RBrownie User's Guide (Version 0.0.4)

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1 Introducing Brownie

RBrownie is an R package constructed around the C++ command-line program Brownie (available here) which was designed by Brian O'Meara in order to fascilitate the estimation and comparison of rates and means of character evolution over different parts of a phylogeny (O'Meara, et al., 2006). It it built on three primary analyses: First, are censored rate tests under Brownian motion, which allow the comparison of the rates of evolution (under brownian motion) across two monophyletic sister-clades or between a monophyletic and paraphyletic clade. This analysis allows you to answer questions about differences in the evolutionary rate of two distinct clades (e.g. large subspecies vs. small subspecies). Second are non-censored ratetests, so called because they do not remove (censor) any branches of the phylogeny being tested. This class of analyses allows you to compare the rates (Brownian motion model) or means (Ornstein-Uhlenbeck model) of evolution under different character states which are mapped onto the phylogeny. For example, if you had a tree with binary characters mapped onto it branches indicating possession a certain trait and you also had morphological data for all the taxa present in the tree (the tips), then, using a model-testing approach, this analysis allows to you statistically assess whether or not that binary trait is associated with the significantly different rate of evolution of the morphological characteristic measured. Lastly, Brownie performs discrete ancestral state evolution (maximum likelihood approach), which produces branch-annotated trees which can used in these non-censored rate tests.

RBrownie's aim is to exposes all of Brownies capabilities to R users so that they can be used in conjuction with R's existing phylogenetic toolkit. It provides users with easy ways to run common Brownie analyses and the flexibility to write more complex analyses if desired. Below I give a short introduction to Brownie and RBrownie, discusses the main datatypes in RBrownie and how to use them, and then a number of examples of the sort of evolutionary analyses that can be completed in RBrownie, including visualizing and understanding the results.

1.1 More information

Brownie has a large number of functions in addition to the rate tests described above that we cannot describe in full here, but are worth looking into at least because they will help you better understand RBrownie's capabilities and will allow you to exploit all the useful functions RBrownie) has to offer. The manual for Brownie can be found here.

2 Introducing RBrownie

It was mentioned before that RBrownie is an R wrapper for the Brownie program described briefly above. Using the Rcpp package, RBrownie links R with the exact same Brownie code that is used in the Brownie command line program. In addition to providing a streamlined interface to Brownie's core functionality, RBrownie adds plotting and visualization functions which make working with the output data much more easy and flexible.

Like PAUP*, Mesquite, and other phylogenetic programs, Brownie uses specially formatted nexuses files to load phylogenetic data into the program. RBrownie adds functions and classes to help the user read, write, and manipulate these tailored nexus files, supplementing the more generic NEXUS read/write functions found in the ape and phylobase packages (see ape:::read.nexus.data, ape:::write.nexus.data, phylobase:::writeNexus).

To get started using the package (once it has been installed), type the following:

> require(RBrownie)

Note that the other packges on which RBrownie depends are listed. Now that we have the package loaded, we can look into how to load and/or manipulate our phylogenetic data in RBrownie.

2.1 Reading nexus files

Nexus files can be read into R using a number of packages. ape and phylobase both include the functionality to read in standard trees and sequence data, while phylobase also reads standard CHARACTERS (or DATA) blocks. For more information on the options presented by these too packages, check out their documentation:

> ?read.nexus # ape > ?read.nexus.data # ape

> ?readNexus # phylobase

However, Brownie uses nexus blocks beyond those supported by ape and phylobase and RBrownie extends the methods above to accomadate. For example, Brownie uses an ASSUMPTIONS block to define subsets of the taxa which can be used in its various rate tests. It also uses a BROWNIE block to

store a list of brownie commands to be run automatically if the file is "executed" within the Brownie environment (this is similar to how PAUP* executes files). So, RBrownie has its own functions for reading and writing nexus files with these special blocks, readBrownie and writeBrownie.

To illustrate how they work we'll use the parrot dataset which is included with RBrownie (NOTE: you need to have write access to the directory you are in to run this example).

```
> data(parrot)
> writeBrownie(parrot, file = "parrotdata.nex")
> newparrot = readBrownie("parrotdata.nex")
```

The two objects parrot and newparrot should be indentical. It would also be instructive to open the parrotdata.nex file to see how RBrownie writes ASSUMPTIONS blocks, but this isn't necessary. You may have also noticed that parrot and newparrot are both of class list. The list class (a very common class in R) is used when multiple trees are found in a single nexus file. For example, if you openned parrotdata.nex, you would find 10 trees in the TREES block. Each element in the parrot (and newparrot) list contains one of these trees and is of class brownie:

```
> class(parrot)  # list class
> length(parrot)  # == 10
> class(newparrot)  # list class
> length(newparrot)  # == 10
> #
> class(parrot[[1]]) # brownie class representing the first tree in the list
```

brownie is the main class that RBrownie uses to represent phylogenetic data. It and other RBrownie classes are discussed in the next section,

2.2 Classes in RBrownie

At its core RBrownie is built around two major classes, phylo4d_ext and brownie, which both extend phylo4d from phylobase. In fact, phylo4d_ext extends phylo4d and brownie in turn extends phylo4d_ext. In non-computerese this mean that phylo4d_ext adds a few new data containers (or "slots" as they are known in the S4 world) to phylo4d and brownie adds a few new data containers to phylo4d_ext. To illustrate this, run the R code below, it shows the new slots that are added to phylo4d.

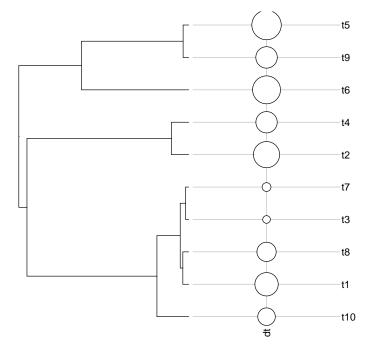
```
> require(RBrownie,quietly=TRUE,warn.conflicts=FALSE)
> phylo4d_slots = names(getSlots("phylo4d"))
> phylo4d_ext_slots = names(getSlots("phylo4d_ext"))
> brownie_slots = names(getSlots("brownie"))
> phylo4d_slots
> # set of differences between phylo4d and phyl4d_ext
```

- > setdiff(phylo4d_ext_slots,phylo4d_slots)
- > # set of differences between phyl4d_ext and brownie
- > setdiff(brownie_slots,phylo4d_ext_slots)

2.2.1 phylo4d_ext class

This class (as implied by the name) is an extension of the phylo4d class from phylobase. This base class holds a phylogenetic tree and a data.frame containing data for each node of the tree. The code below shows how to construct a random coalescent tree and add junk data to the tips of tree which in practice might represent morphological data for each of the taxa.

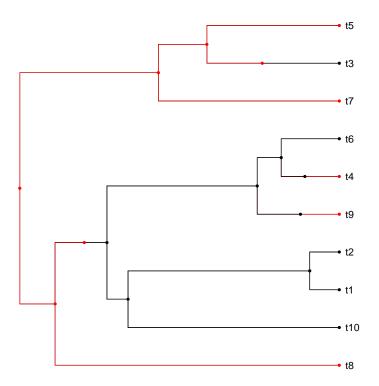
```
> require(phylobase)
> # generate a random coalescent tree with 10 tips
> ape_tree = rcoal(10)
> # convert tree from 'phylo' (ape) to 'phylo4' (phylobase) format
> phy_tree = as(ape_tree, "phylo4")
> # generate junk data for the tips
> sample_tipdata = runif(10)
> # combine the phylogenetic tree and the sample data
> phyd_tree = phylo4d(phy_tree, tip.data=sample_tipdata)
> plot(phyd_tree)
```



phylo4d_ext adds to this class "subnodes" and tree weights. Subnodes are internal nodes mapped directly onto an existing branches and are conceptually similar to singleton nodes (but offer a bit more flexibility from a programming standpoint). The phylo4d_ext class was created mainly to handle SIMMAP-formatted nexus trees, but instances can be created from scratch as well. For example, the code below constructs a random coalescent tree like we did in the last section, but with binary data representing say the presence or absence of a life history trait and also with 4 subnodes along 4 different branches of the tree each representing either loss(0) or gain(1) of the trait along the branch.

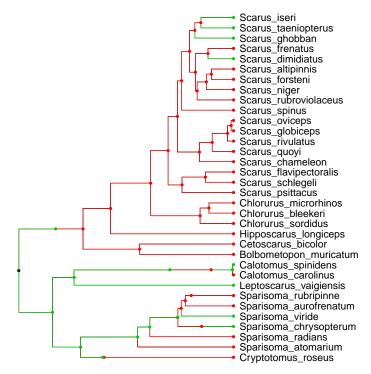
```
> require(RBrownie)
> ape_tree = rcoal(10,br=runif(9,5,15))
> phy_tree = as(ape_tree,"phylo4")
> # Discrete trait data:
> # (Loosely simulate the evolution of a binary trait down a tree,
    choosing a random value for the tip and then changing
     the trait based on the the change matrix.)
> #
> phy_tree = reorder(phy_tree,order="preorder")
> sample_binarydata = data.frame(rep(0,19)) # junk data
> names(sample_binarydata) <- "hasTrait"</pre>
> sample_binarydata[rootNode(phy_tree),1] = sample(c(0,1),1) # choose root value
> changemat = matrix(c(0.8,0.2,0.2,0.8),byrow=T,nrow=2) # prob. of changing on branch
> sample_edges = integer(0) # track edges where change occurs
> for(i in seq(nrow(edges(phy_tree))))
          anc = edges(phy_tree)[i,1]
          desc = edges(phy_tree)[i,2]
          if(anc == 0)
                  next
          newvalue = sample(c(0,1),1,prob=changemat[(sample_binarydata[anc,1]+1),])
          sample_binarydata[desc,1] = newvalue
          if( newvalue != sample_binarydata[anc,1] )
                  sample_edges = append(sample_edges,i)
> #
> #
> # Subnodes
> # (put subnodes on the branches where the trait changed showing
    where on the branch the change happened. Note that the
    subnode data.frame must have the same structure as the
    parent data.frame)
> #
> tmpdata = as.integer(!sample_binarydata[edges(phy_tree)[sample_edges,2],1])
```

```
> sample_subnodedata = data.frame(tmpdata)
> names(sample_subnodedata) <- "hasTrait"</pre>
> #
> # Position the subnodes along their respective branches
> # using the fraction of the overall branch length. In
> # this case, I'll put them near the middle of the branch.
> sample_positions = runif(n=4,min=0.4,max=0.6)
> #
> # Create the phylo4d_ext object, which first creates
> # a phylo4d object and then adds the subnodes to it.
> # NOTE: any "extra" arguments (all.data) are passed
> \# on to the phylo4d constructor via the ellipsis argument (...).
> phyext_tree = phyext(phy_tree,
                          snode.data=sample_subnodedata,
                          snode.branch=edges(phy_tree)[sample_edges,],
                          snode.pos=sample_positions,
                          all.data=sample_binarydata)
> # nodes, subnodes, and branches without the "trait" (state 0) are colored black
> # nodes, subnodes, and branches with the "trait" (state 1) are colored red
> plot(phyext_tree,states=c(0,1),states.col=c("black","red"))
```



In the plot above, the subnodes are clearly seen as dots on an existing branch. This is usually used to indicate state changes along a branch, but because of the randomly generated data above you might see an entirely red branch with a red subnode or something similarly odd. A better example might be seen in the parrot data set.

- > require(RBrownie)
- > data(parrot)
- > plot(parrot[[1]])



Now the binary state data are represented by the default colors and a state change is clearly visible on the branch from the taxon Sparisoma_chrysopterum. In this case, state 0 and state 1 represent a life history trait of parrot fish, namely whether they scrape (0) or excavate (1) coral. You can clearly see that, in this tree at least, that life history trait changes throughout the evolutionary history of the parrot fish at both internal nodes and along a branch at a subnode. Brownie uses this state information, as we will see later, to address questions such as "Does coral excavatingscraping limit morphological evolution in parrotfishes" (Bodega Phylogenetics Wiki, 2010)

Anyhow, to view subnode information these accessor functions are available.

```
> snid(phyext_tree)
```

> sndata(phyext_tree)

You might notice that the subnode position is represented currently as a twocolumn matrix. In a future version, the subnodes will float between two values representing our confidence on where the state change (and thus the subnode) might occur. weight is another aspect of the phylo4d_ext class that will be supported better in future versions.

node.position,
node.data)

Finally, subnodes will most be read in from SIMMAP-formatted nexus files and not added manually. The SIMMAP format was created represent state mappings onto branches (subnodes) in nexus tree files and the standard has gone through a number of iterations. Currently, only SIMMAP version 1.0 and 1.1 are supported by RBrownie and we are working to add support for the latest version (1.5) which is used by Mesquite. RBrownie has functions for reading and / writing in this format with functions similar to ape's read.nexus and write.nexus functions. (NOTE: you need to have write access to the directory you are in to run this example).

```
> data(parrot)
> write.nexus.simmap(parrot, file = "parrottmp.nex")
> newparrot = read.nexus.simmap("parrottmp.nex")
```

2.2.2 brownie class

Many examples in the last section where the phylo4d_ext class was explained used the parrot dataset included in RBrownie. But, the parrot object is a list of brownie objects, right? Yes, the brownie class extends the phylo4d_ext class further (adding two new slots). We can confirm that this is the case:

```
> data(parrot)
> inherits(parrot[[1]], "phylo4d_ext")
```

brownie simply adds a few new slots to phylo4d_ext. The two classes were not combined into one larger class as we thought that the extended phylo4d class would hopefully have some use outside of RBrownie and furture development should reflect that. Moving on though, the two new slots are commands, to hold brownie text commands, and datatypes, which adds annotations to the data slot.

First, the commands slot. Its purpose is to hold the text strings that will eventually be executed in Brownie itself. That is, once the phylogenetic tree and any node data is loaded into the Brownie core, these commands are run. They describe actions that the brownie core should take: analyses might need to be run on certain trees and not others, choose which models to use when model tests are done, etc. A full list and description of all the commands can be found in the Brownie manual.

In common use however, these commands will be filled in automatically by some higher level functions (seen later on) which were designed to make it easy to set up and execute common analysis. If you do find yourself needing to manipulate these commands, there are a number of accessor functions set up for you to do this

```
> require(RBrownie)
> data(geospiza_ext)
> commands(geospiza_ext)
[1] "discrete model= nonrev freq=unif reconstruct=yes file=nonrev_unif.txt replace=y;"
[2] "discrete model=nonrev freq=equilib reconstruct=yes file=nonrev_equil.txt replace=y;"
> # Removal
> geo_none1 = clearCommands(geospiza_ext) # remove all commands
> geo_none2 = removeCommands(geospiza_ext, "discrete") # remove all 'discrete' commands
> geo_3 = removeCommands(geospiza_ext,1) # remove the first command
> # Addition
> geo_add = geospiza_ext
> # add a command to the end of the vector
> commands(geo_add,add=TRUE) <- "choose t=1"
> # add a command at a certain index
> commands(geo_add,add=TRUE,index=1) <- "choose c=1"</pre>
> # replace ALL commands with this one
> commands(geo_add,replace=TRUE) <- "choose d=1"</pre>
> #
> geo_add = geospiza_ext
> # replace command at a certain index with this one
> commands(geo_add,add=TRUE,index=1,replace=TRUE) <- "choose d=1"</pre>
```

Second, the datatypes slot. datatypes is a character vector which contains a description of each column in the data slot. RBrownie uses this information in order to write different parts of the data data.frame to different blocks in a nexus file. For example, discrete and continuous data are treated differently in Brownie

and each require their own nexus block (CHARACTERS and CHARACTERS2 are used). Also, taxa sets are stored in the data slot as binary data (1 is taxon is in the set, 0 if not) and they are written to an ASSUMPTIONS nexus block. To illustrate this, lets look at the data in parrot data set. (NOTE: The data slot comes from phylobase, its accessor functions also come from there. tdata is an example)

```
> require(RBrownie)
> data(parrot)
> # How many columns are in the data.frame?
> ncol(tdata(parrot[[1]]))
[1] 5
> # What are their names?
> names(tdata(parrot[[1]]))
[1] "simmap_state"
[3] "pc2"
                         "TAXSET_all"
[5] "TAXSET_intrajoint"
> # What are their types?
> datatypes(parrot[[1]])
[1] "discrete" "cont"
                           "cont"
                                      "taxset"
                                                  "taxset"
```

We can see here that this brownie object contains five columns in its data.frame and they are of 3 different types. The first discrete, indicates discrete data; the second and third cont indicate continuous data, and the forth and fifth taxset indicate a special binary data column showing which taxa are in a certain set of taxa.

There are handy ways to access these datatype strings when using them:

```
> discData()
> contData()
> taxaData()
> genericData() # undefined data (shouldn't be used)
> brownie.datatypes() # show all
```

Adding new data to a **brownie** object is almost identical to how it is done in phylobase - the **addData** function is actually specified for class **brownie** - with the exception that you can specify which datatype you want to label it as. For example:

```
> require(RBrownie)
> data(parrot)
> # junk morphological data for tips
> junkcont = runif((nNodes(parrot[[1]])+1),-10,10)
```

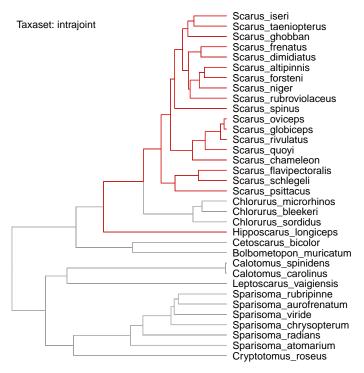
```
> # junk discrete data for tips / nodes
> junkdisc = sample(letters[1:4], length(junkcont) + nNodes(parrot[[1]]),replace=TRUE)
> parrot_new = addData(parrot, tip.data=junkcont, known.types=contData())
> parrot_new = addData(parrot_new, all.data=junkdisc, known.types=discData())
> datatypes(parrot_new)

[1] "discrete" "cont" "cont" "taxset" "taxset"
[6] "cont" "discrete"
```

Now there are two new datatypes in the parrot data which we specified. Also note how addData and datatypes can be called on list objects as well as brownie objects - the list variant of these functions actually adds the data to each object in that list. RBrownie has a number of such convenient functions.

Adding taxa set data is a bit different. There is an accessor function in RBrownie called taxasets which gives the user access to all the data columns representing sets of taxa. It can also be used to create new taxasets:

```
> hasTaxasets(parrot) # see if there are any taxasets
> taxasets(parrot) # return taxasets as data.frame
> #
> # use binary data to indicate membership in a taxa set
> junktaxaset = sample(c(0,1),nTips(parrot[[1]]),replace=TRUE)
> taxasets(parrot,taxnames="Random_set") <- junktaxaset
> #
> # OR
> #
> # use taxa names to indicate membership in a taxa set
> all_sparisoma = grep("Sparisoma",tipLabels(parrot[[1]]),value=TRUE)
> taxasets(parrot,taxnames="sparisomas") <- all_sparisoma
  And you can plot a taxaset as well:
> require(RBrownie)
> data(parrot)
> plot.taxaset(parrot[[3]],2) # plot the second taxaset
Renaming internal nodes (might take a while if there are a lot of taxa):
```



And that is the **brownie** class which can be thought of as a specialization of generic phylogentic trees + data. It would be instructive to take some of the examples above, where data is added to data contained within RBRownie, write it out to a file using **writeBrownie**, and look at how the different changes manifest in the nexus file itself.

3 Running Analyses

Now that you are familiar with the data structures in RBrownie, we can move on to using their methods effectively. RBrownie includes a number of high-level functions to perform the most common evolutionary analyses (discussed in the first three subsections), but also exposes a number of lower-level functions making it easier for the user to tweak or customize the commands the will eventually be run on the Brownie core.

RBrownie provides two general ways to run any of these analyses. First, they can be run directly by calling one of the pre-packaged functions of the form runTEST. These functions clear away anything currently in the commands block and run the TEST specified. Alternatively, commands can be added to a brownie objects one by one using functions of the form addTEST or addOPTION and then all the commands in the brownie object can be executed using run.asis. We'll see examples of all of these below.

3.1 Censored Rate Test

Briefly, the purpose of this test is to calculate and compare rate of continuous character evolution in different parts of a tree. A detailed description of this test is beyond the scope of this guide, but can be found in the Brownie manual.

Brownie uses the 'ratetest' command to perform this test allowing the user to specify a number of optional parameters and so function RBrownie provides functions which add this command along with various options to the commands slot of a brownie object. It requires the use of a phylo4d_ext object as input (see the manual) - Let's look at an example:

[1] "ratetest taxset=intrajoint reps=1000 treeloop=no charloop=yes quiet=no;"

As you can see, the 'ratetest' command has been added to all the brownie objects in the parrot list. Now that we have added this command, we can run the analysis manually:

```
> # run the brownie objects as is and get the text
> # output from the brownie core
> rawtxt = run.asis(par.new)$textout
> rawtxt = scan.textout(rawtxt) # process text output
> test.results = read.ratetest.output(txt=rawtxt)
> summaryRatetest(test.results)
```

The last command, summaryRatetest, provides a summary of the ratetest results - it's a very handy command, providing a rough interpretation of the data in test.results. Again, the reader is referred to the brownie manual for a more detailed description of the data.frame and its summary. But getting back to RBrownie, there is a more straight-forward way to run this test than the manual method used above that adds commands and processes the text output automatically. Here it is:

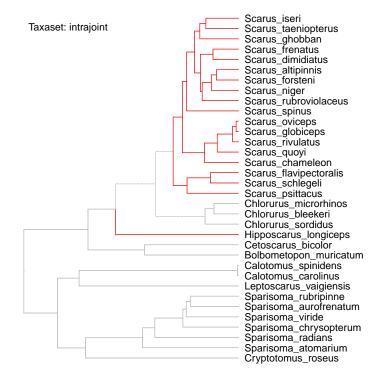
```
> require(RBrownie)
> data(parrot)
> test.results = runCensored(parrot, taxset = "intrajoint",
+    reps = 1000, charloop = T)
> summaryRatetest(test.results)
```

3.1.1 Plots

In addition to the summaryRatetest function, there is also a specialized plotting function for visualizing which branches of a tree the ratetest command

will remove (see the brownie manual for more information on this). The branches to be removed are shown as dashed, the branches of the taxa in the taxaset 'intrajoint' are shown in red.

Renaming internal nodes (might take a while if there are a lot of taxa):



3.2 Non-Censored Rate Test

The non-censored rate test is similar to the original censored version, except that where exactly on the branch the state changes matters to the likelihood estimate. Agian, describing the test in full is outside of the scope of this guide, but a full description is available in the Brownie manual and (O'Meara, et al., 2006).

Brownie uses the 'cont' command to perform this test - let's look at an example where we will run and compare two different models of morphological evolution (in this case, the two models are based on brownian motion):

As you can see, in the non-censored case it is necessary to add more than one command - two for the two different models and two to instruct the Brownie core execute a Non-censored ratetest using which ever continuous evolution model is set. Executing the whole bunch is similar to the censored test:

```
> # run the brownie objects as is and get the text
> # output from the brownie core
> rawtxt = run.asis(par.new)$textout
> rawtxt = scan.textout(rawtxt) # process text output
> test.results = read.continuous.output(txt=rawtxt)
> summaryCont(test.results)
```

test.results is again a data frame with the log-likelihood and AIC values calculated for each tree, character, and model used and summaryCont prints out the model comparisons for each character, averaging over all the trees used. For more information about the test and how to interpret it, check out the Brownie manual and (Bodega Phylogenetics Wiki, 2010).

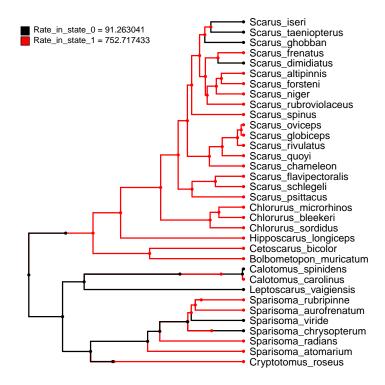
Again, there is a more direct way to execute a non-censored rate test:

In the example above runNonCensored removes all commands currently in the commands slot and replaces them with the same commands we added earlier manually and then executes the resulting brownie list (all within the function). In returns the same data frame.

3.2.1 Plots

Some of the models tested return values that can be plotted on top of phylogenetic trees for easier visualization. One such model is "BMS", Brownian motion with rate varying by state of the stochastically. It returns two different evolutionary rates, one for when the internal state is 0 and one is 1.

```
> # run non-censored rate test
> require(RBrownie)
> data(parrot)
> test.results = runNonCensored(parrot,models=brownie.models.continuous()[1:2],
          treeloop=T, charloop=T)
> #
> # Plot the evo rates from the BMS model
> # use tree #1 and character #1
> modname = "BMS"
> tree = 1
> char = 1
> x = parrot[[tree]]
> states = c(0,1)
> state.cols = c(1,2)
> line.widths = c(2,2)
> state.ratename = c("Rate_in_state_0", "Rate_in_state_1")
> rowind = which(test.results$Tree==tree &
          test.results$Char==char &
         test.results$Model == modname)
> junkrow = test.results[rowind,]
> colinds = which(colnames(junkrow) %in% state.ratename)
> legends = paste(names(junkrow[colinds]), junkrow[colinds], sep=" = ")
> plot(x,states=states,states.col=state.cols,line.widths=line.widths,plot.points=T)
> #
> # Add jury-rigged legend
> fontsize=10
> boxSize = unit(fontsize, "points")
> for(ii in seq(length(states)))
          yoff = unit( ((ii-1)*fontsize)+ ifelse(ii==1,0,2) ,"points")
          grid.text(legends[ii],
          x=unit(2, "mm"),
          y=unit(1, "npc") - unit(2, "mm") - yoff,
          just=c("left", "top"),
          gp=gpar(fontsize = fontsize))
          grid.rect(x=unit(2, "mm") - boxSize,
          y=unit(1, "npc") - unit(2, "mm") - boxSize - yoff,
          width = boxSize, height = boxSize,
          just = "bottom", gp = gpar(fill = state.cols[ii]))
+ }
```



3.3 Discrete Ancestral State Reconstruction

Brownie can do discrete character reconstructions about which more information can be found here. Also check out the addDiscrete documentation.

Brownie uses the 'discrete' command to run these tests. Unlike the Censored and Non-Censored rate tests, this command simply requires a phylogenetic tree (with no branch annotations) and discrete data at the tree tips. Let's see an example using the geospiza_ext dataset built into RBrownie:

```
+ freq="equilib",
+ reconstruct=T,
+ fileappend=T)
> commands(geoext)
```

- [1] "discrete model=nonrev freq=unif treeloop=no charloop=no allchar=no variable=no reconstr
- [2] "discrete model=nonrev freq=equilib treeloop=no charloop=no allchar=no variable=no recor

At this point we have essentially re-created the geospiza_ext object. Now we will manually run the analysis:

```
> rawout = run.asis(geoext)
```

If we look at the rawout we will see that both list elements textout and treesout are non-null. The treesout element includes the trees with reconstructed discrete values mapped on to either branches - in this case two model, freqs combinations were analyzed so two tree reconstructions are returned. But first, process the text output using some familiar commands.

```
> rawtext = scan.textout(rawout$textout) # process text output
> test.results = read.discrete.output(txt=rawtext)
```

And the trees can be retrieved like so:

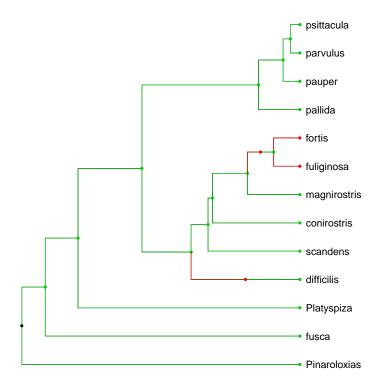
```
> # condition brownie output, by removing the ticks it adds
> treesout = gsub("'","",rawout$treesout)
> #
> # retrieve phylo4d objects with subnodes as singletons
> tree1 = read.simmap(text=treesout[1])
> tree2 = read.simmap(text=treesout[2])
> #
> # Convert singletons to subnodes
> # (and phylo4d objects to phylo4d_ext objects)
> tree1.ext = phyext(tree1)
> tree2.ext = phyext(tree2)
> #
> # Visualize the results:
> plot(tree1.ext)
> plot(tree2.ext)
```

Once more, there is a less round-about way to run this analysis:

```
> require(RBrownie)
> data(geospiza_ext)
> junkrun=runDiscrete(geospiza_ext,
+ models=c("nonrev","nonrev"),
+ freqs=c("unif","equilib"),
+ reconstruct=T)
```

Analysis returned some warnings:

- [1] "WARNING: if some characters have fewer than 2 character states, the output in the output
- [2] "WARNING: if some characters have fewer than 2 character states, the output in the output
- > #
- > # View the reconstruction results for the first tree:
- > plot(junkrun\$trees[[1]])



3.4 Custom Tests

As shown above all of the runTEST commands can also be put together manually using addTEST and addOPTION commands. There is an add command for most Brownie options and commands - to view the available options, see:

> help.search("add", package = "RBrownie")

4 Final Thoughts

More work is currently in progress to have RBrownie support more branch annotation formats (those currently supported by Mesquite and BEAST) and work being done to support returning confidence intervals for rate tests and

reconstructions. Check back soon to see if there are any updates or more detailed tutorials.

We are continually updating RBrownie and would love your feedback on how it can be improved. If there is a feature you would like to see supported in either Brownie or RBrownie, please post a feature request to the brownie-users google group.

References

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