

A jrich worked example

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29/02/2016

These functions calculate the taxonomic measures presented in Miranda-Esquivel (2016). The package introduces Jack-knife resampling in evolutionary distinctiveness prioritization analysis, as a way to evaluate the support of the ranking in area prioritization, and the persistence of a given area in a conservation analysis.

For further information, you could read http://link.springer.com/chapter/10.1007/978-3-319-22461-9_11

1. An example with a single topology and distribution to reproduce Figure 1 in Miranda-Esquivel (2016)

First of all, we remove everything from the R environment.

```
#  
rm(list=ls())
```

And close the graphic devices, if there is any.

```
#  
if (dev.cur()!=1){dev.off()}
```

```
## null device  
##          1
```

Now, we get the latest version of the library, from GitHub.

```
#library("devtools")  
#install_github("Dmirandae/jrich")
```

We load the library before proceeding.

```
library(jrich)
```

```
## Loading required package: ape
```

ans set working directory to the R Data (make sure you have changed it to your own directory).

```
# setwd("./myData/")
```

Now, we can read a tree from `figure1.tre`, which is written in Newick format or we can use directly the tree from the data file.

To read a file form a file use:

```
tree.figure1 <- read.tree ("yourTree.tre")
```

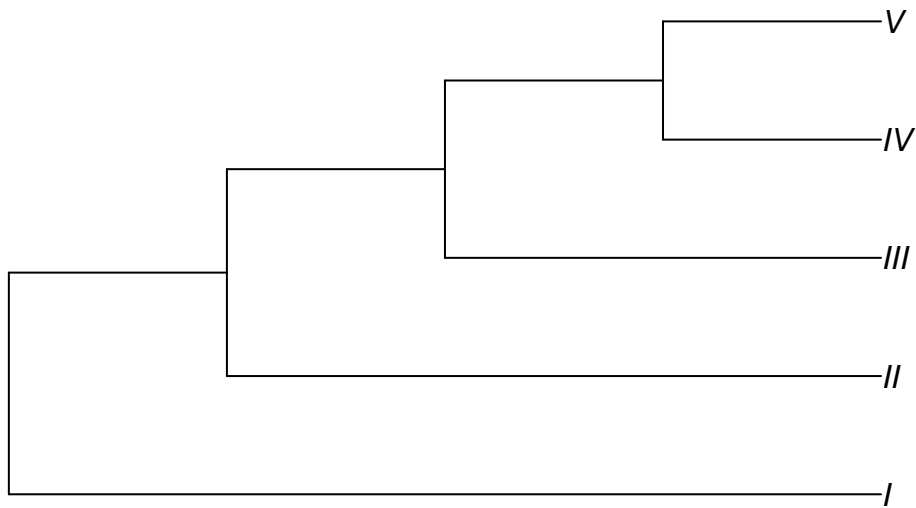
Or, upload the data set, using the data function.

```
data(tree)
```

We can plot the tree.

```
plot(tree, main= "Figure 1. Area Cladogram")
```

Figure 1. Area Cladogram



We can also read distributions using the species per area format. The distributions could be a csv file with several features: Each line has a species name and an area, and multiple areas for the same species means a widespread taxon in multiple lines.

The following function creates a data frame for the species distributions.

```
distrib.figure1 <- Read.Data("figure1.csv")
```

```
head(distrib.figure1)
```

Also, we could read the example distribution.

```
data(distribution)
```

We run the initial Index calculation, with a verbose output.

```
library(jrich)
```

```
initial.Values <- Calculate.Index(tree=tree, distrib = distribution, verbose=T)
```

```
## [1] "Deleted 0 out of 5"
```

```
initial.Values
```

```
##   area I Ie   Is   Ise   W   We   Ws   Wse rich endem jtopol jtip
## 1   A 8   8 0.200 0.364 4.000 4.000 0.158 0.300   1    1    0    0
## 2   B 4   4 0.100 0.182 2.000 2.000 0.079 0.150   1    0    0    0
## 3   C 2   2 0.050 0.091 1.333 1.333 0.053 0.100   1    0    0    0
## 4   D 1   1 0.025 0.045 1.000 1.000 0.039 0.075   1    0    0    0
## 5   E 1   1 0.025 0.045 1.000 1.000 0.039 0.075   1    0    0    0
## 6   F 8   2 0.200 0.091 5.333 1.333 0.211 0.100   4    0    0    0
## 7   G 8   2 0.200 0.091 5.333 1.333 0.211 0.100   4    0    0    0
## 8   H 8   2 0.200 0.091 5.333 1.333 0.211 0.100   4    0    0    0
```

Note that the figures for **Is/Ws** indices here are different from Figure 1 in Miranda-Esquivel 2016, as here are re-scaled to sum 1, but the proportions are exactly the same.

To obtain the same figures for **Is/Ws** indices as in Figure 1, we must use this code:

```
figure1.Values <- Calculate.Index(tree=tree, distrib = distribution, verbose=F, standard = "tree")
```

```
figure1.Values
```

```
##   area I Ie   Is   Ise   W   We   Ws   Wse rich endem jtopol jtip
## 1   A 8   8 0.500 0.500 4.000 4.000 0.429 0.429   1    1    0    0
## 2   B 4   4 0.250 0.250 2.000 2.000 0.214 0.214   1    0    0    0
## 3   C 2   2 0.125 0.125 1.333 1.333 0.143 0.143   1    0    0    0
## 4   D 1   1 0.062 0.062 1.000 1.000 0.107 0.107   1    0    0    0
## 5   E 1   1 0.062 0.062 1.000 1.000 0.107 0.107   1    0    0    0
## 6   F 8   2 0.500 0.125 5.333 1.333 0.571 0.143   4    0    0    0
## 7   G 8   2 0.500 0.125 5.333 1.333 0.571 0.143   4    0    0    0
## 8   H 8   2 0.500 0.125 5.333 1.333 0.571 0.143   4    0    0    0
```

```
all.equal(initial.Values,figure1.Values)
```

```
## [1] "Component \"Is\": Mean relative difference: 1.499"
## [2] "Component \"Ise\": Mean relative difference: 0.374"
## [3] "Component \"Ws\": Mean relative difference: 1.71029"
## [4] "Component \"Wse\": Mean relative difference: 0.429"
```

1. Correlations between values

Plot the initial Values for the index that “explains” the most with this line:

```
correlations <- cor(initial.Values[,2:10],initial.Values[,2:10])
```

Do not forget using the following code to avoid the “autocorrelation”.

```
diag(correlations) <- 0.0
```

Now we can determine the ‘Best’ descriptor:

```
best.Index <- which.max(apply(correlations,2,sum))
```

```
best.Index
```

```
## I  
## 1
```

With the following instruction we can get the name of index that displays the highest value, and the row of the column **rich** in which the object is found.

Keep in mind that richness is not a good predictor for all indices.

```
which(abs(correlations[, "rich"]) == max(abs(correlations[, "rich"])))
```

```
## Ws  
## 7
```

To plot aesthetic graphics, load the library `ggplot2`.

```
library(ggplot2)
```

2. The index that “explains the most”, without resampling is:

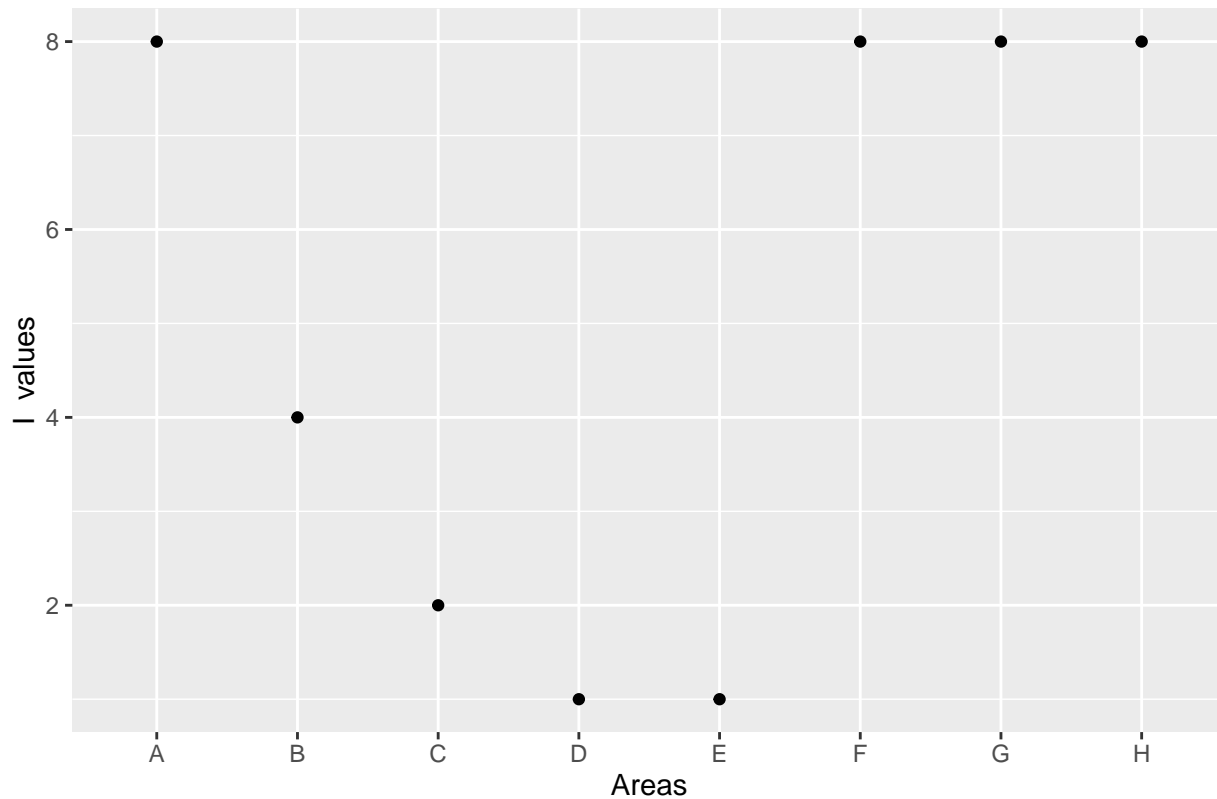
```
best.Index <- which.max(apply(correlations,2,sum))
```

```
best.Index
```

```
## I  
## 1
```

```
qplot(initial.Values$area, initial.Values[, names(best.Index)], xlab = "Areas",  
       ylab = paste(names(best.Index), " values"), main = paste("Figure 2. Values of the most informative :"))
```

Figure 2. Values of the most informative index: I Index



In this example, Areas **A** / **F** / **G** / **H** have the same ranking, as area **A** harbors species **I**, while areas **F** / **G** / **H** have the highest absolute richness.

We run the analysis with a single Jackknife replicate with a jtip value of 0.5.

```
jack.Values <- Calculate.Index(tree=tree, distrib = distribution,jtip = 0.5)
```

```
## [1] "Deleted 4 out of 5"
```

The absolute difference between these two outputs can be computed with this instruction:

```
all.equal(initial.Values, jack.Values)
```

```
## [1] "Component \"I\": Mean relative difference: 0.9230769"
## [2] "Component \"Ie\": Mean relative difference: 0.8571429"
## [3] "Component \"Is\": Mean relative difference: 0.75"
## [4] "Component \"Ise\": Mean relative difference: 1"
## [5] "Component \"W\": Mean relative difference: 0.8767056"
## [6] "Component \"We\": Mean relative difference: 0.7567305"
## [7] "Component \"Ws\": Mean relative difference: 0.6563437"
## [8] "Component \"Wse\": Mean relative difference: 1"
## [9] "Component \"rich\": Mean relative difference: 0.8125"
## [10] "Component \"endem\": Mean relative difference: 1"
## [11] "Component \"jtip\": Mean absolute difference: 0.5"
```

3. But a single replicate is not interesting, therefore we can repeat the process 100 times, using two approaches:

- **3.1** Jackknifing with a jtip value of 0.5, 100 replicates, using the `Calculate.Index` function.

```
jack.Ranking.100 <- list()
for (i in 1:100){
  ## if you want to get the output number replicate, uncomment this line:
  ##print(paste("replicate #",i), )

  jack.Ranking.100[[i]] <- as.data.frame(Rank.Indices(Calculate.Index(tree=tree, distrib = distribution
})
```

- **3.2** Comparing the whole ranking for the best index, 100 times.

```
initial.Ranking <- as.data.frame(Rank.Indices(initial.Values))

jack.Ranking.100.comparison <- NULL

for (i in 1:100){

  if(!all(jack.Ranking.100[[i]][,best.Index] == "XOX")){
    jack.Ranking.100.comparison[i] <- all.equal(initial.Ranking[,best.Index],
                                                jack.Ranking.100[[i]][,best.Index])
  }else jack.Ranking.100.comparison[i] <- 0
}
```

Get the number of hits or the extent of the Jackknife replicate to recover the initial ranking:

```
length(which(jack.Ranking.100.comparison==TRUE))
```

```
## [1] 9
```

To estimate the error, we use this code:

```
length(which(as.data.frame(jack.Ranking.100.comparison)!=TRUE))
```

```
## [1] 91
```

These two figures indicate that the best index is not **I**. As it is also important to estimate the error, we can use the following code to compute it.

```
jack.Mismatch <- jack.Ranking.100.comparison[jack.Ranking.100.comparison!=TRUE]

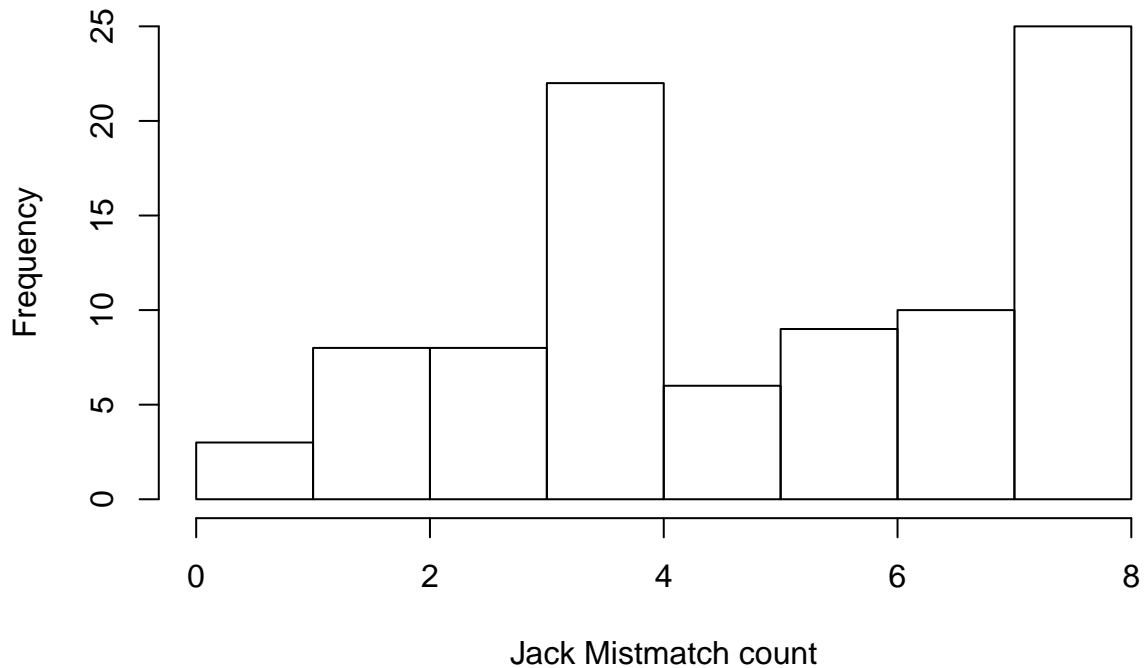
count.Jack.Mismatch <- gsub(" string mismatches","",jack.Mismatch)

count.Jack.Mismatch <- as.numeric(count.Jack.Mismatch)
```

Plot the distribution of the error, not so bell-shaped.

```
hist(sort(count.Jack.Mismatch,na.last = NA), main = "Figure 3. Histogram of Error Distribution", xlab=
```

Figure 3. Histogram of Error Distribution



4. A wrap to the previous function, and evaluating the number of times we recovered 1/2/3 position in the ranking.

Note that `Calculate.Index` recovers the index values while `Best.Index` recovers the ranking comparison.

```
jack.figure1.jtip05.100replicates <- Best.Index(tree=tree, distrib = distribution, jtip = 0.5, replicates=100)
```

```
## [1] "Deleted 0 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 4 out of 5"
## [1] "Deleted 4 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 4 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 0 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 4 out of 5"
## [1] "Deleted 5 out of 5"
## [1] "Deleted 2 out of 5"
```

```
## [1] "Deleted 3 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 4 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 4 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 3 out of 5"
## [1] "Deleted 0 out of 5"
## [1] "Deleted 1 out of 5"
## [1] "Deleted 2 out of 5"
## [1] "Deleted 4 out of 5"
```

```
jack.figure1.jtip05.100replicates
```

```
## Wse Ws We W Ise Is Ie I
## 1 4 48 28 38 4 38 28 38
```

```
best.Index = names(jack.figure1.jtip05.100replicates)[c(which(jack.figure1.jtip05.100replicates == max(
```

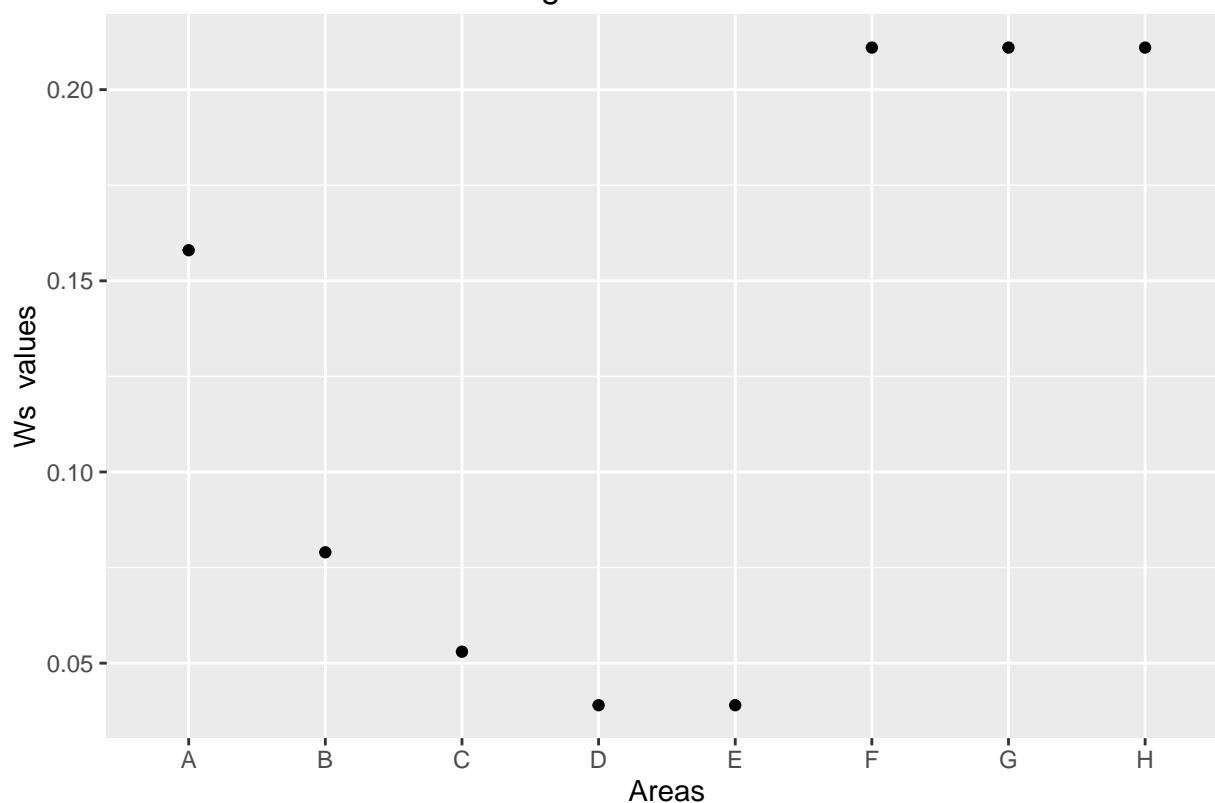
W / Ws explains better than I, as they have a Jackknife value. In this context we can plot **Ws**.

```
for (i in 1:length(best.Index)){
  print(best.Index[i])

  print(qplot(initial.Values$area,initial.Values[,best.Index[i]], xlab = "Areas",
    ylab =paste(best.Index[i], " values"), main = paste("Figure 4. ",best.Index[i], " Index")))
}
```

```
## [1] "Ws"
```


Figure 4. Ws Index



As it is shown in **Figure 4.**, areas **F/G/H** have a higher value as they have higher richness, but even so, the support is relatively low.

2. An example with two topologies and distributions

For this example, We will work with a tree and a distribution for an real taxon: *Puranius* (These two files could be found in in the data directory).

```
tree.Puranius <- read.tree ("puranius.tre")
distrib.Puranius <- Read.Data ("puranius.csv.gz")
```

Now we will assign a list to the object `data.Puranius` in order to join the tree and the distribution, which could certainly be made with a function as well. However, I prefer this approach.

```
data.Puranius <- list()
data.Puranius[[1]] <- tree.Puranius
data.Puranius[[2]] <- distrib.Puranius
head(data.Puranius)
```

We will proceed just as we did for the first Taxon, thence we got a tree and a distribution for another Taxon: *Janus*.

```

tree.Janus <- read.tree ("Janus.tre")

distrib.Janus <- Read.Data("Janus.csv.gz")

data.Janus  <- list()

data.Janus[[1]] <- tree.Janus

data.Janus[[2]] <- distrib.Janus

data.Janus

```

The following code creates a list to handle multiple datasets. In this case, the list `Multitaxon1` contains two objects.

```

Multitaxon1 <- list()

Multitaxon1[[1]] <- data.Janus

Multitaxon1[[2]] <- data.Puranus

```

The function `Multi.Index.Calc` calculates indices values for a `MultiData` list. For this case, it computes the Initial ranking for `Multitaxon1` with default values. This action is performed just once.

```

data(Multitaxon1)

initial.Values.Multi <- Multi.Index.Calc(Multitaxon1)

```

```

## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"

```

Before plotting the initial Values for the Index that explains the most, make sure that the library `ggplot2` is loaded.

```

library(ggplot2)

```

1. Correlations between values

Plotting as example, the index value that explains the most is:

```

correlations.Multi <- cor(initial.Values.Multi[,2:10], initial.Values.Multi[,2:10])

```

To avoid the highest “autocorrelation”, use this line:

```

diag(correlations.Multi) <- 0.0

```

In case we want to know the index that explains the most, use this code:

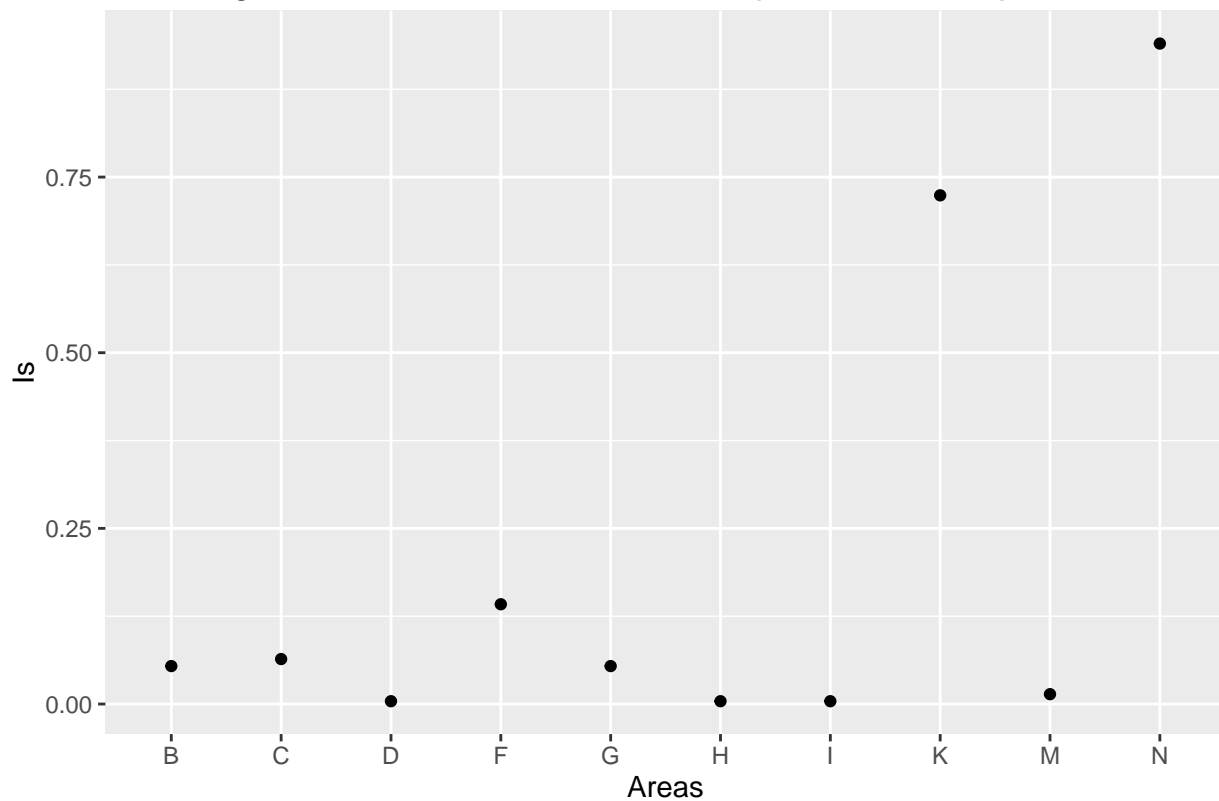
```

best.Index.Multi <- which.max(apply(correlations.Multi,2,sum))

qplot(initial.Values.Multi$area,initial.Values.Multi[,names(best.Index.Multi)], xlab = "Areas", ylab = "n")

```

Figure 5. Values of the Index that explains the most per area



2. A delete experiment: Prob jtip jtopol 0.5

```
jack.Multi <- Multi.Index.Calc(Multitaxon1,jtip = 0.5,jtopol = 0.5)
```

```
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
```

See that the jtopol and jtip present the deleted proportions, and the two data frames are different.

```
all.equal(jack.Multi,initial.Values.Multi)
```

```
## [1] "Component \"I\": Mean relative difference: 0.06126296"
## [2] "Component \"Ie\": Mean relative difference: 0.2730821"
## [3] "Component \"Is\": Mean relative difference: 0.06204757"
## [4] "Component \"Ise\": Mean relative difference: 1"
## [5] "Component \"W\": Mean relative difference: 0.2548686"
## [6] "Component \"We\": Mean relative difference: 0.2546368"
## [7] "Component \"Ws\": Mean relative difference: 0.2176088"
## [8] "Component \"Wse\": Mean relative difference: 1"
## [9] "Component \"rich\": Mean relative difference: 0.3658537"
## [10] "Component \"endem\": Mean relative difference: 0.5757576"
## [11] "Component \"jtopol\": Mean relative difference: 1"
## [12] "Component \"jtip\": Mean relative difference: 1"
```

```
vector <- which.max(apply(correlations,2,sum))
```

3. Jack Ranking Multitaxon 100 times

This operation computes the index values and returns a data matrix. Diverse from the first method, it allows to perform this operation 100 times. The number of replicates can be easily modified, by changing the rank in for (i in 1:100).

```
jack.Ranking.100 <- list()

for (i in 1:100){
  print(paste("replicate #",i))
  jack.Ranking.100[[i]] <- as.data.frame(Rank.Indices(Multi.Index.Calc(Multitaxon1,
    jtip = 0.5,jtopol = 0.5)))
}
```

```
## [1] "replicate # 1"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 15 out of 20"
## [1] "replicate # 2"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 3"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 4"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 5"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 6"
## [1] "Deleted 13 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 7"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 8"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 9"
## [1] "Deleted 14 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 10"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 11"
## [1] "Deleted 8 out of 20"
## [1] "Deleted 9 out of 20"
## [1] "replicate # 12"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 13"
```

```
## [1] "Deleted 10 out of 20"
## [1] "Deleted 7 out of 20"
## [1] "replicate # 14"
## [1] "Deleted 9 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 15"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 16"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 17"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 18"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 19"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 8 out of 20"
## [1] "replicate # 20"
## [1] "Deleted 7 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 21"
## [1] "Deleted 15 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 22"
## [1] "Deleted 8 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 23"
## [1] "Deleted 14 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 24"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 8 out of 20"
## [1] "replicate # 25"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 26"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 27"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 28"
## [1] "Deleted 7 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 29"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 30"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 31"
```

```
## [1] "Deleted 0 out of 20"
## [1] "Deleted 8 out of 20"
## [1] "replicate # 32"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 33"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 34"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 35"
## [1] "Deleted 15 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 36"
## [1] "Deleted 9 out of 20"
## [1] "Deleted 12 out of 20"
## [1] "replicate # 37"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 38"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 9 out of 20"
## [1] "replicate # 39"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 40"
## [1] "Deleted 15 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 41"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 42"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 43"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 44"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 45"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 46"
## [1] "Deleted 8 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 47"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 5 out of 20"
## [1] "replicate # 48"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 49"
```

```
## [1] "Deleted 12 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 50"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 51"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 9 out of 20"
## [1] "replicate # 52"
## [1] "Deleted 13 out of 20"
## [1] "Deleted 14 out of 20"
## [1] "replicate # 53"
## [1] "Deleted 8 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 54"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 12 out of 20"
## [1] "replicate # 55"
## [1] "Deleted 13 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 56"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 57"
## [1] "Deleted 13 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 58"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 59"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 60"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 7 out of 20"
## [1] "replicate # 61"
## [1] "Deleted 8 out of 20"
## [1] "Deleted 12 out of 20"
## [1] "replicate # 62"
## [1] "Deleted 6 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 63"
## [1] "Deleted 9 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 64"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 65"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 66"
## [1] "Deleted 6 out of 20"
## [1] "Deleted 8 out of 20"
## [1] "replicate # 67"
```

```
## [1] "Deleted 12 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 68"
## [1] "Deleted 12 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 69"
## [1] "Deleted 8 out of 20"
## [1] "Deleted 12 out of 20"
## [1] "replicate # 70"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 71"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 72"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 73"
## [1] "Deleted 12 out of 20"
## [1] "Deleted 13 out of 20"
## [1] "replicate # 74"
## [1] "Deleted 9 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 75"
## [1] "Deleted 9 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 76"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 13 out of 20"
## [1] "replicate # 77"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 78"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 79"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 13 out of 20"
## [1] "replicate # 80"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 13 out of 20"
## [1] "replicate # 81"
## [1] "Deleted 9 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 82"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 83"
## [1] "Deleted 5 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 84"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 85"
```



```
## [1] "Deleted 0 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 86"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 9 out of 20"
## [1] "replicate # 87"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 14 out of 20"
## [1] "replicate # 88"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 89"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 90"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 10 out of 20"
## [1] "replicate # 91"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 92"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 93"
## [1] "Deleted 11 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 94"
## [1] "Deleted 8 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 95"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 9 out of 20"
## [1] "replicate # 96"
## [1] "Deleted 10 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 97"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 0 out of 20"
## [1] "replicate # 98"
## [1] "Deleted 7 out of 20"
## [1] "Deleted 7 out of 20"
## [1] "replicate # 99"
## [1] "Deleted 13 out of 20"
## [1] "Deleted 11 out of 20"
## [1] "replicate # 100"
## [1] "Deleted 0 out of 20"
## [1] "Deleted 5 out of 20"
```

We can get the ranking in the first run. It is only necessary to run this line:

```
jack.Ranking.100[[1]][,vector]
```

```
## [1] N K F B G C M D H I
## Levels: B C D F G H I K M N
```

4. Convert the initial values to a ranking

The initial ranking for Multitaxon is assigned as follows:

```
initial.Ranking <- as.data.frame(Rank.Indices(Multi.Index.Calc(Multitaxon1)))
```

```
## [1] "Deleted 0 out of 20"  
## [1] "Deleted 0 out of 20"
```

```
initial.Ranking[,vector]
```

```
## [1] N K F C B G M D H I  
## Levels: B C D F G H I K M N
```

5. Compare the whole ranking for the best index, 100 times.

```
jack.Ranking.100.comparison <- NULL  
  
for (i in 1:100){  
  jack.Ranking.100.comparison[i] <- all.equal(initial.Ranking[,vector],  
                                              jack.Ranking.100[[i]][,vector])  
}
```

Compute the number of hits or the Jackknife replicate ability to recover the initial ranking:

```
length(which(jack.Ranking.100.comparison==TRUE))
```

```
## [1] 30
```

As in the first example, get an estimation of the error.

```
length(which(as.data.frame(jack.Ranking.100.comparison)!=TRUE))
```

```
## [1] 70
```

And with this instruction we can identify the type of error.

```
jack.Mismatch <- jack.Ranking.100.comparison[jack.Ranking.100.comparison!=TRUE]  
  
count.Jack.Mismatch <- gsub(" string mismatches", "", jack.Mismatch)  
  
count.Jack.Mismatch <- as.numeric(count.Jack.Mismatch)
```

This code computes the distribution of the error, which will rather be bell shaped.

```
hist(sort(count.Jack.Mismatch), main="Figure 6. Histogram of Error Distribution", xlab= "Jack Mismatch")
```

Figure 6. Histogram of Error Distribution

