## Week 09: Comparative Methods

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Due April  $3^{rd}$ 

## General Ι.

## Type up all of your work in a text editor.

Before you begin, make a new folder in Tasks called Task\_08, and save an empty file named task08.r in that folder.

You must also make a separate "Project" folder within Tasks. Inside your Project folder, make "code", 8 "data", and "text" subfolders. By Friday, you should have some data in the data folder, some code to read 9 the data in to R in the code folder, and a document with a *single sentence* describing your hypothesis. 10

When you're **done** with this assignment, turn it in by (1) saving your text document, (2) opening your Terminal or GitBash, (3) navigating to the appropriate directory using cd and (4) typing: 12

```
git add -A (enter)
13
      git commit -m "Task 08" (enter)
14
      git push -u origin master (enter)
15
```

The main goal this week is a bit meta. Almost everyone is going to end up analyzing data along a 16 phylogeny for their final project. So we're going to walk through some functions that allow you to do that 17 clearly and cleanly.

## II. Your Code!

```
# We're going to unite data and phylogenies into one grand thing
20
21
  # First, set your working directory to your working folder
22
  # Now, we need to tell R that we want to use functions from the phytools
23
     library
24
  # Next, read the phylogeny into R using the read.tree() function (see the
25
      last R assignment for how to do this)
26
27
  plot(tree, type="fan")
28
  # QUESTION 1: How many tips are there in the tree, and are branch lengths
29
     present?
30
  # Now, let's read in the data. It's in the file svl.csv. SVL stands for "
32
     snout-vent length", and it's a standard measure of how long a vertebrate
33
34
  data <- read.csv("https://jonsmitchell.com/data/svl.csv", stringsAsFactors=F
35
      , row.names=1)
36
  # QUESTION 2: What kind of object is ''data''? What are its dimensions?
```

```
39
  # We're going to convert the object data into a vector
40
   svl <- setNames(data$svl, rownames(data))
41
42
  # Now, we have an evolutionary tree of these lizards, and the body size of
43
      the living species. Let's RECONSTRUCT the ancestral states. That is, let'
44
      s estimate how large the ancestors were using the phylogeny and the data.
45
46
   Ancestors <- fastAnc(tree, svl, vars=TRUE, CI=TRUE)
47
  # Examine the Ancestors object and the help file for fastAnc
49
  # QUESTION 3: Where are the estimated values stored? What is the CI95
50
      element?
51
  # QUESTION 4: What are two assumptions made in the estimation of the
52
      ancestral states using fastAnc?
53
54
  # Let's plot all of this together!
55
56
  # First, we'll plot the tree
57
  par(mar=c(0.1,0.1,0.1,0.1))
58
   plot (tree, type="fan", lwd=2, show.tip.label=F)
59
60
  # Now we'll put points instead of names at the tips. The size of the points
61
      will be proportional to the size of the lizard. To do this, we'll use
62
      tiplabels (), but tell it to plot points instead of words (using the pch
      parameter), and to scale the points by the size of each lizard using the
64
      parameter cex. We'll make sure the tips match the points by reorganizing
65
      the svl vector using the tip labels of the tree using square brackets.
66
   tiplabels (pch=16, cex=0.25*svl[tree$tip.label])
67
68
  # Now, we'll add the ancestral states to the tree using nodelabels(), as
69
      each node
70
   nodelabels(pch=16, cex=0.25*Ancestors$ace)
71
72
  # Looking at the points, you can see SOME variation in size, but it's hard
73
      to tell how large or small any given tip is relative to its ancestors or
74
      even to other tips. So let's try visualizing this data in a different way
75
76
77
   obj <- contMap(tree, svl, plot=F)
   plot(obj, type="fan", legend=0.7*max(nodeHeights(tree)), sig=2, fsize=c(0.7,
79
       0.9)
80
81
  # Now, we're going to add some fossils in using data from 'Amber fossils
82
      demonstrate deep-time stability of Caribbean lizard communities' by Emma
83
       Sherratt et al. 2015 in PNAS
84
  # We'll take the measurements and the node closest to each fossil species
   fossilData \leftarrow data.frame(svl=log(c(25.4, 23.2, 17.7, 19.7, 24, 31)), tip1=c(
87
      "aliniger", "aliniger", "occultus", "christophei", "cristatellus", "
88
```

```
occultus"), tip2=c("chlorocyanus", "coelestinus", "monticola", "cybotes",
89
       "angusticeps", "angusticeps"))
90
91
   # What we'll do is: for each fossil, find what node corresponds to the Most
92
      Recent Common Ancestor (MRCA) of the pair of tips in the dataframe.
93
94
   # QUESTION 5: I've set up most of the for loop. You'll need to add the
95
      actual for () line, and close the loop in the approprioate places.
96
   fossilNodes \leftarrow c()
97
   nodeN \leftarrow c()
98
99
   Node <- fastMRCA(tree, fossilData[i, "tip1"], fossilData[i, "tip2"])
100
   fossilNodes[i] <- fossilData[i, "svl"]
101
   nodeN[i] <- Node
102
103
   names(fossilNodes) <- nodeN
104
105
   # Now we estimate the ancestral states, but this time use the fossils to
106
      constrain the estimates at the relevant nodes
107
   Ancestors_withFossils <- fastAnc(tree, svl, anc.states=fossilNodes, CI=TRUE,
108
       var=TRUE)
109
110
     QUESTION 7: How do fossils, in general, change the estimated ancestral
111
      sizes? (Hint, making a plot of the two different ancestral state
112
      estimates against one another will help!)
113
114
     QUESTIONS 8 - 10: Now, I want you to determine if the assumptions you
115
      identified in question 4 hold for these data. The way you'll want to go
116
      about this is by installing a library called "geiger", and use a
117
      function in that package called "fitContinuous". I want you to 1) fit
118
      several different models of evolution to these data, 2) determine which
119
      model fits the data best, and 3) describe whether or not that model is
120
      different from what fastAnc assumes (look at the help page for fastAnc).
121
122
     This will require that you read the help pages for the pertinent functions
123
      , and possible some other functions. Google will be your friend here.
124
      Akaikie Information Criteria (AIC) will be an important facet. We'll talk
125
       about what AIC is in class, but for this assignment all you need to know
126
       is that the model with the LOWEST AIC is the best-fitting model.
127
```