

Computational Skills for Biostatistics I: Lecture 1

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Welcome!

- ▶ Biost 561 is an add-on to Biost 514
- ▶ 514 covers basic R commands, 561 covers more advanced R and other programming skills
- ▶ As a MS/PhD student in Biostats, you will do some serious programming!
- ▶ Good programming practices will help you with research, collaborating, your job search, and your long-term career... whichever path you choose!

Structure and expectations

- ▶ Weekly lectures
- ▶ Weekly homeworks
- ▶ Weekly office hours: HSB F-657. Tuesday 3-4?
- ▶ Office hours by appointment

Public folder of course materials at

<https://github.com/adw96/biostat561>

Topics

Subject to change

- ▶ 9/28 Lecture 1: Advanced classes, methods, objects, debugging
- ▶ 10/5 Lecture 2: Efficient loops, functions
- ▶ 10/12 Lecture 3: Pipes
- ▶ 10/19 Lecture 4: ggplot
- ▶ 10/26 Lecture 5: latex *
- ▶ 11/2 Lecture 6: Markdown *
- ▶ 11/9 Lecture 7: unix, shell, UW cluster computing *
- ▶ 11/16 Lecture 8: version control, git, github, packages
- ▶ 11/23 no class; Thanksgiving
- ▶ 11/30 Lecture 9: Calling C/C++ in R *
- ▶ 12/7 Lecture 10: Python *

* indicates guest lecture by one of your classmates. A great opportunity to learn the latest and greatest!

Resources

- ▶ Available via github
<https://github.com/adw96/biostat561>
 - ▶ Syllabus
 - ▶ Slides (& source code)
 - ▶ Examples
 - ▶ Homeworks
- ▶ Available via github classroom
 - ▶ Homework submission
- ▶ Available via email
 - ▶ Announcements

Expectations

What you should expect of me

- ▶ I will make your learning a priority
- ▶ I will give you timely feedback on your homeworks
- ▶ I will treat you as adults, I will treat you with respect
- ▶ I will talk slowly (tell me if I'm speaking too fast!)
- ▶ I will try to make class engaging and fun!

Expectations

What I will expect of you

- ▶ You make attending class a priority
- ▶ You submit your best work for homework: your own work, on time
- ▶ You engage in classroom discussion and quizzes!
- ▶ You learn from the class; you learn to teach yourself programming skills
- ▶ You treat me, guest lecturers, and each other with respect

Assessment

The only assessment in this course is homework.

- ▶ 10 or fewer homeworks
- ▶ You must submit a good attempt at every homework to receive credit for this course

I won't record attendance, but if you consistently do not show up you will not receive credit.

About me

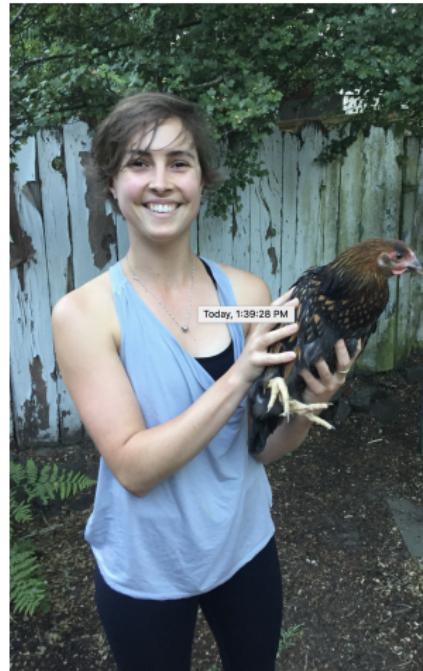
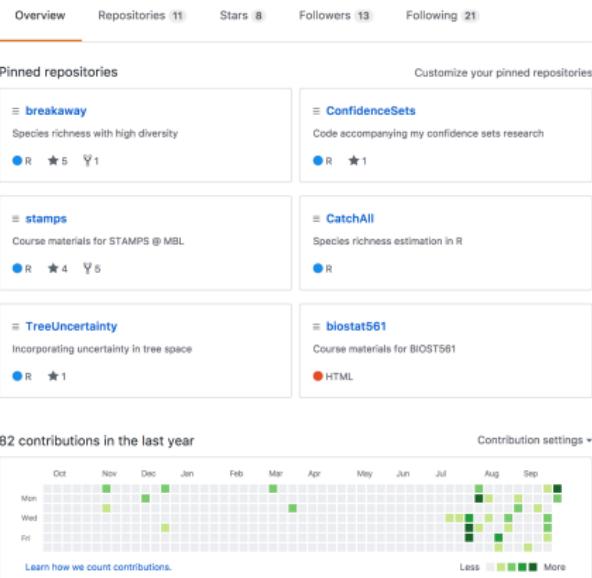


AmyW
adw96

statistician, biodiversity
enthusiast, assistant professor

📍 Seattle, WA
➲ <http://faculty.washington.edu...>

Organizations



About you

Everyone is here with different backgrounds in programming and computing. Let's get statistical!

[Analyse survey here]

Class

There are many different *classes* of objects in R

```
x <- c(1, 2, 5)
y <- c("a", "b")
z <- as.factor(y)
c(class(x), class(y), class(z), class(c))
```

```
## [1] "numeric"    "character"   "factor"      "function"
```

Others include logical (TRUE, FALSE), complex numbers . . .

Modes

R has different modes. The mode tells the way a variable is stored.

```
mode(x)
```

```
## [1] "numeric"
```

```
mode(y)
```

```
## [1] "character"
```

```
mode(z) # factors are stored as numerics
```

```
## [1] "numeric"
```

```
mode(c)
```

```
## [1] "function"
```

Modes

`is.[class]` asks about the class. Normally you will be interested in the class, not the mode.

```
is.numeric(x)
```

```
## [1] TRUE
```

```
is.factor(z)
```

```
## [1] TRUE
```

```
is.numeric(z) # we asked about class, not the mode
```

```
## [1] FALSE
```

Data structures

R can store data in various *objects*

- ▶ vector: one-dimensional, all data points have same mode
- ▶ matrix: two-dimensional, all data points have same mode
- ▶ data frame: two-dimensional, all data points in same column have same mode
- ▶ list: one-dimensional, data points can be of any type

Matrices

Matrices vs data frames: all elements have same mode in matrices

```
cbind(c(1,2), c("a", "b"))
```

```
##      [,1] [,2]
## [1,] "1"  "a"
## [2,] "2"  "b"
```

Matrices

Be very careful with matrix operations!

```
aa <- matrix(c(1, 2, 3, 5), nrow = 2, byrow = T)
bb <- c(0.5, 2)
aa
```

```
##      [,1] [,2]
## [1,]     1     2
## [2,]     3     5
```

```
bb
```

```
## [1] 0.5 2.0
```

Matrices

```
aa %*% bb # matrix multiplication
```

```
##      [,1]
## [1,] 4.5
## [2,] 11.5
```

```
aa * bb # careful! pointwise
```

```
##      [,1] [,2]
## [1,] 0.5   1
## [2,] 6.0   10
```

Data frames

```
data.frame(c(1,2), c("a", "b"))
```

```
##   c.1..2. c..a....b..
## 1           1           a
## 2           2           b
```

```
dd <- data.frame("ID"=c(1,2), "name"=c("a", "b")) # better
dd
```

```
##   ID name
## 1  1    a
## 2  2    b
```

```
str(dd) # structure: compact info about frame & variables
```

```
## 'data.frame': 2 obs. of 2 variables:
## $ ID : num 1 2
## $ name: Factor w/ 2 levels "a","b": 1 2
```

vectorisation

vectorisation: doing many calculations with a single command

```
x <- c(0.5, 2, 3, 6)  
x^2
```

```
## [1] 0.25 4.00 9.00 36.00
```

```
y <- c(3, 1, 2, 1)  
x^y
```

```
## [1] 0.125 2.000 9.000 6.000
```

vectorisation

The slow way: with loops. Avoid where possible!

```
z <- rep(NA, 4)
for (i in 1:4) {
  z[i] <- x[i]^y[i]
}
z
```

Code is easier to read, and usually easier to debug, when vectorised

recycling

In many tasks, R recycles elements of one input until it has enough to match the other

```
aa
```

```
##      [,1] [,2]
## [1,]     1     2
## [2,]     3     5
```

```
bb
```

```
## [1] 0.5 2.0
```

```
aa + bb # vectors are treated as columns!
```

```
##      [,1] [,2]
## [1,]    1.5   2.5
## [2,]    5.0   7.0
```

```
cc <- c(1, 2, 3, 4)
aa + cc # the silent killer
```

Speed comparison

Vectorisation can cause major speed-ups, because task is optimised and precompiled in C/Fortran, not interpreted R.

```
dd <- matrix(rnorm(1e6), nrow = 1000)
cor(dd)

ee <- matrix(NA, 1000, 1000)
for (i in 1:1000) {
  for (j in 1:1000) {
    ee[i, j] <- cor(ee[, i], ee[, j])
  }
}
```

Speed up factor of vectorisation: 36!

Lists

Lists store information of many different types. Names are optional, but recommended!

```
amy <- list(office.num = 657, pets = TRUE,
            pets.names = c("Princess Jaws", "Friendly", "Mohawk",
                           "Canada", "USA", "Regina George"),
            is.cat = c(TRUE, rep(FALSE, 5)))
```

```
amy
```

```
## $office.num
## [1] 657
##
## $pets
## [1] TRUE
##
## $pets.names
## [1] "Princess Jaws" "Friendly"      "Mohawk"       "Canada"
## [5] "USA"           "Regina George"
##
## $is.cat
## [1] TRUE FALSE FALSE FALSE FALSE
```

Lists

Double square brackets pull out individual elements. Single square brackets pull out subsets of the list. I recommend using names wherever possible!

```
amy[[3]] # subset third element
```

```
## [1] "Princess Jaws" "Friendly"      "Mohawk"       "Canada"  
## [5] "USA"           "Regina George"
```

```
amy[3] # third element -- a list!
```

```
## $pets.names  
## [1] "Princess Jaws" "Friendly"      "Mohawk"       "Canada"  
## [5] "USA"           "Regina George"
```

Lists

```
amy[2:3] # second and third elements -- a list!
```

```
## $pets
## [1] TRUE
##
## $pets.names
## [1] "Princess Jaws" "Friendly"      "Mohawk"       "Canada"
## [5] "USA"           "Regina George"
```

```
amy$office # can also refer by name
```

```
## [1] 657
```

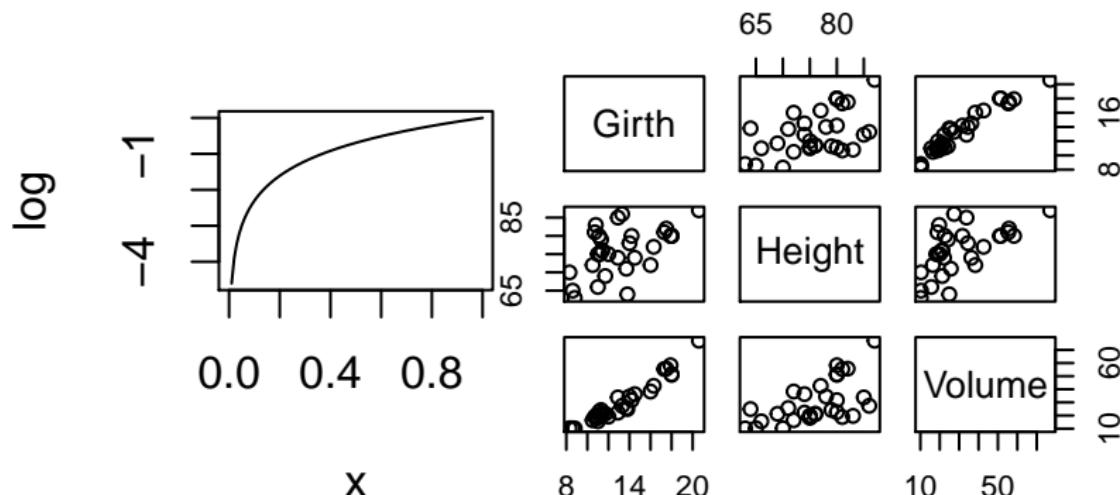
Generic functions

The same function can apply to objects of different classes. How does R know what to do?

```
c(class(log), class(trees))
```

```
## [1] "function"    "data.frame"
```

```
layout(t(1:2), widths = c(3,1)); plot(log); plot(trees)
```



Generic functions

plot is a *generic* function. Generic functions don't do anything themselves – they call *methods*, which are tailored to the class.

plot

```
## function (x, y, ...)
## UseMethod("plot")
## <bytecode: 0x7f94433b7b50>
## <environment: namespace:graphics>
```

```
methods("plot") # lists all types R knows how to plot
```

```
## [1] plot.acf*           plot.data.frame*   plot.decomp*
## [4] plot.default         plot.dendrogram*  plot.density*
## [7] plot.ecdf            plot.factor*      plot.formula*
## [10] plot.function        plot.hclust*      plot.histogram*
## [13] plot.HoltWinters*   plot.isoreg*     plot.lm*
## [16] plot.medpolish*    plot.mlm*       plot.ppr*
## [19] plot.pie*             plot.princomp*  plot.spline*
```

Generic functions

To find the functions that apply to a class

```
methods(class = "lm")
```

```
## [1] add1           alias          anova         case.numeric  
## [5] coerce         confint        cooks.distance deviance  
## [9] dfbeta         dfbetas        drop1         dummy  
## [13] effects        extractAIC   family        formula  
## [17] hatvalues      influence     initialize    kappa  
## [21] labels         logLik        model.frame  model  
## [25] nobs           plot          predict       print  
## [29] proj            qr            residuals    rstudent  
## [33] rstudent        show          simulate    slotsK  
## [37] summary         variable.names vcov  
## see '?methods' for accessing help and source code
```

Generic functions

To see the code for a generic function, type [function].[class]

```
dimnames.data.frame
```

```
## function (x)
## list(row.names(x), names(x))
## <bytecode: 0x7f94460a20e8>
## <environment: namespace:base>
```

Help

There are many ways to get help with using functions or debugging code

1. The internet

The screenshot shows a Google search results page. The search query is "r How do I change the color of points". The results are filtered under the "All" tab. The first result is a link to Stack Overflow titled "r - Setting the color for an individual data point - Stack Overflow". The second result is "R color scatter plot points based on values - Stack Overflow". The third result is "change the color of certain data points in r - Stack Overflow". Below the search bar, there are links for "Videos", "Shopping", "Maps", "News", and "More", along with "Settings" and "Tools" buttons.

About 3,360,000 results (1.24 seconds)

[r - Setting the color for an individual data point - Stack Overflow](https://stackoverflow.com/questions/.../setting-the-color-for-an-individual-data-point)

<https://stackoverflow.com/questions/.../setting-the-color-for-an-individual-data-point> ▾

Jan 7, 2012 - To expand on @Dirk Eddelbuettel's answer, you can use any function for col in the call to plot . For instance, this colors the x==3 point red, ...

[R color scatter plot points based on values - Stack Overflow](https://stackoverflow.com/questions/.../r-color-scatter-plot-points-based-on-values)

<https://stackoverflow.com/questions/.../r-color-scatter-plot-points-based-on-values> ▾

Jul 9, 2013 - Best thing to do here is to add a column to the data object to represent the point colour. Then update sections of it by filtering.

[change the color of certain data points in r - Stack Overflow](https://stackoverflow.com/questions/.../change-the-color-of-certain-data-points-in-r)

<https://stackoverflow.com/questions/.../change-the-color-of-certain-data-points-in-r> ▾

Nov 17, 2013 - The following will work given that your data is in a data.frame called "dat".
cols <- rep('black', nrow(dat)) cols[c(7, 8, 15)] <- 'red'. In your plot ...

[Quick-R: Graphical Parameters](http://www.statmethods.net/advgraphs/parameters.html)

www.statmethods.net/advgraphs/parameters.html ▾

Help

2. `?fn` shows the documentation for `fn...` if it exists!



Generic X-Y Plotting

Description

Generic function for plotting of `R` objects. For more details about the graphical parameter arguments, see [par](#).

For simple scatter plots, [plot.default](#) will be used. However, there are `plot` methods for many `R` objects, including [functions](#), [data.frames](#), [density](#) objects, etc. Use `methods(plot)` and the documentation for these.

Usage

```
plot(x, y, ...)
```

Arguments

- x the coordinates of points in the plot. Alternatively, a single plotting structure, function or *any R object with a plot method* can be provided.
- y the y coordinates of points in the plot, *optional* if x is an appropriate structure.
- ... Arguments to be passed to methods, such as [graphical parameters](#) (see [par](#)). Many methods will accept the following arguments:

Help

2. ... if it exists!

| Secure | <https://adw96.github.io/breakaway/reference/simpson.html>

breakaway



Get Started

Reference

Articles ▾

News

Plug-in Simpson

TODO

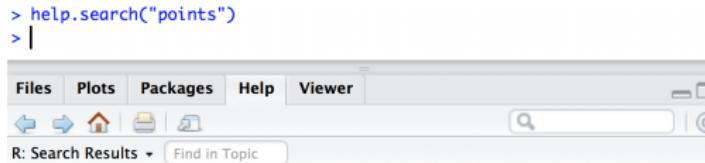
```
simpson(input)
```

Arguments

input TODO

Help

3. `help.search("topic")` searches help pages for “topic”



Search Results



Code demonstrations:

[tcltk::tkcanvas](#) Creates a canvas widget showing a 2-D plot with data points that can be dragged with the mouse. [\(Run demo in console\)](#).

Help pages:

ade4::ichtyo	Point sampling of fish community
ade4::procuste	Simple Procrustes Rotation between two sets of points
ade4::s.class	Plot of factorial maps with representation of point classes
ade4::triangle.class	Triangular Representation and Groups of points
ade4::triangle.plot	Triangular Plotting
base::pretty	Pretty Breakpoints
base::utf8ToInt	Convert Integer Vectors to or from UTF-8-encoded

Examples

The documentation pages often show examples (`example(plot)`) and have demos (`demo(plotmath)`). `vignette()` opens longer worked examples that are great for playing with new packages.



Secure | <https://adw96.github.io/breakaway/articles/betta-figure.html>

breakaway Home Get Started Reference Articles News

Comparing samples visually with betta

Amy Willis

2017-09-22

`betta` is useful for formally testing differences between communities with respect to their alpha diversity. However, before ever doing any inferential test, it's best to try to visualise the difference that you are looking for. Here is a example to show you how to do that using `betta_pic`.

```
library(breakaway)

frequencytablelist <- lapply(apply(toy_otu_table, 2, table), as.data.frame)
frequencytablelist <- lapply(frequencytablelist, function(x) x[x[,1]!=0,])

ob_results <- lapply(frequencytablelist[1:15], objective_bayes_negbin, answers = T, plot=F, print = F)

lower <- unlist(lapply(ob_results, function(x) x$results["LCI.C", ]))
upper <- unlist(lapply(ob_results, function(x) x$results["UCI.C", ]))
means <- unlist(lapply(ob_results, function(x) x$results["mean.C", ]))
standard_deviations <- unlist(lapply(ob_results, function(x) x$results["stddev.C", ]))

## Find how many otus are in the cyanobacteria genus
cyano_nodes <- apply(toy_otu_table[grepl("Cyano", toy_taxonomy), ], 2, function(x) sum(x>0))
cyano_nodes <- cyano_nodes

bloom1 <- toy_metadata[, "bloom2"]
bloom <- bloom1 == "yes"
```

The following is the default option for plotting. It shows the mean estimates with lines up to +/- 2 standard deviations. In this way, it is a visual mimic of the procedure that underpins `betta`. Feel free to pull apart the source code to optimise it for your purposes. It uses

Debugging

1. Stare at it until you identify the problem a.k.a. psychic debugging
2. Breakdown the components until you find the problem (bisection method converges linearly!)
3. `traceback()` – covered later in the course

Homework 1 and next week

- ▶ Slides:
 - ▶ <https://github.com/adw96/biostat561/lecture1/lecture1.pdf>
- ▶ Homework 1 is due next Thursday at 2 p.m.
 - ▶ <https://github.com/adw96/biostat561/lecture1/homework1.pdf>
- ▶ Complete Question 0 by next Tuesday so you can attend office hours if you have trouble
- ▶ Submission via github classroom (instructions included):
 - ▶ <https://classroom.github.com/classrooms/32249780-biost-561>