, must be considered as a highly interactive, iterative process, whose actual steps are selected segments of a stubbily branching, tree-like pattern of possible actions". [J. W. Tukey]

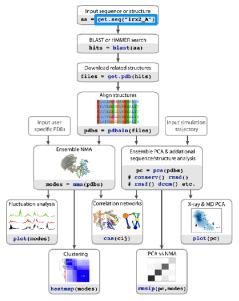
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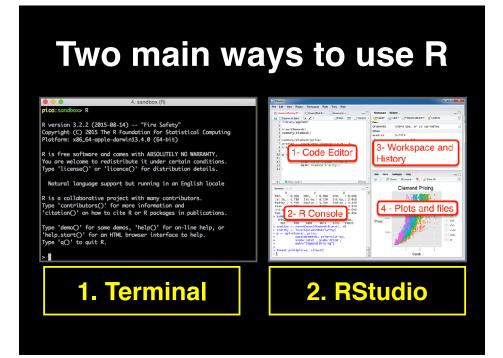
Bioinformatics data is intrinsically **high dimensional** and frequently 'messy' requiring **exploratory data analysis** to find patterns - both those that indicate interesting biological signals or suggest potential problems.

Input sequence or structure aa = get.seq("1rx2_A") BLAST or HMMER search hits = blast(aa) Download related structures files = get.pdb(hits) Align structure: specific PDRs pdbs = pdbaln(files) trajectory Ensemble PCA & addational sequence/structure analysis pc = pca(pdbs) # conserv() rmsd() modes = nma(pdbs) # rmsf() dccm() etc. Fluctuation analysis X-ray & MD PCA TV - Will plot(modes) cna(cij) plot(pc) PCA vs NMA Clusterina rmsip(pc,mode:

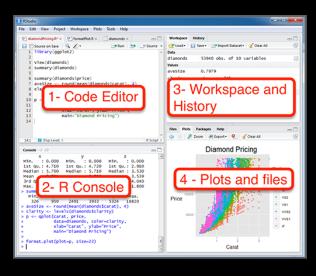
R Features = functions()

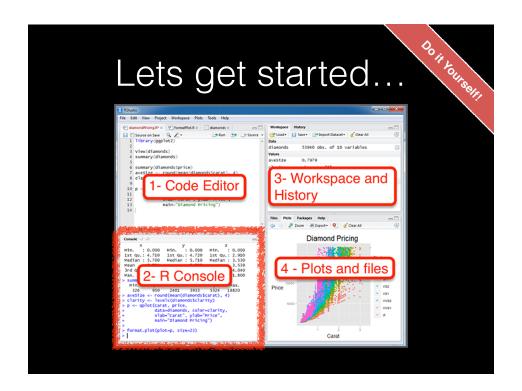


How do we use R?



We will use RStudio today





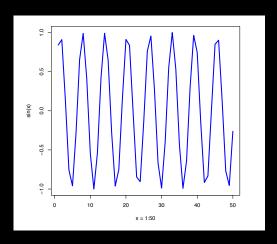
Some simple R commands

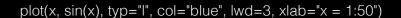
√R prompt!

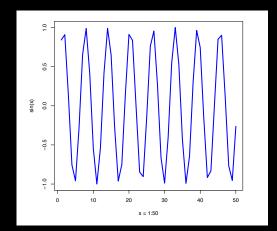
- 1 > 2+2
 [1] 4 Result of the command
- 2 > 3^2 [1] 9
- 3 > sqrt(25) [1] 5
- 4 > 2*(1+1) [1] 4
- $\begin{array}{c|c}
 \hline
 5 & > 2 \times 1 + 1 \\
 \hline
 [1] & 3
 \end{array}$ Order of precedence

- 6 > exp(1) [1] 2.718282
- 7 > log(2.718282) [1] 1
- 8 > log(10, base=10)
 [1] 1 Optional argument
- 9 > log(10
 - + , base = 10)
 [1] 1 Incomplete
- 10 > x=1:50 command > plot(x, sin(x))

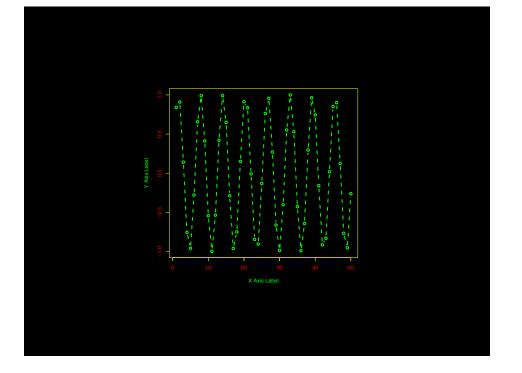
Does your plot look like this?

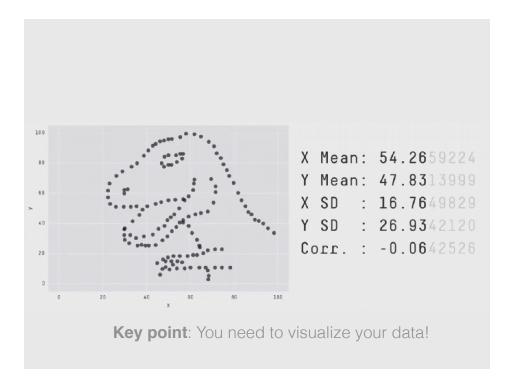






Options: ?plot ?plot.default







Error Messages

Sometimes the commands you enter will generate errors. Common beginner examples include:

• Incomplete brackets or quotes e.g.

((4+8)*20 <enter>

This eturns a + here, which means you need to enter the remaining bracket - R is waiting for you to finish your input.

Press <ESC> to abandon this line if you don't want to fix it.

- Not separating arguments by commas e.g. plot(1:10 col="red")
- Typos including miss-spelling functions and using wrong type of brackets e.g.

exp{4}



Topics Covered:

Calling Functions
Getting help in R
Vectors and vectorization
Workspace and working directory
RStudio projects

Topics Covered:

Calling Functions
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Vectors and vectorization

Workspace and working directory RStudio projects

Vectors

- Vectors are the most basic data structure in R
- All elements of a vector must be the same type

 When you attempt to combine different types they will be <u>coerced</u> to the most flexible type.

```
var <- c(1, "G", "4", 0.05, TRUE)
```

Names

- You can name a vector in several ways:
 - When creating it: x < -c(a = 1, b = 2, c = 3)
 - By modifying an existing vector in place:

 You can then use the names to access (subset) vector elements:

Why is this useful?

Because if you know the name (i.e. your label)
then you don't have to remember which element
of a vector the data you are after was stored in.
Consider this fictional example:

What would happen?

- 1 > x <- 1:3; names(x) <- c("a", "b", "c", "d")
- **2** x <-1:3; names(x) <- 3:1; x[3]
- **3** > x["3"]

R has many data structures

These include:

- vector
- data frame
- list
- matrix
- factors

data.frame

- data.frame is the *de facto* data structure for most tabular data and what we use for statistics and plotting with ggplot2 - more on this later!
- · Arguably the most important R data structure
- Data frames can have additional attributes such as rownames() and colnames(), which can be useful for annotating data, with things like subject_id or sample_id

data.frame continued...

Created with the function data.frame()

```
dat <- data.frame(id = letters[1:10], x = 1:10, y = 11:20)
```

Or more commonly when reading delimited files
 (i.e. importing data) with the functions
 read.csv(), read.table(), read_xlsx() etc...

dep <- read.csv2("http://bio3d.uib.no/data/pdb_deposition2.csv")</pre>

R Studio can do this for you via:
 File > Import Dataset > From CSV...

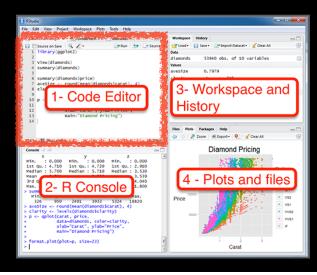
Useful data.frame Functions

- head() -and tail() shows first 6 rows and last 6 rows respectively
- **dim()** returns the dimensions (i.e. number of rows and columns)
- **nrow()** and **ncol()** returns the number of rows and columns separately.
- rownames() and colnames()- shows the names attribute for rows and columns
- **str()** returns the structure including name, type and preview of data in each column

Key Points

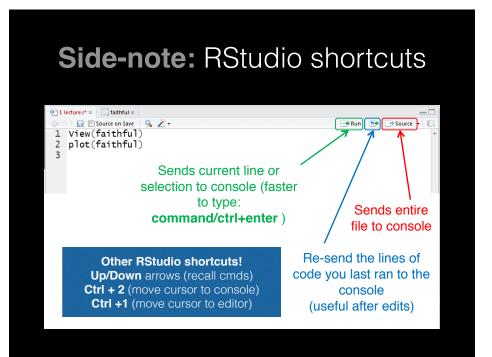
- R's basic data types are **logical**, **character**, **numeric**, integer and complex.
- R's basic data structures include **vectors**, lists, **data frames**, matrices and factors.
- Objects may have attributes, such as **name**, **dim**ension, and **class**.

Side-note: Use the code editor for R scripts



R scripts

- A simple text file with your R commands (e.g. lecture7.r) that contains your R code for one complete analysis
- Scientific method: complete record of your analysis
- Reproducible: rerunning your code is easy for you or someone else
- In RStudio, select code and type <ctrl+enter> to run the code in the R console
- Key point: Save your R script!



Rscript: Third way to use R





> Rscript --vanilla my_analysis.R

1. Terminal

2. RStudio

3. Rscript

From the command line!

> Rscript --vanilla my_analysis.R
or within R: source(my_analysis.R)

Side-Note: R workspaces

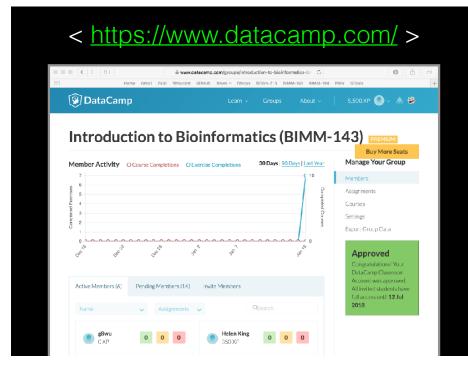
- When you close RStudio, SAVE YOUR .R SCRIPT
- You can also save data and variables in an R workspace, but this is generally not recommended
- Exception: working with an enormous dataset
- Better to start with a clean, empty workspace so that past analyses don't interfere with current analyses
- rm(list = ls()) clears out your workspace
- You should be able to reproduce everything from your R script, so save your R script, don't save your workspace!

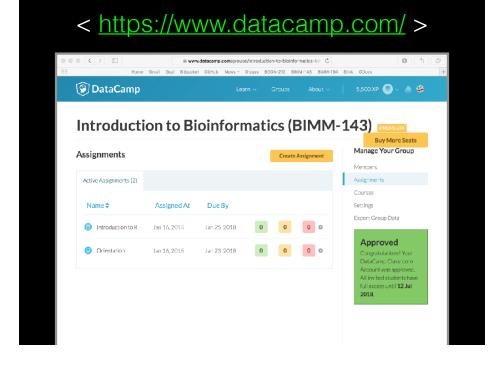
Learning Resources

- TryR. An excellent interactive online R tutorial for beginners. < http://tryr.codeschool.com/ >
- RStudio. A well designed reference card for RStudio.
 - < https://help.github.com/categories/bootcamp/ >
- DataCamp. Online tutorials using R in your browser.
 - < https://www.datacamp.com/ >
- R for Data Science. A new O'Reilly book that will teach you how to do data science with R, by Garrett Grolemund and Hadley Wickham.
 - < http://r4ds.had.co.nz/ >

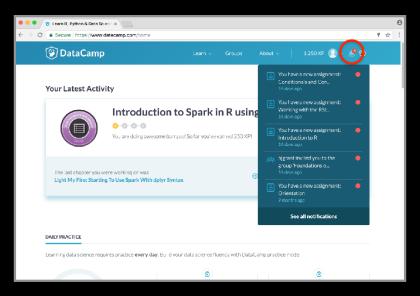
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 - < https://www.datacamp.com/ >
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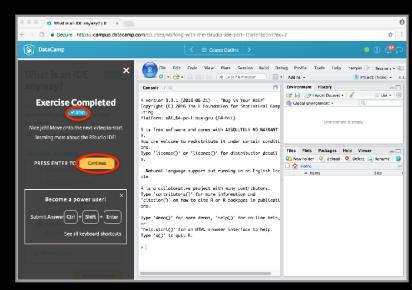




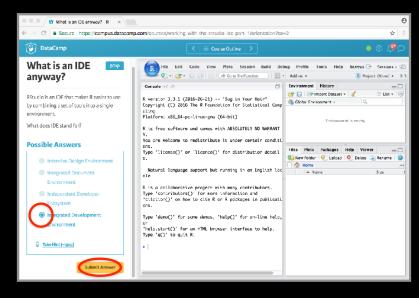
< https://www.datacamp.com/ >



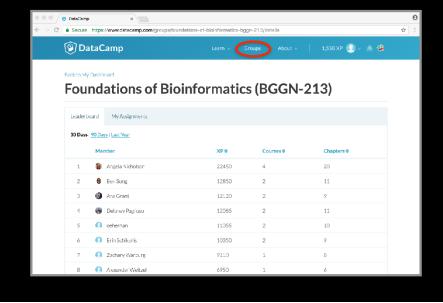
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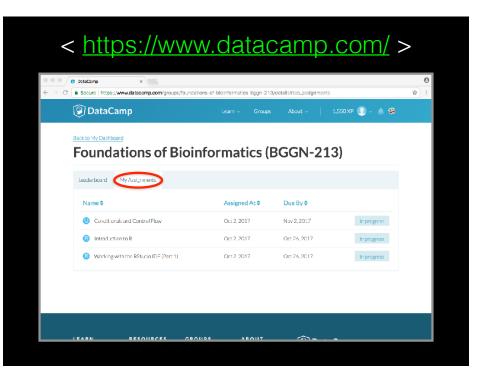


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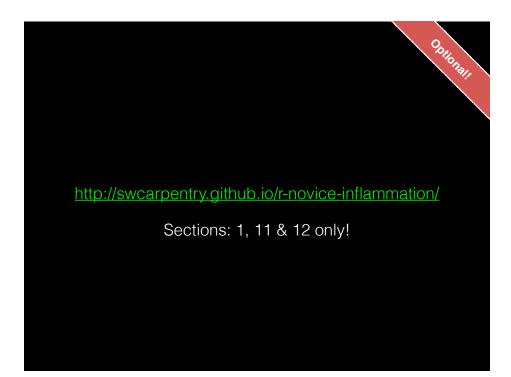
Key Points

- R's basic data types are **logical**, **character**, **numeric**, integer and complex.
- R's basic data structures include **vectors**, lists, **data frames**, matrices and factors.
- Objects may have attributes, such as name, dimension, and class.
- DataCamp, StackOverflow and help() are your friends.

Final Knowledge Check!

- What is R and why should we use it?
- Familiarity with R's basic syntax.
- Familiarity with major R data structures namely *vectors* and *data.frames* (with more on *lists* and *matrices* next day).
- Understand the basics of using functions (arguments, vectorizion and re-cycling).
- Be able to use R to read and parse comma-separated (.csv) formatted files ready for subsequent analysis.
- · Appreciate how you can use R scripts to aid with reproducibility.

Link: Muddy point assessment



R: Logarithms and Exponentials • Find in Top R Documentation log (base)

Logarithms and Exponentials

 \log computes legarithms, by default natural logarithms, $\log 10$ computes common (i.e., bood 10) logarithms, and $\log 2$ computes binary (i.e., base 2) logarithms. The general form $\log (x)$ base) computes logarithms with base base

logic (x) computes log(1+x) accurately also for |x| << 1 (and less accurately when x /s

exp computes the exponential function.

 $= x_1 m1$ (x) computes exp(x) - f accurately also for |x| << 1

Usage How to use the function

log(x, base = exp(1))
logb(x, sase = exp(1) logiu(x) 1002 (x)

loglp(x

Arguments What does the function need

- x a numeric or complex vector.
- base a positive or complex number: the base with respect to which locarithms are computed. Dafaults to e=exp(1).

All except logb are generic functions; methods can be defined for them individually or via the Math group generic.

10210 and 1022 are only convenience wrappers, but logs to bases 10 and 2 (whether computed via lag or the wrappers) will be computed more efficiently and accurately where supported by the OS. Methods can be set for them individually (and otherwise methods for log

Logic is a wrapper for Log for compatibility with S. If (S3 or S4) methods are set for Log they

All except log are <u>primitive</u> functions.

R: Logarithms and Exponentials =

Value What does the function return

Description What the function does in general terms Avector of the same length of a containing the transformed values. Log (0) gives Inf., and log(x) for negative values of x is NaN. exp(-Inf) is 0.

> For complex inputs to the loc functions, the value is a complex number with imaginary part in the range [-pi, pi]: which end of the range is used might be platform-specific

exp, expm1, log, log10, log2 and log1p are S4 generic and are members of the $\underline{\mathtt{Math}}$

Note that this means that the S4 generic for tog has a signature with only one argument, \mathbf{x} but that have can be passed to methods (but will not be used for method selection). On the other hand, if you only set a method for the Math group generic then have argument of log will be ignored for your class.

Source

log1p and expm1 may be taken from the operating system, but if not available there are based on the Fortran subroutine dlared by W. Fullerton of Los Alamos Scientific Laboratory (see http://www.netib.org/slotte/finib/direct f and (for small x) a angle Newton step for the solution of Legip $\{y\} = x$ respectively.

Becker, R. A., Chambers, J. M. and Wilks, A. R. (1988) The New S Language. Wadsworth & Brooks/Cole. (for log, log10 and exp.)

Chambers, J. M. (1998) Programming with Data. A Guide to the S Languege. Springer. (for

See Also Discover other related functions

[Examples] Sample code showing how it works

v <- 100-(140*1:0) cbind(x, log(1+x), log(p(x), exp(x)-1, expml(x))

[Fackage base version 3.0.1 Index]

Help from within R

- Getting help for a function
- > help("log")
- > ?loa
- Searching across packages
- > help.search("logarithm")
- · Finding all functions of a particular type

```
> apropos("log")
[7] "SSlogis" "as.data.frame.logical" "as.logical"
  "as.logical.factor" "dlogis" "is.logical"
[13] "log" "log10" "log1p" "log2" "logLik" "logb"
[19] "logical" "loglin" "plogis" "print.logLik" "glogis"
  "rlogis"
```

Optional Exercise

Use R to do the following. Create a new script to save your work and code up the following four equations:

$$1 + 2(3 + 4)$$

$$\ln(4^{3} + 3^{2+1})$$

$$\sqrt{(4+3)(2+1)}$$

$$\left(\frac{1+2}{3+4}\right)^{2}$$