# Phylogenetics in R

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# 1. Introduction and resources

This practical introduces you to basic phylogenetic computing in R. We will review importing phylogenetic trees as data files, displaying phylogenetic trees visually, and some basic evolutionary computations that can be conducted with phylogenetic trees. This practical will deliver some of the important background for Coursework 1. Below you will find some of the relevant resources required for this practical.

Parts (sections 2,3,4 and 5) of this practical are written by Natalie Cooper. The original can be found here.

The data used throughout the practical can be downloaded from Blackboard or from here and here (right click and save link as).

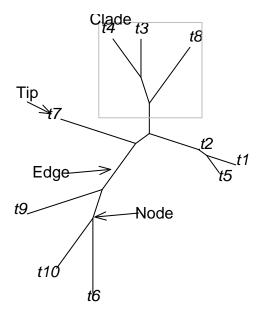
Further information can be found in Liam Revell's book chapter here.

# 2. A refresher of phylogenetic trees

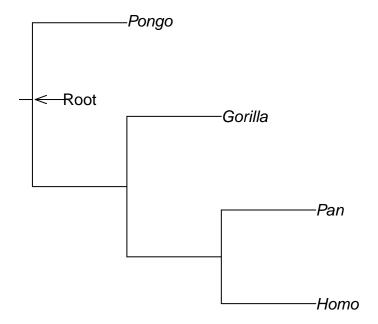
This section will review some basic aspects of phylogenetic trees and introduce how trees are handled at the level of software. Because you are now interacting with phylogenetic trees at a 'lower level' (i.e. at the bioinformatics level), it is also helpful to know some of the names for parts of phylogenetic trees used in computer science.

## A. Tree parameters

A phylogenetic tree is an ordered, multifurcating graph with labeled **tips** (or **leaves**) (and sometimes labeled histories). It represents the relative degrees of relationships of species (i.e. tips or OTUs). The graph consists of a series of **branches** (or **edges**) with join successively towards **nodes** (or **vertices**, *sing.* **vertex**). Each node is subtended by a single branch, representing the lineage of ancestors leading to a node. The node is thus the common ancestor of two or more descendant branches. All the descendant branches of a given node (and all of the their respective descendants) are said to form a **clade** (or **monophyletic group**).

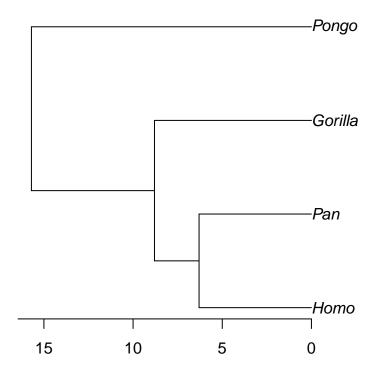


When we select a node to act as the base of a tree, the tree is said to be **rooted**. At the bottom of a tree, is the **root node** (or simply the **root**).



Phylogenetic trees of the kind shown above are fairly simple and lack information about time or character changes occurring along a branch. We can assign branch length in the form of either time or the amount of change/substitution along a branch. A tree with **branch lengths** depicted can be called a **phylogram**.

When (an implied) dimension of time is being considered, all the tips of the tree must be at the level representing the time in which they are observed. For trees where all the species are extant, the tips are flush at the top. This representation is called an **ultrametric** tree.



#### B. Informatic representations of tree

To perform any useful calculations on a tree, we need both a computer-readable tree format and (in part) to understand how trees are constructed in computer memory.

#### Text based formats

Storage of trees for transfer between different software is essential. This is most commonly achieved with a text-based format stored in a file. The most common file format for representing phylogenetic trees is **Newick format**. This consists of clades represented within parentheses. Commas separate each clade. Either tip names or symbols representing the tips are nested within the lowest orders of parentheses. Each tip or branch can be associated with a branch length scalar that follows a colon.

For example:

```
"(((Homo, Pan), Gorilla), Pongo);"
```

Or with branch length:

```
"(((Homo:6.3, Pan:6.3):2.5, Gorilla:8.8):6.9, Pongo:15.7);"
```

Trees are also increasing use of XML formats such as PhyloXML and NeXML.

In this practical we are going to use the elopomorph.tre newick tree. You can open it with a simple text editor to see the newick tree structure.

#### Edge table

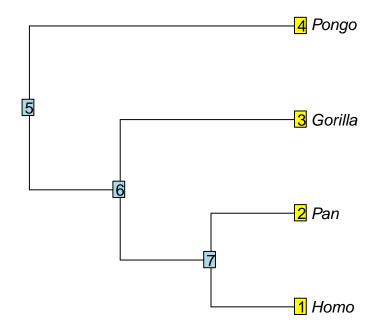
It is also possible to represent a phylogenetic tree as a matrix of edges and vertices called an edge table. This is an even less intuitive representation, but it is implemented in R and worth reviewing here.

There are a number of conventions that can be used to create an edge table. The general concept consists of numbering the tips 1 - n, and all internal nodes labeled  $n+1 \dots n+n-1$ . The numbers for the internal nodes can be assigned arbitrarily or according to an algorithm.

In R packages like ape, edge tables are constructed as follows:

node	connects to
5	6
6	7
7	1
7	2
6	3
5	4

You read the table as follows: node 5 (root) connects to node 6. The node 6 connects to node 7. Node 7 connects to node 1 that happen to be the first tip (Homo) and to node 2 (Pan) etc... Note that in a binary tree (i.e. a tree where each node has only two descendants) each node always connects to two elements (nodes or tips).



## Records & pointers

At a lower level, phylogenetic trees can be represented in computer memory as more complex data objects. We don't need to go into detail here, but if you consider nodes and tips as data objects (i.e. a dataframe), a tree could be stored as an array of dataframes which store information about which store information about which members of that same array are descendants and which are ancestors.

# 2. Installing and loading extra packages in R

To plot phylogenies (or use any specialized analysis) in R, you need to download one or more additional packages from the basic R installation. For this practical you will need to install the following packages:

- ape
- phytools

To install the package ape:

### install.packages("ape")

Pick the closest mirror to you if asked. Now install phytools.

You've installed the packages but they don't automatically get loaded into your R session. Instead you need to tell R to load them **every time** you start a new R session and want to use functions from these packages. To load the package ape into your current R session:

## library(ape)

You can think of install.packages like installing an app from the App Store on your smart phone - you only do this once - and library as being like pushing the app button on your phone - you do this every time you want to use the app.

Don't forget to load phytools too!

```
library(phytools)
```

Note that the latest version of ape (3.5) now requires you to have R v3.2.0 installed (or higher). If you are using an older version of R you can always install an older version of ape as follows:

```
old_ape <- "https://cran.r-project.org/src/contrib/Archive/ape/ape_3.4.tar.gz"
install.packages(old_ape, repos = NULL, type = "source")</pre>
```

Or change the version number to something even lower (e.g. change ape\_3.4.tar.gz to ape\_3.0.tar.gz). However, remember that this is a bad practice in informatics and that you should always prefer the most up to date (and corrected!) version.

# 3. Loading your phylogeny and data into R

#### Reading in a phylogeny from a file

To load a tree you need the function read.tree. read.tree can read any newick format trees (see above) like the elopomorph.tre file.

```
fishtree <- read.tree("elopomorph.tre")</pre>
```

Be sure you are always in the right directory. Remember you can navigate in R using setwd(), getwd() and list.files() (to see what's in the current directory).

## Reading in a phylogeny that is already built into R

The bird and anole phylogenies are already built into R so we don't need to read them in using read.tree. Instead we just use:

```
data(bird.orders)
data(anoletree)
```

## Reading and viewing your data in R

Later we will use some Greater Antillean *Anolis* lizard data to add data to a phylogeny. Before we can add data to our tree, we need to load the data we are going to use. R can read files in lots of formats, including comma-delimited and tab-delimited files. Excel (and many other applications) can output files in this format (it's an option in the Save As dialogue box under the File menu). To save time I have given you a comma-delimited text file called anole.data.csv which we are going to use. Load these data as follows. I am assuming you have set your working directory, if not don't forget the path.

```
anoledata <- read.csv("anole.data.csv", header = TRUE)</pre>
```

You can use read.delim for tab delimited files or read.csv for comma delimited files (comma separated values). header = TRUE, indicates that the first line of the data contains column headings.

This is a good point to note that unless you **tell** R you want to do something, it won't do it automatically. So here if you successfully entered the data, R won't give you any indication that it worked. Instead you need to specifically ask R to look at the data.

We can look at the data by typing:

#### str(anoledata)

```
'data.frame':
                    100 obs. of 23 variables:
##
    $ species
                             : Factor w/ 100 levels "ahli", "alayoni", ...: 1 2 3 4 5 6 7 8 9 10 ...
##
    $ AVG.SVL
                                    4.04 3.82 3.53 4.04 4.38 ...
##
    $ AVG.hl
                                    2.88 2.7 2.38 2.9 3.36 ...
                                    2.36 1.99 1.56 2.37 2.69 ...
##
    $ AVG.hw
                             : num
##
    $ AVG.hh
                             : num
                                    2.13 1.75 1.39 2.05 2.32 ...
##
   $ AVG.1j1
                                    2.85 2.71 2.32 2.9 3.38 ...
                             : num
    $ AVG.outlever
                             : num
                                    2.75 2.62 2.26 2.83 3.29 ...
##
    $ AVG.jugal.to.symphysis: num
                                    2.54 2.37 2.08 2.6 3.07 ...
##
    $ AVG.femur
                             : num
                                    2.74 2.07 2.17 2.48 2.8 ...
##
    $ AVG.tibia
                                    2.69 2.02 2.09 2.34 2.69 ...
                             : num
##
    $ AVG.met
                                    2.25 1.54 1.55 1.87 2.18 ...
                             : num
    $ AVG.ltoe.IV
##
                             : num
                                    2.55 1.88 1.73 2.26 2.53 ...
##
    $ AVG.toe.IV.lam.width : num
                                    0.1795 0.0488 -0.5361 0.4904 0.8441 ...
##
    $ AVG.humerus
                                    2.46 1.95 1.63 2.3 2.62 ...
                             : num
##
    $ AVG.radius
                                    2.27 1.69 1.4 2.09 2.34 ...
                             : num
##
    $ AVG.lfing.IV
                             : num
                                    1.94 1.4 1.04 1.7 1.98 ...
##
    $ AVG.fing.IV.lam.width : num
                                    0.0754 -0.0739 -0.755 0.3155 0.6584 ...
##
    $ AVG.pelv.ht
                                    1.99 1.51 1.19 1.87 2.1 ...
                             : num
    $ AVG.pelv.wd
##
                                    1.71 1.419 0.946 1.752 2.014 ...
                             : num
##
    $ Foot.Lam.num
                                    3.28 3.43 3.2 3.58 3.72 ...
                             : num
##
    $ Hand.Lam.num
                                    2.87 3.08 2.73 3.16 3.24 ...
                             : num
##
    $ Avg.lnSVL2
                             : num
                                    3.94 3.74 3.48 3.93 4.36 ...
##
  $ Avg.ln.t1
                             : num
                                   4.41 4 4.37 4.44 5.04 ...
```

Always look at your data before beginning any analysis to check it read in correctly.

str shows the structure of the data frame (this can be a really useful command when you have a big data file). It also tells you what kind of variables R thinks you have (characters, integers, numeric, factors etc.). Some R functions need the data to be certain kinds of variables so it's useful to check this.

#### head(anoledata)

```
##
      species AVG.SVL
                         AVG.hl
                                   AVG.hw
                                            AVG.hh AVG.ljl AVG.outlever
         ahli 4.039125 2.882657 2.356126 2.131599 2.854745
                                                                 2.754934
## 2
     alayoni 3.815705 2.702116 1.990610 1.751588 2.708966
                                                                 2.617852
      alfaroi 3.526655 2.378156 1.556037 1.390037 2.323857
                                                                 2.263844
## 4 aliniger 4.036557 2.898836 2.366592 2.046660 2.903946
                                                                 2.828555
## 5 allisoni 4.375390 3.358957 2.690339 2.317309 3.381306
                                                                 3.288402
      allogus 4.040138 2.861027 2.351275 2.142107 2.848331
## 6
                                                                 2.750404
##
     AVG.jugal.to.symphysis AVG.femur AVG.tibia AVG.met AVG.ltoe.IV
## 1
                   2.538052
                            2.741378
                                        2.686032 2.254620
                                                              2.553344
                             2.065121
## 2
                   2.371995
                                        2.016402 1.535253
                                                              1.875258
## 3
                   2.077565
                             2.172476
                                        2.094946 1.547563
                                                             1.732540
                   2.596821
## 4
                                        2.344015 1.867176
                             2.475445
                                                             2.256541
                                                              2.533697
## 5
                             2.795145
                                        2.687734 2.178155
                   3.071149
## 6
                   2.539320
                             2.739175
                                       2.684476 2.217314
                                                              2.494900
     AVG.toe.IV.lam.width AVG.humerus AVG.radius AVG.lfing.IV
##
## 1
               0.17953991
                             2.460728
                                         2.268511
                                                      1.943288
               0.04879016
## 2
                             1.947873
                                         1.687093
                                                      1.403029
## 3
              -0.53614343
                             1.628260
                                         1.401183
                                                      1.040277
```

```
## 4
               0.49041881
                              2.298878
                                          2.091864
                                                        1.702199
                                                        1.983412
## 5
               0.84407840
                              2.618551
                                          2.341565
## 6
               0.19625907
                              2.459994
                                          2.267176
                                                        1.919010
     AVG.fing.IV.lam.width AVG.pelv.ht AVG.pelv.wd Foot.Lam.num Hand.Lam.num
##
## 1
                0.07541664
                               1.987646
                                           1.7095849
                                                          3.279600
                                                                        2.866197
## 2
               -0.07391568
                               1.507128
                                           1.4194873
                                                          3.432372
                                                                        3.075269
## 3
               -0.75502258
                               1.189367
                                           0.9458495
                                                          3.198016
                                                                        2.733866
## 4
                 0.31554040
                               1.867949
                                           1.7521520
                                                          3.582875
                                                                        3.156774
## 5
                 0.65838319
                               2.097302
                                           2.0142361
                                                          3.721185
                                                                        3.239211
## 6
                 0.05329130
                               2.008426
                                           1.6875679
                                                          3.335990
                                                                        2.808270
##
     Avg.lnSVL2 Avg.ln.t1
       3.939015
## 1
                 4.406652
## 2
       3.743863
                 4.002788
## 3
       3.478776
                 4.369448
## 4
       3.933181
                 4.441203
## 5
       4.355582
                 5.039851
## 6
       4.028863
                 4.510920
```

This gives you the first few rows of data along with the column headings.

#### names (anoledata)

```
[1] "species"
                                   "AVG.SVL"
##
    [3] "AVG.hl"
##
                                   "AVG.hw"
    [5] "AVG.hh"
                                   "AVG.1j1"
##
    [7] "AVG.outlever"
                                   "AVG.jugal.to.symphysis"
       "AVG.femur"
                                   "AVG.tibia"
##
    [9]
       "AVG.met"
                                   "AVG.ltoe.IV"
##
  [11]
  [13] "AVG.toe.IV.lam.width"
                                   "AVG.humerus"
## [15] "AVG.radius"
                                   "AVG.lfing.IV"
##
  [17]
       "AVG.fing.IV.lam.width"
                                   "AVG.pelv.ht"
  [19] "AVG.pelv.wd"
                                   "Foot.Lam.num"
## [21] "Hand.Lam.num"
                                   "Avg.lnSVL2"
## [23] "Avg.ln.t1"
```

This gives you the names of the columns.

#### anoledata

This will print out all of the data!

# 4. Basic tree viewing in R

Now let's visualise some phylogenies! We'll use the Elopomorpha (eels and similar fishes) tree to start as it is simple.

```
fishtree <- read.tree("elopomorph.tre")</pre>
```

Let's examine the tree by typing:

#### fishtree

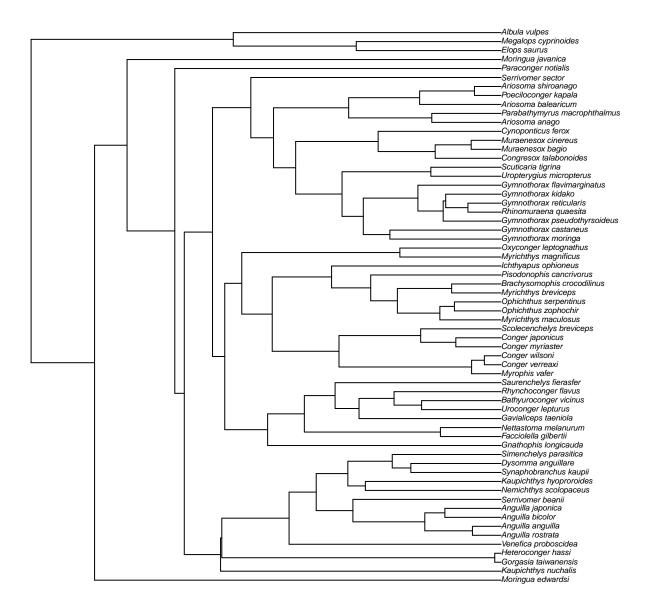
```
##
## Phylogenetic tree with 62 tips and 61 internal nodes.
## Tip labels:
## Moringua_edwardsi, Kaupichthys_nuchalis, Gorgasia_taiwanensis, Heteroconger_hassi, Venefica_probosc
##
## Rooted; includes branch lengths.
```

```
str(fishtree)
```

```
## List of 4
                : int [1:122, 1:2] 63 64 64 65 66 67 68 68 69 70 ...
## $ edge
## $ Nnode
                : int 61
## $ tip.label : chr [1:62] "Moringua_edwardsi" "Kaupichthys_nuchalis" "Gorgasia_taiwanensis" "Hetero
## $ edge.length: num [1:122] 0.0105 0.0672 0.00537 0.00789 0.00157 ...
## - attr(*, "class")= chr "phylo"
## - attr(*, "order")= chr "cladewise"
```

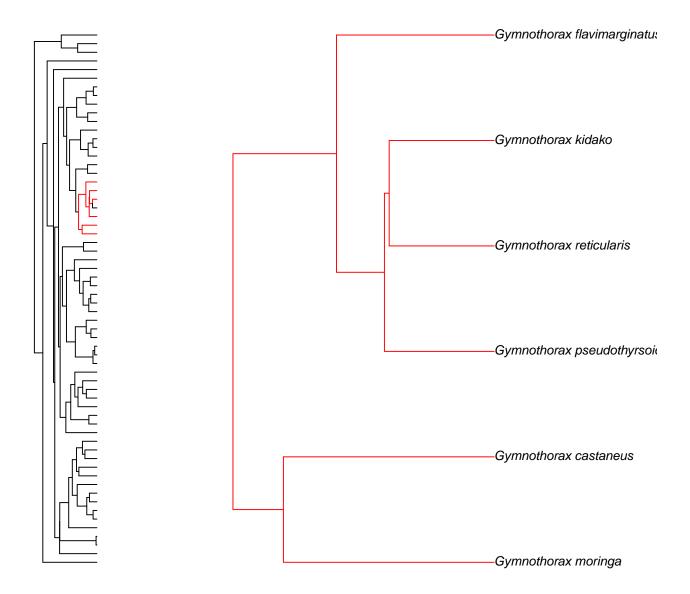
fishtree is a fully resolved tree with branch lengths. There are 62 species and 61 internal nodes. We can plot the tree by using the plot.phylo function of ape. Note that we can just use the function plot to do this as R knows if we ask it to plot a phylogeny to use plot.phylo instead!

```
plot(fishtree, cex = 0.5)
```



cex = 0.5 reduces the size of the tip labels so we can read them. We can also zoom into different sections of the tree that you're interested in:

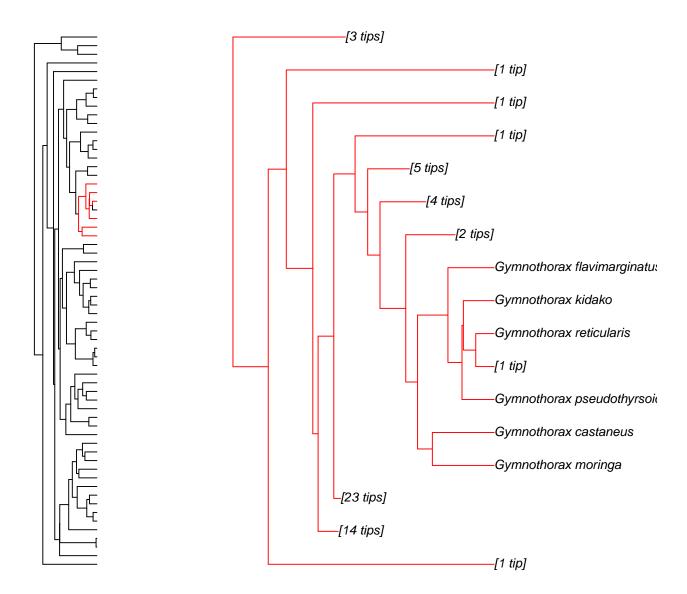
```
zoom(fishtree, grep("Gymnothorax", fishtree$tip.label), subtree = FALSE, cex = 0.8)
```



The grep function is a generic function in R that allows to *grab* any element in an object containing the desired characters. In this example, grep is going to search for all the elements in fishtree\$tip.label that contains Gymnothorax (e.g. Gymnothorax\_kidako, Gymnothorax\_reticularis). Try using only grep("thorax", fishtree\$tip.label) to see if it also only selects the members of the *Gymnothorax* genus.

In this example, we just display the tree for the *Gymnothorax* genus but you can also see how the species fit into the rest of the tree using:

```
zoom(fishtree, grep("Gymnothorax", fishtree$tip.label), subtree = TRUE, cex = 0.8)
```



Note that zoom is a specific plotting function that will automatically set the plotting window to display two plots at once. This might create some conflicts if you're using RStudio. The bug can be easily solved though by typing dev.off() to reinitialise the plotting window and then proceed to the normal zoom(...) function as written above.

You can also reset this to one plot only per window by using:

```
par(mfrow = c(1, 1))
```

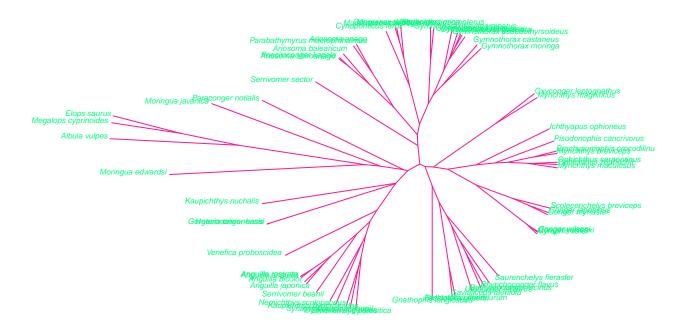
To get further options for the plotting of phylogenies:

```
?plot.phylo
```

Using the question mark (?) can also be done for every function if you want more details!

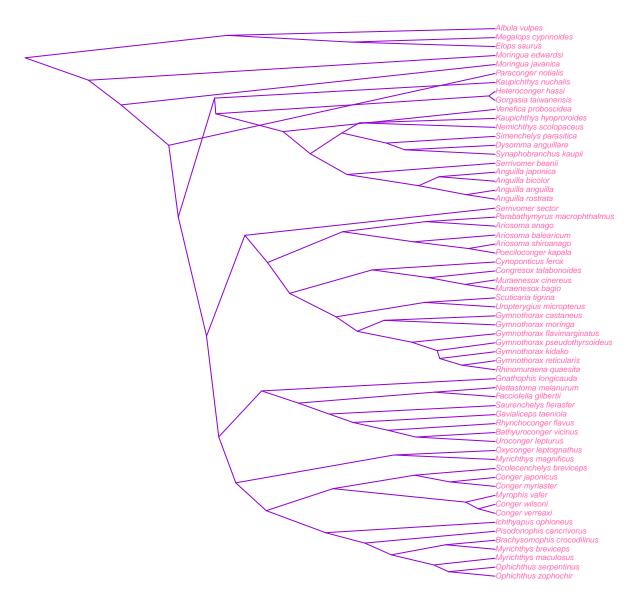
Note that although you can use plot to plot the phylogeny, you need to specify plot.phylo to find out the options for plotting trees. You can change the style of the tree (type), the color of the branches and tips (edge.color, tip.color), and the size of the tip labels (cex). Here's an fun/hideous example!

# plot(fishtree, type = "unrooted", edge.color = "deeppink", tip.color = "springgreen", cex = 0.5)



Or try

plot(ladderize(fishtree), type = "c", edge.color = "darkviolet", tip.color = "hotpink", cex = 0.5)



The ladderize function allows to display the branches from shortest to longest.

Try to modify the graphical options (colors, display, size, ordering of the nodes, etc.) to obtain the most beautiful or ugliest Elopomorpha phylogeny!

# 5. Adding trait data to trees in R

#### A. Ancestral state reconstructions on discrete data

For this exercise we will use the bird data. Remember we already loaded the phylogeny and data as follows:

## data(bird.orders)

First we will invent some data for each bird order that we can reconstruct along the tree. Lets pretend we went in the field and studied each taxa's dietary preferences and found out that they can be sorted into three categories: herbivores, omnivores and carnivores.

```
diets <- c(rep("Herbivore", 5), "Omnivore", rep("Carnivore", 4), rep("Omnivore", 5),
    rep("Herbivore", 2), rep("Omnivore", 3), rep("Carnivore", 2), "Omnivore")</pre>
```

The function rep simply repeats the dietary name n times. Here we assigned the dietary category to each bird group in the order they appear in the phylogeny. You can then check your information by using the function paste that will simply paste together the birds orders and their diet:

```
paste(bird.orders$tip.label, "-", diets)
```

```
"Tinamiformes - Herbivore"
##
    [1] "Struthioniformes - Herbivore"
    [3] "Craciformes - Herbivore"
                                        "Galliformes - Herbivore"
##
    [5] "Anseriformes - Herbivore"
                                        "Turniciformes - Omnivore"
##
   [7] "Piciformes - Carnivore"
                                        "Galbuliformes - Carnivore"
##
   [9] "Bucerotiformes - Carnivore"
                                        "Upupiformes - Carnivore"
## [11] "Trogoniformes - Omnivore"
                                        "Coraciiformes - Omnivore"
## [13] "Coliiformes - Omnivore"
                                        "Cuculiformes - Omnivore"
## [15] "Psittaciformes - Omnivore"
                                        "Apodiformes - Herbivore"
## [17] "Trochiliformes - Herbivore"
                                        "Musophagiformes - Omnivore"
## [19] "Strigiformes - Omnivore"
                                        "Columbiformes - Omnivore"
## [21] "Gruiformes - Carnivore"
                                        "Ciconiiformes - Carnivore"
## [23] "Passeriformes - Omnivore"
```

We can then use the ape function ace to reconstruct ancestral characters along the nodes of the tree. type = d means the character to be reconstructed is discrete.

```
ancestors <- ace(diets, bird.orders, type = "d")</pre>
```

ace uses a likelihood function to estimate what is the probability of each node to be one of the three dietary categories. You can find more information on how ace works by looking at its manual page (?ape) or you can look at the results:

## ancestors

```
##
##
       Ancestral Character Estimation
##
##
  Call: ace(x = diets, phy = bird.orders, type = "d")
##
##
       Log-likelihood: -23.28865
##
## Rate index matrix:
##
             Carnivore Herbivore Omnivore
## Carnivore
                                1
## Herbivore
                     1
                                         1
## Omnivore
                     1
                                1
##
## Parameter estimates:
   rate index estimate std-err
##
                  0.013 0.0049
##
## Scaled likelihoods at the root (type '...$lik.anc' to get them for all nodes):
## Carnivore Herbivore
                           Omnivore
## 0.04505018 0.37554243 0.57940739
```

This gives use information on:

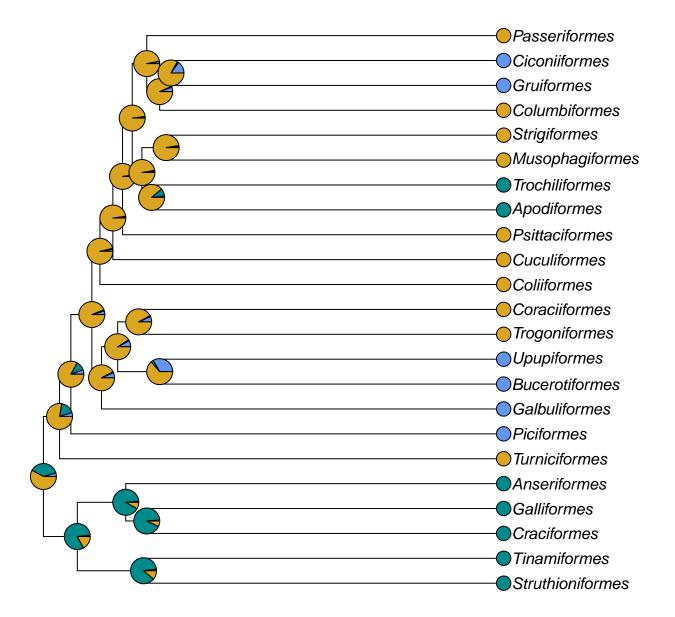
- the rate index matrix (i.e. the probability of changing from one dietary category to another here every probability are as likely)
- the parameter estimates (i.e. the rate or speed at which changes in diet can occur along the branches)
- the scaled likelihood at the root (i.e. the likelihood of the ancestor of all birds to be of one dietary category or the other)

For more clarity, we can plot all this on a phylogeny. First decide which colours we'd like. To look at a list of colours in R type in colors().

```
colours <- c("cornflowerblue", "cyan4", "goldenrod")</pre>
```

Now plot the tree and add labels to the tips and the nodes using the results in ancestors. We use label.offset = 1 to move the labels to the right a bit so the pie charts will fit.

```
plot(bird.orders, label.offset = 1)
tiplabels(pch = 21, bg = colours[as.factor(diets)], cex = 2, adj = 1)
nodelabels(pie = ancestors$lik.anc, piecol = colours)
```



pch = 21 sets the tip labels to be unfilled circles, bg defines the colours of the circles using the list of colours we provided, and ordering them based on what the species value was for mydata (i.e. 0, 1 or 2). cex = 2 doubles the point size, and adj = 1 moves the tip labels sideways so they don't obscure the ends of the branches. pie makes pie charts coloured using the ancestral state reconstructions in ancestors, and piecol tells it to use the colours we have defined.

Again, try to change the graphical parameters to obtain the ugliest or most beautiful figure that tells us about the evolution of diet in birds.

Can you tell what is the most likely diet of the ancestor of these birds? Also what is the rate of change in diet in birds per unit of time?

#### B. Ancestral state reconstructions on continuous data

Similarly to discrete categorical data (see above) we can analysis continuous data to estimate the value of this data for the ancestor of a phylogeny. We are going to use the *Anolis* data to create a phylogeny with

different colours for different observed and reconstructed body sizes (snout-to-vent length, SVL). Remember we already loaded the phylogeny and data as follows:

```
data(anoletree)
anoledata <- read.csv("anole.data.csv", header = TRUE, row.names = 1)</pre>
```

Note the names in anoledata are the species names without the Genus. In the phylogeny the species names are Anolis\_species. So to get the two to match we need to add Anolis\_ to each name.

```
rownames(anoledata) <- paste("Anolis", rownames(anoledata), sep = "_")</pre>
```

As seen before, paste just sticks together Anolis with the names in anoles already with an underscore (\_) separating them (sep = "\_"). We then need to make sure the order of the species in the data matches that of the phylogeny.

```
anoledata <- anoledata[anoletree$tip.label, ]</pre>
```

Next we make a matrix containing only the average snout-to-vent length values for each Anolis species:

```
SVL <- as.matrix(anoledata)[,"AVG.SVL"]</pre>
```

This code selects only the variable AVG.SVL from anoledata (square brackets subset in R in the form [rows, columns]), and then uses as.matrix to make this data frame into a matrix.

Take a look at the first values in SVL:

## head(SVL)

```
## Anolis_ahli Anolis_allogus Anolis_rubribarbus
## 4.039125 4.040138 4.078469
## Anolis_imias Anolis_sagrei Anolis_bremeri
## 4.099687 4.067162 4.113371
```

We can use the ace function as before but by changing the data, the tree and the method (c for continuous):

```
SVLancestors <- ace(SVL, anoletree, type = "c")
```

The estimation uses a similar Maximum Likelihood function as before but now gives us:

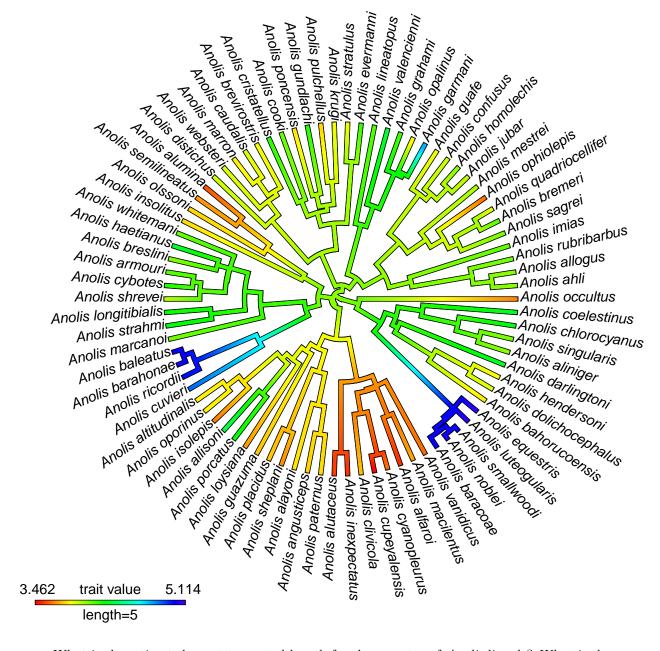
- the estimated value for each node (use SVLancestors\$ace to display it)
- the 95% confidence intervals for each estimate value (use SVLancestors\$CI95 to display it)
- the sigma2 value that is the parameter estimation from the evolutionary model used (by default a Brownian motion model; google it, it's fun!)

To visualise this data, we will us the function contMap from the phytools package. contMap creates a tree with a mapped continuous character, i.e. where the value of the character is known at the tips, and estimated along the tree. The estimating of the character along the tree also uses a Maximum Likelihood estimation procedure similar to ace. Here we will tell contMap not to automatically plot the tree (using plot = FALSE) so we can make some modifications.

```
SVLplot <- contMap(anoletree, SVL, plot = FALSE)</pre>
```

Finally let's plot the tree as a fan (legend = 5 just spreads the legend out so it is readable).

```
plot(SVLplot, type = "fan", legend = 5)
```



What is the estimated snout-to-ventral length for the ancestor of *Anolis* lizards? What is the confidence around this estimation?

By using the plot, can you spot the clades with small or big lizards? Is there any clade where the size seems to be changing a lot in evolutionary time?