# An introduction to pez

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## 1 Preamble

You can install pez by typing install.packages("pez"), and get a listing of the functions in the package by typing library(help=pez). If you find any bugs, or have any feature requests for the package, please use the online tracker at http://github.com/willpearse/pez/iIndeed, please contribute to the package using at its GitHub site—help is always welcome! If you just can't wait to get the latest version, you can install the latest version directly from GitHub (require(devtools); install\_github('willpearse/pez')).

While **pez** contains much novel code, it relies heavily on the R ecosystem. Much of the community phylogenetic metric functions are wrappers around existing code (detailed in the help files for each function); notably caper (Orme et al., 2013) and picante (Kembel et al., 2010) but many others. Please cite the authors of these packages so that their hard-work is rewarded!

## 2 Data formats in pez

pez functions work with comparative community ecology objects, called comparative.comm objects. These are designed to help keep phylogenies, community data matrices, species trait data, and environmental data all in the same place in a format that makes it easy to work with them. They're much less scary than they sound!

Below we load *pez*, some example data that comes with it, and then make a comparative.comm object. You can examine the phylogeny (tree), community data (comm), and trait data (data) that went into making dataset for yourself, although all the data types are explained in more detail below. Below we use the Helmus & Ives (2012) dataset to show *pez*'s features.

```
library(pez)
data(laja)
data <- comparative.comm(invert.tree, river.sites, invert.traits)
data

## Comparative community dataset of 60 taxa:
## Phylogeny: invert.tree
## 54 internal nodes, VCV matrix present
## Community data: river.sites
## 11 sites, 60 taxa
## Trait data: invert.traits
## 2 variables
## Environmental data: None</pre>
```

pez is conservative; if you give it trait data for only half of the species in your community data, the comparative.comm object will only contain data on those species that have both trait data and community data. The same goes for the phylogeny, and for sites with environmental data. pez will warn you about the loss of species or traits when you print the object to screen, and while it's making the comparative.comm object (unless you set the argument warm=FALSE).

You can also subset your comparative.comm object to exclude certain species or sites, in much the same way you can a data.frame. Note that *pez* will not (by default) warn you if this operation drops out certain species or sites. For example:

```
data[1:5,]
## Comparative community dataset of 60 taxa:
## Phylogeny: x$phy
      54 internal nodes, VCV matrix present
## Community data: comm
##
        5 sites, 60 taxa
## Trait data: new.x$data
##
       2 variables
## Environmental data: None
data[,1:3]
## Comparative community dataset of 3 taxa:
## Phylogeny: phy
      2 internal nodes, VCV matrix present
## Community data: comm
        11 sites, 3 taxa
## Trait data: traits
##
       2 variables
## Environmental data: None
data[,1:3,warn=TRUE]
## Warning: Mismatch between phylogeny and other data, dropping 57 tips
## Warning: Mismatch between community matrix and other data, dropping
57 columns
## Warning: Mismatch between traits and other data, dropping 57 columns
## Comparative community dataset of 3 taxa:
## Phylogeny: phy
      2 internal nodes, VCV matrix present
## Community data: comm
        11 sites,
                    3 taxa
## Trait data: traits
       2 variables
## Environmental data: None
```

### 2.1 Phylogenies

pez uses the phylo format in the ape package to store phylogenies. You can load your own phylogenies using the ape functions read.tree and read.nexus.

### 2.2 Community data

pez uses the same community data format as the vegan package: a matrix with sites in the rows and species in the columns. The elements of the community matrix can be species abundances or presence/absence (1/0). Not all the species in your matrix have to be present in a site—there can be empty columns in your data. This is particularly important when using the dispersion measures (see below). Your data should be named, with row names that correspond to sites, and column names that correspond to species.

#### 2.3 Trait data

Trait data should be a data.frame with row names that correspond to the species in the phylogeny, and named columns for each separate trait.

#### 2.4 Environemntal data

Environmental data should be a data.frame with row names that correpsond to the sites in your community data, and separate (named) columns for each kind of environmental data.

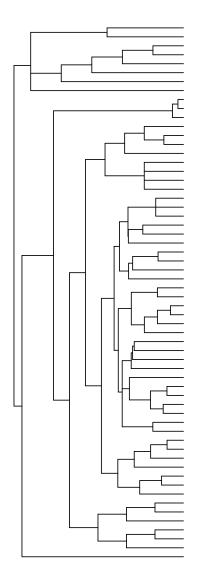
## 3 Plotting and exploring data

pez comes with a few functions that are intended to make exploring your data slightly easier. For instance, you can plot out graphs of species abundances in communities

```
cc.dotplot(data)
## Error: could not find function "cc.dotplot"
```

```
plot(data)
```

## Error: invalid 'cex' value



## 4 Community phylogenetic metrics

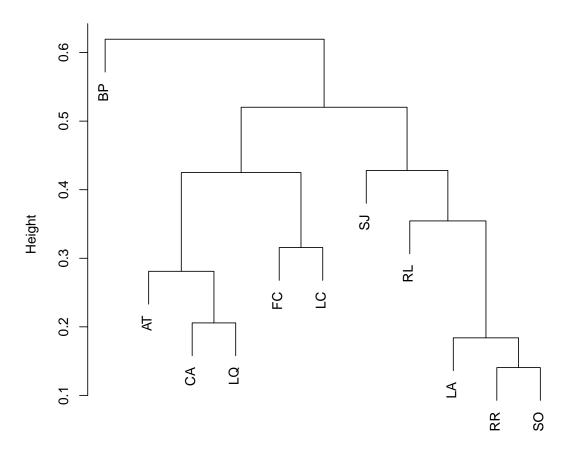
pez splits community phylogenetic metrics into four functions according to the scheme outlined by Pearse et al. (2014): shape, evenness, dispersion, and dissimilarity. Shape metrics measure the structure of an community phylogeny, while evenness metrics additionally incorporate species abundances. Dispersion metrics calculate examine whether phylogenetic biodiversity in an assemblage differs from the expec-

tation of random assembly from a given set of species. Finally, dissimilarity measures the pairwise difference in phylogenetic biodiversity between assemblages.

You can calculate all metrics within a class at the same time (which is what we recommend), or you can pick a particular one. Below we show how to calculate the metrics, and give examples of how to work with their output.

```
shape.output <- shape(data)</pre>
coef(shape.output)
##
                                            Delta DeltaStar LambdaPlus DeltaPlus
         psv
                 psr
                        mpd
                               pd
                                   pd.ivs
## AT 0.6721 19.490
                      985.3 9178
                                    24.43
                                            985.3
                                                       985.3
                                                                  123300
                                                                             985.3
## BP 0.7923
               6.339 1161.6 3726
                                     6.06 1161.6
                                                      1161.6
                                                                  121966
                                                                            1161.6
## CA 0.6308 15.139
                      924.8 8117
                                   257.22
                                            924.8
                                                       924.8
                                                                  106561
                                                                             924.8
## FC 0.6729 12.112
                      986.4 6333
                                    26.05
                                            986.4
                                                       986.4
                                                                   92822
                                                                             986.4
## LA 0.6660
              7.993
                      976.4 4792
                                    37.00
                                            976.4
                                                       976.4
                                                                 122011
                                                                             976.4
## LC 0.7621 16.005 1117.3 7549
                                   465.97 1117.3
                                                      1117.3
                                                                  123918
                                                                            1117.3
## LQ 0.6481 18.147
                      950.1 8696 -198.32
                                            950.1
                                                       950.1
                                                                             950.1
                                                                  100305
## RL 0.5551
               9.992
                      813.8 5826 -481.20
                                            813.8
                                                       813.8
                                                                  76909
                                                                             813.8
## RR 0.6307 10.091
                      924.6 5612 -178.00
                                            924.6
                                                                 105977
                                                                             924.6
                                                       924.6
## SJ 0.6098 12.197
                      894.0 6589 -235.56
                                            894.0
                                                       894.0
                                                                  94568
                                                                             894.0
## SO 0.6678 10.016
                      978.9 5807
                                   276.35
                                            978.9
                                                       978.9
                                                                  103177
                                                                             978.9
##
      S.DeltaPlus eigen.sum
                                EED
                                       HED
## AT
                    0.017566 1.737
             28573
                                     5.848
## BP
              9292
                    0.017852 5.141 10.690
## CA
             22194
                    0.010539 2.008
                                     6.380
## FC
            17756
                    0.010592 2.621
                                     7.575
## LA
             11717
                    0.012937 3.670
                                     9.119
## LC
             23463
                    0.023911 2.205
                                     6.712
## LQ
             26604
                    0.010663 1.824
                                     6.077
## RL
             14648
                    0.007254 2.773
                                     8.015
                    0.009443 2.964
## RR
             14793
                                     8.218
## SJ
             17881
                    0.008546 2.461
                                     7.374
## SO
             14684
                    0.011426 2.966
                                     8.032
shape.output$mpd
    [1]
##
         985.3 1161.6
                        924.8
                                986.4 976.4 1117.3 950.1
                                                              813.8
                                                                             894.0
## [11]
         978.9
dissimilarity.output <- dissimilarity(data, metric="phylosor")</pre>
plot(hclust(dissimilarity.output$phylosor))
```

### **Cluster Dendrogram**



dissimilarity.output\$phylosor hclust (\*, "complete")

Note that phylosor (Bryant et~al., 2008) is reported as a dissimilarity in pez: it's not the fraction of shared branch lengths, but 1-fractionshared branch length. Remember: the function is called dissimilarity!

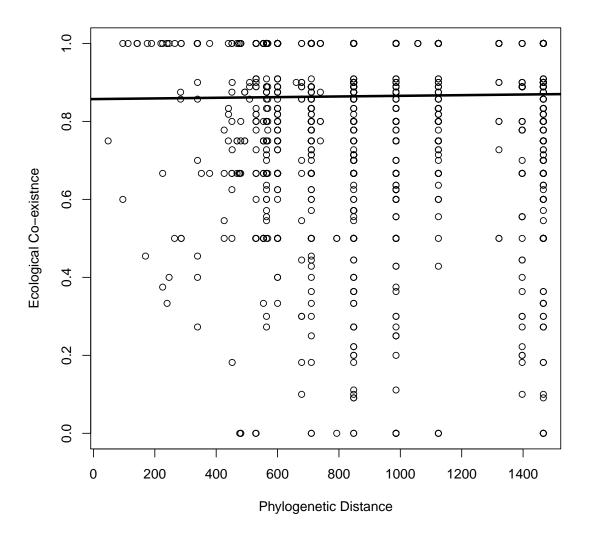
# 5 Eco-evolutionary regression (pez)

pez is intended to replace and improve upon earlier Visual Basic/C programs called EcoPhyl. You can regress the relative coexistence of species in your dataset against

those species phylogenetic (eco.phy.regression) and trait (eco.trait.regression) dissimilarity, as well as shared habitat preferences based on environmental tolerances (eco.env.regression).

```
model <- eco.trait.regression(data, method="lm")</pre>
model
##
## eco.trait.regression
## Method: lm
## Randomisation: NONE
## Observed slope: 0.01
## Observed model summary:
##
## Call:
## lm(formula = as.numeric(eco.mat) ~ as.numeric(trait.mat))
## Residuals:
      Min
               1Q Median
                                       Max
## -0.8387 -0.1097 0.0636 0.1756 0.2025
##
## Coefficients:
##
                        Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                          0.79748
                                     0.01012
                                              78.80 <2e-16 ***
## as.numeric(trait.mat) 0.01283
                                     0.00506
                                               2.54
                                                       0.011 *
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.218 on 1767 degrees of freedom
     (1 observation deleted due to missingness)
## Multiple R-squared: 0.00363, Adjusted R-squared: 0.00306
## F-statistic: 6.43 on 1 and 1767 DF, p-value: 0.0113
more.complex.model <- eco.trait.regression(data, method="mantel", altogether=FALSE, p
## Error: incorrect number of dimensions
more.complex.model
## Error: object 'more.complex.model' not found
```

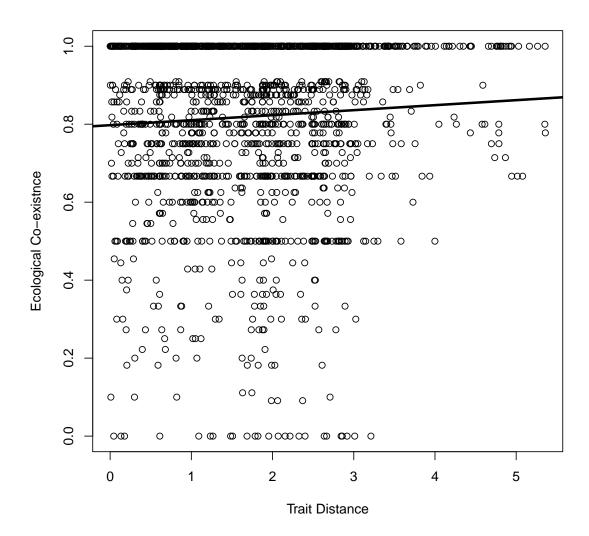
```
more.complex.model[[1]]
## Error: object 'more.complex.model' not found
plot(eco.phy.regression(data, method="quantile"))
```



However, the real power in this approach comes from combining information about the phylogenetic signal of species traits with the output from an eco.trait.regression of those traits. This amounts to plotting something about the evolution of those traits

against their ecology. This sort of approach was first proposed by Cavender-Bares et al. (2004) (figure 4), and you can use the fingerprint.regression function to carry it out. Originally, Cavender-Bares et al. (2004) proposed the use of Mantel regressions (eco.phy.regression), but in pez you can use measures of phylogenetic signal (phy.signal). There's no need to do all these steps separately, though:

```
model <- fingerprint.regression(data, eco.permute=10)
## Error: object of type 'closure' is not subsettable
plot(model)</pre>
```



## 6 TO-DO: Traits

## 7 Simulation

A good simulation is one that does exactly what you want it to do, and pez provides a number of simulation functions that may be useful to you as (1) tools, or (2) starting

points for your own simulations.

scape allows you to repeat the analysis of Helmus & Ives (2012), simulating the assembly of species across a landscape given phylogenetically structured assembly. The parameters are complex, but they can generate some useful expected distributions, and give you a feel for regional assembly. You'd do well to read the paper that generated these models, but here's an example of their use:

```
## Loading required package: ape

tree<-stree(8,type="balanced")
kk<-scape(tree, scape.size=100, g.center=100, g.range=1, g.repulse=1, wd.all=150, sig

## Error: could not find function "scape"

require(plotrix)
par(mfrow=c(1,Ntip(tree)),mar=c(.1,.1,.1,.1))
for(j in 1:Ntip(tree)){color2D.matplot(1 - kk$sppXs[,,j]/max(kk$sppXs[,,j]), xlab = '

## Error: object 'kk' not found

par(mfrow=c(2,1))
matplot((kk$X1), type = "l", xlab="gradient",ylab = "probability", main = "Gradient 1")

## Error: object 'kk' not found

matplot((kk$X2), type = "l", xlab="gradient",ylab = "probability", main = "Gradient 2")

## Error: object 'kk' not found</pre>
```

TO-DO: Alternatively, you can model the evolution of species and, at the same time, their assembly through a community. The only problem here is that the models are much simpler, but hopefully they are tunable to your liking!:

TO-DO

## References

Bryant, J.A., Lamanna, C., Morlon, H., Kerkhoff, A.J., Enquist, B.J. & Green, J.L. (2008) Microbes on mountainsides: contrasting elevational patterns of bacterial

- and plant diversity. *Proceedings of the National Academy of Sciences* **105**, 11505–11511, URL http://www.pnas.org/content/105/suppl.1/11505.abstract.
- Cavender-Bares, J., Ackerly, D.D., Baum, D.a. & Bazzaz, F.a. (2004) Phylogenetic overdispersion in Floridian oak communities. *The American Naturalist* **163**, 823–43.
- Helmus, M.R. & Ives, A.R. (2012) Phylogenetic diversity-area curves. *Ecology* **93**, S31–S43.
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- Pearse, W.D., Cavender-Bares, J., Puvis, A. & Helmus, M.R. (2014) Metrics and models of community phylogenetics. *Modern Phylogenetic Comparative Methods and their Application in Evolutionary Biology—Concepts and Practice* (ed. L.Z. Garamszegi), Springer-Verlag, Berlin, Heidelberg.