

Introduction to FestR

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This package is based on the [MCMCglmm package](#) and runs a MCMCglmm analysis on multiple trees using the [MulTree package](#).

Installation

To install FestR its dependency [MulTree package](#) must first be installed directly from GitHub using the following:

```
if(!require(devtools)) install.packages("devtools")
install_github("TGuillaume/mulTree", ref = "master")
```

Following this you can then install FestR directly from GitHub using the following:

```
install_github("healyke/FestR", ref = "master")
```

And load in the data

```
library(mulTree)
library(FestR)
```

Read in the data

```
#read in the data
mydata<-read.csv(file=system.file("extdata", "FestR_data.csv", package = "FestR"),
  header=T, stringsAsFactors = F)

# re-order it into species
mydata <- mydata[order(mydata$species),]

#read in the phylogentic information
#first the mammal trees
mammal_trees <- read.tree(system.file("extdata", "3firstFritzTrees.tre", package = "FestR"))
#than the bird trees
bird_trees <- read.tree(system.file("extdata", "3firstJetzTrees.tre", package = "FestR"))

#combine them together using the tree.bind function from the mulTree package
combined_trees<-tree.bind(x = mammal_trees, y = bird_trees, sample = 2, root.age = 250)
```

Testing the new data: `set.tef.est`

In order to estimate a trophic enrichment factor for a new species we need to check that the species is already present in our phylogeny and check what data is available for the new species.

This data set.tef.est checks for includes:

tissue type (“blood”, “claws”, “collagen”, “feather”, “hair”, “kidney”, “liver”, “milk”, “muscle”),

habitat (“terrestrial”, “marine”),

diet.type (“carnivore”, “herbivore”, “omnivore”, “pellet”),

and the isotopic ratios of the food sources for carbon

(source.iso.13C)

and nitrogen

(source.iso.15N).

```
#####function that checks the dat for some species we want to estimate TEF for
new.data.test <- setTefEst(species = "Meles_meles",
                           habitat = "terrestrial",
                           taxonomic.class = "mammalia",
                           tissue = "blood",
                           diet.type = "omnivore",
                           source.iso.13C = c(-24.1),
                           source.iso.15N = c(7.0),
                           tree = combined_trees)
```

```
## Meles_meles present in phylogeny
```

If the species is not in the phylogeny already such as the Komodo dragon (*Varanus komodoensis*), or we are missing on of the values, in this case diet.type we get warning messages to indicate what is missing from our data.

```
## Varanus_komodoensis not present in phylogeny
```

Formating the new data: tef.mul.clean

We now need to format the data by combining both the isotpic data already avialbe within the package and the data from the new species and matching these species to the included phylogeny. We also include what isotope we want to estimate a trophic discrimination value for (either “carbon” or “nitrogen”).

```
tef_data_c <- tefMulClean(new.data = new.data.test, data = mydata, species.col.name = "species", tree =
```

```
## The following taxa were dropped from the analysis:
```

```
## Tachyglossus_aculeatus Zaglossus_bruijni Zaglossus_bartoni Ornithorhynchus_anatinus Anomalurus_beec:
```

We now have a mulTree class object, which is required by the imputation analysis, that contains the matched phylogeny and a dataset containing isotpic data with the new species for which you want to estimate a trophic enrichment factor at the top with a NA for either delta13C or delta15N depending on isotope.

```
tef_data_c$phy
```

```
## 2 phylogenetic trees
```

```
head(tef_data_c$data)
```

```
##           sp.col      habitat taxonomic.class tissue diet.type
## 1      Meles_meles terrestrial      mammalia  blood  omnivore
## 130     Bos_taurus terrestrial      mammalia   hair  herbivore
## 94  Callorhinus_ursinus      marine      mammalia  blood  carnivore
## 95  Callorhinus_ursinus      marine      mammalia  blood  carnivore
## 132     Capra_hircus terrestrial      mammalia   hair  herbivore
## 31     Enhydra_lutris      marine      mammalia   hair  carnivore
##      source.iso.13C delta13C           animal
## 1             -24.1      NA      Meles_meles
## 130            -27.0      2.7      Bos_taurus
## 94             -18.9      0.2 Callorhinus_ursinus
## 95             -18.9      1.0 Callorhinus_ursinus
## 132            -27.0      3.2      Capra_hircus
## 31             -17.8      2.8      Enhydra_lutris
```

Running the analysis: `tef.mul.clean`

With the data formatted as a `mulTree` object we now need to set up the model. In this case we will run the full model to estimate `delta13C` with the fixed factors of source isotopic ratio, diet type and habitat type.

```
formula.c <- delta13C ~ source.iso.13C + diet.type + habitat
```

and random terms that includes the “animal” term which is required to include phylogeny into the analysis, `sp.col` to allow multiple species entries into the analysis and account for variation within a species, and tissue type.

```
random.terms = ~ animal + sp.col + tissue
```

As we rely on Bayesian imputation to estimate the missing value we also need to specify a prior, in this case we follow the standard prior from `MCMCglmm`

```
prior_tef <- list(R = list(V = 1/4, nu=0.002), G = list(G1=list(V = 1/4, nu=0.002),
      G2=list(V = 1/4, nu=0.002), G3=list(V = 1/4, nu=0.002)))
```

finally we can run the analysis

```
Tef_est.c <- tefMcmcglmm(mulTree.data = tef_data_c,
      formula = formula.c,
      random.terms = random.terms,
      prior = prior_tef,
      output = "test_c_run",
      nitt = c(1200000),
      thin = c(500),
      burnin = c(200000),
      no.chains = c(2),
      convergence = c(1.1),
      ESS = c(1000))
```

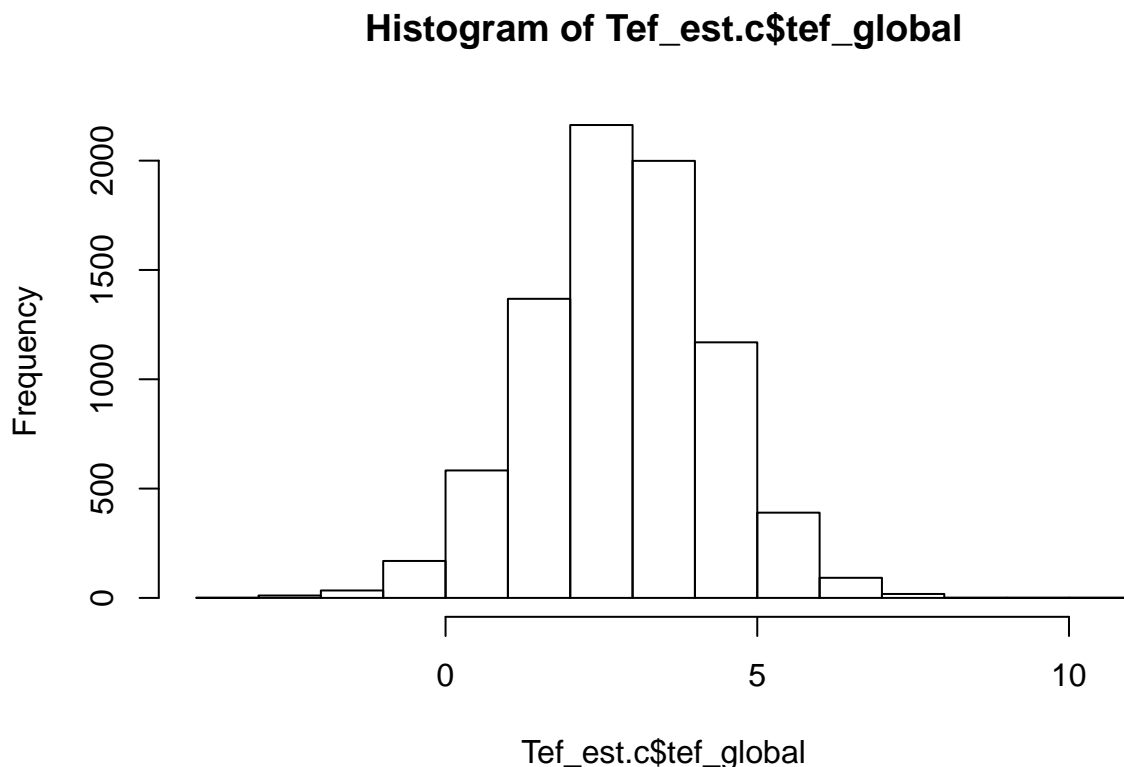
```
## Output chain name "test_c_run" already exists!
## Press [enter] if you wish to overwrite the models or [esc] to cancel.
```

```
## Models will be overwritten...

##
## 2016-04-07 - 10:30:09: MCMCglmm performed on tree 1
## Convergence diagnosis:
## Effective sample size is > 1000: TRUE
## 2000; 2000; 1745.184; 2000; 2000; 2000; 2165.34; 2184.666; 2198.561; 2000; 2388.605; 1931.708
## All levels converged < 1.1: TRUE
## 1.001197; 1.000068; 1.000801; 1.000428; 1.000435; 1.000134
## Individual models saved as: test_c_run-tree1_chain*.rda
## Convergence diagnosis saved as: test_c_run-tree1_conv.rda
##
## 2016-04-07 - 10:34:57: MCMCglmm performed on tree 2
## Convergence diagnosis:
## Effective sample size is > 1000: TRUE
## 2230.65; 2000; 2073.291; 1664.955; 2000; 2358.108; 1822.325; 1792.695; 2177.565; 2149.792; 2000; 2000
## All levels converged < 1.1: TRUE
## 1.00063; 1.001684; 1.000417; 0.9996571; 1.001278; 0.9996844
## Individual models saved as: test_c_run-tree2_chain*.rda
## Convergence diagnosis saved as: test_c_run-tree2_conv.rda
##
## 2016-04-07 - 10:34:57: MCMCglmm successfully performed on 2 trees.
## Total execution time: 9.273651 mins.
## Use read.mulTree() to read the data as 'mulTree' data.
## Use summary.mulTree() and plot.mulTree() for plotting or summarizing the 'mulTree' data.
```

explain what is going on here

```
hist(Tef_est.c$tef_global)
```



```
median(Tef_est.c$tef_global)
```

```
## [1] 2.845063
```