# Package 'abdiv' 

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Title Alpha and Beta Diversity Measures
Version 0.2.0
Description A collection of measures for measuring ecological diversity.
Ecological diversity comes in two flavors: alpha diversity measures the diversity within a single site or sample, and beta diversity measures the diversity across two sites or samples. This package overlaps considerably with other R packages such as 'vegan', 'gUniFrac', 'betapart', and 'fossil'. We also include a wide range of functions that are implemented in software outside the R ecosystem, such as 'scipy', 'Mothur', and 'scikit-bio'. The implementations here are designed to be basic and clear to the reader.

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## $R$ topics documented:

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berger_parker_d Berger-Parker dominance

## Description

The Berger-Parker dominance is the proportion of the most abundant species.

## Usage

berger_parker_d(x)

## Arguments

[^0]
## Details

- Equivalent to berger_parker_d() in skbio.diversity.alpha.
- Equivalent to the bergerparker calculator in Mothur.


## Value

The Berger-Parker dominance, $0<D_{B P} \leq 1$. If the vector sums to zero, the Berger-Parker dominance is undefined, and we return NaN .

## References

Berger WH, Parker FL. Diversity of Planktonic Foraminifera in Deep-Sea Sediments. Science. 1970;168(3937):1345-1347.

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
berger_parker_d(x) # 15 / 28
```

binomial_deviance Binomial deviance and CY index of dissimilarity

## Description

The binomial deviance dissimilarity and the CY (or Cao ) index of dissimilarity were created to compare species counts at sites with moderate to large differences.

## Usage

binomial_deviance(x, y)
cy_dissimilarity(x, y, base = 10, min_value = 0.1)

## Arguments

| $\mathrm{x}, \mathrm{y}$ | Numeric vectors |
| :--- | :--- |
| base | Base of the logarithm |
| min_value | Replacement for zero or near-zero values. Values less than min_value are re- <br> placed with min_value. |

## Details

Both of these measures were designed to be used with whole-numbered counts, and may not make sense for comparing normalized vectors or vectors of species proportions.

For two vectors $x$ and $y$, the binomial deviance dissimilarity is

$$
d(x, y)=\sum_{i} \frac{1}{n_{i}}\left(x_{i} \log \frac{x_{i}}{n_{i}}+y_{i} \log \frac{y_{i}}{n_{i}}-\left(x_{i}+y_{i}\right) \log 2\right)
$$

where $n_{i}=x_{i}+y_{i}$. This value is the weighted average of the deviance for each species, under a binomial model where the expected counts are $n_{i} / 2$ at each site. It was proposed by Anderson and Millar in 2004. Relation to other definitions:

- Equivalent to vegdist() with method = "binomial".

The CY index was proposed by Cao, Williams, and Bark in 1997. For two vectors $x$ and $y$, the CY index is

$$
d(x, y)=\frac{1}{N} \sum_{i}\left(\frac{\left(x_{i}+y_{i}\right) \log _{10}\left(\frac{x_{i}+y_{i}}{2}\right)-x_{i} \log _{10}\left(y_{i}\right)-y_{i} \log _{10}\left(x_{i}\right)}{x_{i}+y_{i}}\right)
$$

where $N$ is the total number of species in vectors $x$ and $y$. Double zeros are not considered in the measure.

When either $x_{i}$ or $y_{i}$ are zero, they need to be replaced by another value in the CY index to avoid infinities. Cao suggested replacing zero values with 0.1 , which is one log lower than the minimum value for whole-numbered counts. Here, we use a min_value argument to allow the user set a lower limit on the values. For vectors of species counts, this function follows the formulation of Cao by default.

Relation of the CY index to other definitions:

- Equivalent to the vegdist() function with method $=$ "cao", if base $=\exp (1)$.


## Value

The Binomial deviance or CY index of dissimilarity. The CY index is undefined if all elements of $x$ and $y$ are zero, in which case we return NaN .

## References

Anderson MJ, Millar RB. Spatial variation and effects of habitat on temperate reef fish assemblages in northeastern New Zealand. Journal of Experimental Marine Biology and Ecology 2004;305:191-221.

Cao Y, Williams WP, Bark AW. Similarity measure bias in river benthic Aufwuchs community analysis. Water Environment Research 1997;69(1):95-106.

## Description

The Bray-Curtis distance is the Manhattan distance divided by the sum of both vectors.

## Usage

bray_curtis(x, y)

## Arguments

$x, y \quad$ Numeric vectors

## Details

For two vectors $x$ and $y$, the Bray-Curtis distance is defined as

$$
d(x, y)=\frac{\sum_{i}\left|x_{i}-y_{i}\right|}{\sum_{i} x_{i}+y_{i}}
$$

The Bray-Curtis distance is connected to many other distance measures in this package; we try to list some of the more important connections here. Relation to other definitions:

- Equivalent to vegdist() with method = "bray".
- Equivalent to the braycurtis() function in scipy.spatial. distance for positive vectors. They take the absolute value of $x_{i}+y_{i}$ in the denominator.
- Equivalent to the braycurtis and odum calculators in Mothur.
- Equivalent to $D_{14}=1-S_{17}$ in Legendre \& Legendre.
- The Bray-Curtis distance on proportions is equal to half the Manhattan distance.
- The Bray-Curtis distance on presence/absence vectors is equal to the Sorenson index of dissimilarity.


## Value

The Bray-Curtis distance between $x$ and $y$. The Bray-Curtis distance is undefined if the sum of all elements in $x$ and $y$ is zero, in which case we return NaN .

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
y <- c(10, 2, 0, 1, 1, 0)
bray_curtis(x, y)
# For proportions, equal to half the Manhattan distance
bray_curtis(x / sum(x), y / sum(y))
manhattan(x / sum(x), y / sum(y)) / 2
```

```
bray_curtis_components
```

Balanced variation and abundance gradient components for abundance data

## Description

Balanced variation and abundance gradient components for abundance data

## Usage

bray_curtis_balanced(x, y)
bray_curtis_gradient(x, y)
ruzicka_balanced(x, y)
ruzicka_gradient(x, y)

## Arguments

$\mathrm{x}, \mathrm{y}$
Numeric vectors

## Value

The balanced variation or abundance gradient component of distance between $x$ and $y$. This quantity is undefined when either $x$ or $y$ have all elements equal to zero, in which case we return NaN .

## References

Baselga A. Separating the two components of abundance-based dissimilarity: balanced changes in abundance vs.abundance gradients. Methods in Ecology and Evolution. 2013;4:552-557.
Baselga A. Partitioning abundance-based multiple-site dissimilarity into components: balanced variation in abundance and abundance gradients. Methods in Ecology and Evolution. 2017;8:799-808.

## Description

The Canberra distance and Clark's coefficient of divergence are measures that use the absolute difference over the sum for each element of the vectors.

## Usage

canberra(x, y)
clark_coefficient_of_divergence(x, y)

## Arguments

$x, y \quad$ Numeric vectors

## Details

For vectors x and y , the Canberra distance is defined as

$$
d(x, y)=\sum_{i} \frac{\left|x_{i}-y_{i}\right|}{x_{i}+y_{i}}
$$

Elements where $x_{i}+y_{i}=0$ are not included in the sum. Relation of canberra() to other definitions:

- Equivalent to R's built-in dist() function with method = "canberra".
- Equivalent to the vegdist() function with method = "canberra", multiplied by the number of entries where $x>0, y>0$, or both.
- Equivalent to the canberra() function in scipy.spatial.distance for positive vectors. They take the absolute value of $x_{i}$ and $y_{i}$ in the denominator.
- Equivalent to the canberra calculator in Mothur, multiplied by the total number of species in $x$ and $y$.
- Equivalent to $D_{10}$ in Legendre \& Legendre.

Clark's coefficient of divergence involves summing squares and taking a square root afterwards:

$$
d(x, y)=\sqrt{\frac{1}{n} \sum_{i}\left(\frac{x_{i}-y_{i}}{x_{i}+y_{i}}\right)^{2}}
$$

where $n$ is the number of elements where $x>0, y>0$, or both. Relation of clark_coefficient_of_divergence() to other definitions:

- Equivalent to $D_{11}$ in Legendre \& Legendre.


## Value

The Canberra distance or Clark's coefficient of divergence. If every element in $x$ and $y$ is zero, Clark's coefficient of divergence is undefined, and we return NaN .

## Examples

```
x<- c(15, 6, 4, 0, 3, 0)
y<-c(10, 2, 0, 1, 1, 0)
canberra(x, y)
clark_coefficient_of_divergence(x, y)
```


## Description

The Chebyshev distance is the maximum absolute difference between the vector elements.

## Usage

chebyshev ( $\mathrm{x}, \mathrm{y}$ )

## Arguments

$$
x, y \quad \text { Numeric vectors }
$$

## Details

For vectors x and y , the Chebyshev distance is defined as

$$
d(x, y)=\max _{i}\left|x_{i}-y_{i}\right|
$$

Relation to other definitions:

- Equivalent to the chebyshev() function in scipy. spatial. distance.


## Value

The Chebyshev distance between $x$ and $y$.

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
y <- c(10, 2, 0, 1, 1, 0)
chebyshev(x, y) # should be 5
```

```
correlation_distance Correlation and cosine distance
```


## Description

The correlation and cosine distances, which are derived from the dot product of the two vectors.

## Usage

correlation_distance(x, y)
cosine_distance(x, y)

## Arguments

$$
x, y \quad \text { Numeric vectors }
$$

## Details

For vectors $x$ and $y$, the cosine distance is defined as the cosine of the angle between the vectors,

$$
d(x, y)=1-\frac{x \cdot y}{|x||y|}
$$

where $|x|$ is the magnitude or L2 norm of the vector, $|x|=\sqrt{\sum_{i} x_{i}^{2}}$. Relation to other definitions:

- Equivalent to the cosine() function in scipy. spatial. distance.

The correlation distance is simply equal to one minus the Pearson correlation between vectors. Mathematically, it is equivalent to the cosine distance between the vectors after they are centered $(x-\bar{x})$. Relation to other definitions:

- Equivalent to the correlation() function in scipy. spatial.distance.
- Equivalent to the 1 -mempearson calculator in Mothur.


## Value

The correlation or cosine distance. These are undefined if either x or y contain all zero elements, that is, if $|x|=0$ or $|y|=0$. In this case, we return NaN .

## Examples

```
\(x<-c(2,0)\)
\(y<-c(5,5)\)
cosine_distance( \(\mathrm{x}, \mathrm{y}\) )
\# The two vectors form a 45 degree angle, or pi / 4
\(1-\cos (\mathrm{pi} / 4)\)
\(v<-c(3.5,0.1,1.4)\)
w <- c(3.3, 0.5, 0.9)
correlation_distance(v, w)
1 - cor (v, w)
```

diversity_measures Diversity measures implemented

## Description

The diversity functions offered in abdiv are organized based on the function signature.

## Usage

alpha_diversities
beta_diversities
phylogenetic_alpha_diversities
phylogenetic_beta_diversities

## Format

Four objects of class character.

## Details

The following character vectors are provided:
alpha_diversities All non-phylogenetic alpha diversity measures. These functions take a single numeric vector as an argument.
beta_diversities All non-phylogenetic beta diversity measures. These functions take two numeric vectors as arguments.
phylogenetic_alpha_diversities There is only one phylogenetic alpha diversity measure implemented, but we use the plural to be consistent with the other vectors. This function takes a numeric vector, a phylogenetic tree object, and optionally a character vector of species labels.
phylogenetic_beta_diversities Phylogenetic measures of beta diversity. These functions take two numeric vectors, a phylogenetic tree object, and optionally a character vector of species labels.

```
euclidean
```

Euclidean and related distances

## Description

These distance and diversity measures are mathematically similar to the Euclidean distance between two vectors.

## Usage

euclidean ( $\mathrm{x}, \mathrm{y}$ )
rms_distance(x, y)
$\operatorname{chord}(\mathrm{x}, \mathrm{y})$
hellinger (x, y)
geodesic_metric(x, y)
euclidean

## Arguments

$x, y \quad$ Numeric vectors

## Details

For vectors $x$ and $y$, the Euclidean distance is defined as

$$
d(x, y)=\sqrt{\sum_{i}\left(x_{i}-y_{i}\right)^{2}}
$$

Relation of euclidean() to other definitions:

- Equivalent to R's built-in dist() function with method = "euclidean".
- Equivalent to vegdist() with method = "euclidean".
- Equivalent to the euclidean() function in scipy. spatial.distance.
- Equivalent to the structeuclidean calculator in Mothur, to speciesprofile if $x$ and $y$ are transformed to relative abundance, and to memeuclidean if $x$ and $y$ are transformed to presence/absence.
- Equivalent to $D_{1}$ in Legendre \& Legendre.
- Equivalent to the distance between species profiles, $D_{18}$ in Legendre \& Legendre if x and y are transformed to relative abundance.

The root-mean-square distance or average distance is similar to Euclidean distance. As the name implies, it is computed as the square root of the mean of the squared differences between elements of $x$ and $y$ :

$$
d(x, y)=\sqrt{\frac{1}{n} \sum_{i}^{n}\left(x_{i}-y_{i}\right)^{2}}
$$

Relation of rms_distance() to other definitions:

- Equivalent to $D_{2}$ in Legendre \& Legendre.

The chord distance is the Euclidean distance after scaling each vector by its root sum of squares, $\sqrt{\sum_{i} x_{i}^{2}}$. The chord distance between any two vectors ranges from 0 to $\sqrt{2}$. Relation of chord() to other definitions:

- Equivalent to $D_{3}$ in Legendre \& Legendre.

The Hellinger distance is equal to the chord distance computed after a square-root transformation. Relation of hellinger () to other definitions:

- Equivalent to $D_{17}$ in Legendre \& Legendre.
- Equivalent to the hellinger calculator in Mothur.

The geodesic metric is a transformed version of the chord distance.

$$
d(x, y)=\arccos \left(1-\frac{d_{c}^{2}(x, y)}{2}\right)
$$

where $d_{c}$ is the chord distance. It gives the length of the arc on a hypersphere between the vectors, if the vectors are normalized to unit length. Relation of geodesic_metric() to other definitions:

- Equivalent to $D_{4}$ in Legendre \& Legendre.


## Value

The distance between $x$ and $y$. The chord distance, Hellinger distance, and geodesic metric are not defined if all elements of either vector are zero. We return NaN in this case.

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
y <- c(10, 2, 0, 1, 1, 0)
euclidean(x, y)
# The "distance between species profiles"
euclidean(x / sum(x), y / sum(y))
rms_distance(x, y)
chord(x, y)
hellinger(x, y)
# Hellinger is chord distance after square root transform
chord(sqrt(x), sqrt(y))
geodesic_metric(x, y)
# No species in common with x
v <- c(0, 0, 0, 5, 0, 5)
chord(v, x)
sqrt(2)
```

faith_pd Faith's phylogenetic diversity

## Description

Faith's phylogenetic diversity gives the total branch length on a phylogenetic tree that is spanned by a community. The abundance of each species in the community is not considered.

## Usage

faith_pd(x, tree, x_labels = NULL)

## Arguments

$x \quad$ A numeric vector of species counts or proportions, or a logical vector of species presence/absence.
tree A phylogenetic tree object..
x_labels A character vector of species labels for x .

## Details

If the vector $x$ is named, the names will be automatically used to match $x$ with the tree. Missing names are filled in with zero counts. If $x$ is not named and $x_{-} l a b e l s$ is provided, these labels are used to match the elements of $x$ with the tree. If $x$ is not named and $x \_l a b e l s$ is not provided, it is assumed that $x$ is already in the correct order, and we simply check that its length matches the number of tips in the tree.

## Value

Faith's phylogenetic diversity, $P D \geq 0$.

## References

Faith DP. Conservation evaluation and phylogenetic diversity. Biol. Conserv. 1992;61:1-10. doi: 10.1016/0006-3207(92)91201-3.

## Examples

```
# Faith's phylogenetic diversity for whole tree is equal to the sum of the
# branch lengths.
sum(faith_tree$edge.length)
faith_pd(c(1, 1, 1, 1, 1), faith_tree)
# Can use named vector or additional argument to match species to tree.
faith_tree$tip.label
faith_pd(c(0, 0, 0, 10, 12), faith_tree)
faith_pd(c(d=10, e=12), faith_tree)
faith_pd(c(10, 12), faith_tree, c("d", "e"))
```

faith_tree Example data for Faith's phylogenetic diversity

## Description

This example was used to illustrate phylogenetic diversity in Faith and Richards (2012).

## Usage

faith_tree

## Format

faith_tree is a phylogenetic tree with five tips, labeled a-e. It was created with the ape library.

## Details

In the paper, they give the total branch length as 41, but they don't assign a length to the branch leading to species " $b$ " and " c ". Looking at the figure, we estimated that the length should be 4 . For that reason, the total branch length of faith_tree is 45 , rather than 41 .

## Source

Faith DP, Richards ZT. Biology (Basel). 2012;1(3):906-32. 10.3390/biology1030906
hamming Hamming distance

## Description

The Hamming distance is the number of positions where the values are different.

## Usage

hamming ( $\mathrm{x}, \mathrm{y}$ )

## Arguments

$x, y \quad$ Numeric vectors.

## Details

For vectors $x$ and $y$, the Hamming distance is defined as

$$
d(x, y)=\sum_{i}\left[x_{i} \neq y_{i}\right]
$$

where the quantity in the brackets is 1 if the elements are not equal, and zero if the elements are equal. Relation to other definitions:

- The hamming() function in scipy. spatial. distance divides the result by the vector length. Our function is equivalent to the SciPy version multiplied by the vector length.
- Equivalent to the hamming calculator in Mothur for presence/absence vectors.


## Value

The Hamming distance between $x$ and $y$.

```
jaccard Beta diversity for presence/absence data
```


## Description

These functions transform the input vectors to binary or presence/absence format, then compute a distance or dissimilarity.

## Usage

jaccard (x, y)
sorenson(x, y)
kulczynski_first(x, y)
kulczynski_second(x, y)
rogers_tanimoto(x, y)
russel_rao(x, y)
sokal_michener (x, y)
sokal_sneath (x, y)
yule_dissimilarity (x, y)

## Arguments

$\mathrm{x}, \mathrm{y} \quad$ Numeric vectors

## Details

Many of these indices are covered in Koleff et al. (2003), so we adopt their notation. For two vectors $x$ and $y$, we define three quantities:

- $a$ is the number of species that are present in both x and y ,
- $b$ is the number of species that are present in $y$ but not $x$,
- $c$ is the number of species that are present in x but not y , and
- $d$ is the number of species absent in both vectors.

The quantity $d$ is seldom used in ecology, for good reason. For details, please see the discussion on the "double zero problem," in section 2 of chapter 7.2 in Legendre \& Legendre.
The Jaccard index of dissimilarity is $1-a /(a+b+c)$, or one minus the proportion of shared species, counting over both samples together. Relation of jaccard() to other definitions:

- Equivalent to R's built-in dist() function with method = "binary".
- Equivalent to vegdist() with method = "jaccard" and binary = TRUE.
- Equivalent to the jaccard() function in scipy.spatial.distance, except that we always convert vectors to presence/absence.
- Equivalent to $1-S_{7}$ in Legendre \& Legendre.
- Equivalent to $1-\beta_{j}$, as well as $\beta_{c c}$, and $\beta_{g}$ in Koleff (2003).

The Sørenson or Dice index of dissimilarity is $1-2 a /(2 a+b+c)$, or one minus the average proportion of shared species, counting over each sample individually. Relation of sorenson() to other definitions:

- Equivalent to the dice() function in scipy.spatial. distance, except that we always convert vectors to presence/absence.
- Equivalent to the sorclass calculator in Mothur, and to 1 -whittaker.
- Equivalent to $D_{13}=1-S_{8}$ in Legendre \& Legendre.
- Equivalent to $1-\beta_{s o r}$ in Koleff (2003). Also equivalent to Whittaker's beta diversity (the second definition, $\left.\beta_{w}=(S / \bar{a})-1\right)$, as well as $\beta_{-1}, \beta_{t}, \beta_{m e}$, and $\beta_{h k}$.
I have not been able to track down the original reference for the first and second Kulczynski indices, but we have good formulas from Legendre \& Legendre. The first Kulczynski index is $1-a /(b+c)$, or one minus the ratio of shared to unshared species.
Relation of kulczynski_first to other definitions:
- Equivalent to $1-S_{12}$ in Legendre \& Legendre.
- Equivalent to the kulczynski calculator in Mothur.

Some people refer to the second Kulczynski index as the Kulczynski-Cody index. It is defined as one minus the average proportion of shared species in each vector,

$$
d=1-\frac{1}{2}\left(\frac{a}{a+b}+\frac{a}{a+c}\right)
$$

Relation of kulczynski_second to other definitions:

- Equivalent to $1-S_{13}$ in Legendre \& Legendre.
- Equivalent to the kulczynskicody calculator in Mothur.
- Equivalent to one minus the Kulczynski similarity in Hayek (1994).
- Equivalent to vegdist() with method = "kulczynski" and binary = TRUE.

The Rogers-Tanimoto distance is defined as $(2 b+2 c) /(a+2 b+2 c+d)$. Relation of rogers_tanimoto() to other definitions:

- Equivalent to the rogerstanimoto() function in scipy. spatial. distance, except that we always convert vectors to presence/absence.
- Equivalent to $1-S_{2}$ in Legendre \& Legendre.

The Russel-Rao distance is defined $(b+c+d) /(a+b+c+d)$, or the fraction of elements not present in both vectors, counting double absences. Relation of russel_rao() to other definitions:

- Equivalent to the russelrao() function in scipy.spatial. distance, except that we always convert vectors to presence/absence.
- Equivalent to $1-S_{11}$ in Legendre \& Legendre.

The Sokal-Michener distance is defined as $(2 b+2 c) /(a+2 b+2 c+d)$. Relation of sokal_michener () to other definitions:

- Equivalent to the sokalmichener() function in scipy.spatial. distance, except that we always convert vectors to presence/absence.
The Sokal-Sneath distance is defined as $(2 b+2 c) /(a+2 b+2 c)$. Relation of sokal_sneath() to other definitions:
- Equivalent to the sokalsneath() function in scipy.spatial.distance, except that we always convert vectors to presence/absence.
- Equivalent to the anderberg calculator in Mothur.
- Equivalent to $1-S_{10}$ in Legendre \& Legendre.

The Yule dissimilarity is defined as $2 b c /(a d+b c)$. Relation of yule_dissimilarity() to other definitions:

- Equivalent to the yule() function in scipy.spatial. distance, except that we always convert vectors to presence/absence.
- Equivalent to $1-S$, where $S$ is the Yule coefficient in Legendre \& Legendre.


## Value

The dissimilarity between $x$ and $y$, based on presence/absence. The Jaccard, Sorenson, SokalSneath, Yule, and both Kulczynski dissimilarities are not defined if both $x$ and $y$ have no nonzero elements. In addition, the second Kulczynski index and the Yule index of dissimilarity are not defined if one of the vectors has no nonzero elements. We return NaN for undefined values.
jaccard_components Nestedness and turnover components for presence/absence data

## Description

Nestedness and turnover components for presence/absence data

## Usage

jaccard_turnover (x, y)
jaccard_nestedness(x, y)
sorenson_turnover(x, y)
sorenson_nestedness(x, y)

## Arguments

$$
x, y \quad \text { Numeric vectors }
$$

## Value

The nestedness or turnover component of distance between $x$ and $y$. This quantity is undefined when either $x$ or $y$ have no observations, in which case we return NaN .

## References

Baselga A. Partitioning the turnover and nestedness components of beta diversity. Global Ecol. Biogeogr. 2010;19:134-143.
Baselga A. The relationship between species replacement, dissimilarity derived from nestedness, and nestedness. Global Ecol. Biogeogr. 2012;21:1223-1232.

```
kempton_taylor_q Kempton-Taylor Q index
```


## Description

The Kempton-Taylor Q index is designed to measure species in the middle of the abundance distribution.

## Usage

kempton_taylor_q(x, lower_quantile $=0.25$, upper_quantile $=0.75$ )

## Arguments

$x \quad$ A numeric vector of species counts or proportions.
lower_quantile, upper_quantile
Lower and upper quantiles of the abundance distribution. Default values are the ones suggested by Kempton and Taylor.

## Details

For a vector of species counts $x$, the Kempton-Taylor $Q$ statistic is equal to the slope of the cumulative abundance curve across a specified quantile range. The cumulative abundance curve is the plot of the number of species against the log-abundance.

Kempton and Taylor originally defined the index as

$$
Q=\frac{\frac{1}{2} S}{\log R_{2}-\log R_{1}}
$$

where $S$ is the total number of species observed, $R_{1}$ is the abundance at the lower quantile, and $R_{2}$ is the abundance at the upper quantile. However, this definition only holds if one uses the interquartile range. Because we allow the user to adjust the upper and lower quantiles, we have to find the number of species at these abundance values. Here, we follow the implementation in scikit-bio and round inwards to find the quantile values, taking the number of species and log-abundance values at these data points exactly.

- Equivalent to kempton_taylor_q() in skbio.diversity.alpha.
- Similar to the qstat calculator in Mothur. Our implementation differs slightly, and this difference affects the result.


## Value

The Kempton-Taylor Q index, $Q<0$. If the vector sums to zero, we cannot compute the quantiles, and this index is undefined. In that case, we return NaN.

## References

Kempton RA, Taylor LR. Models and statistics for species diversity. Nature. 1976;262:818-820.

```
kullback_leibler_divergence
```

                                    Kullback-Leibler divergence
    
## Description

Kullback-Leibler divergence

## Usage

kullback_leibler_divergence(x, y)

## Arguments

$x, y \quad$ Numeric vectors representing probabilities

## Details

Kullback-Leibler divergence is a non-symmetric measure of difference between two probability vectors. In general, $\operatorname{KL}(\mathrm{x}, \mathrm{y})$ is not equal to $\operatorname{KL}(\mathrm{y}, \mathrm{x})$.

Because this measure is defined for probabilities, the vectors x and y are normalized in the function so they sum to 1 .

## Value

The Kullback-Leibler divergence between x and y . We adopt the following conventions if elements of x or y are zero: $0 \log \left(0 / y_{i}\right)=0,0 \log (0 / 0)=0$, and $x_{i} \log \left(x_{i} / 0\right)=\infty$. As a result, if elements of $x$ are zero, they do not contribute to the sum. If elements of $y$ are zero where $x$ is nonzero, the result will be Inf. If either $x$ or $y$ sum to zero, we are not able to compute the proportions, and we return NaN .

## Description

This tree was used in Figure 2 of Leprieur et al. (2005) to demonstrate the nestedness and turnover components of phylogenetic beta diversity.

## Usage

leprieur_tree

## Format

leprieur_tree is a phylogenetic tree with 8 tips, labeled a-h. It was created with the ape library. All edges (branches) in the tree are of length 1.

## Source

Leprieur F, Albouy C, De Bortoli J, Cowman PF, Bellwood DR, Mouillot D. Quantifying phylogenetic beta diversity: distinguishing between "true" turnover of lineages and phylogenetic diversity gradients. PLoS One. 2012;7(8):e42760. 10.1371/journal.pone. 0042760
lozupone_tree Example data for UniFrac distance

## Description

This example was used to illustrate unweighted UniFrac distance in Lozupone and Knight (2005).

## Usage

lozupone_tree
lozupone_panel_a
lozupone_panel_b

## Format

lozupone_tree is a phylogenetic tree with 14 tips, labeled A-N. It was created with the ape library. The data frames lozupone_panel_a and lozupone_panel_b are transcribed from Figure 1 of the paper. They have the following columns:

Species The species label, matching to the phylogenetic tree.
SampleID The community, either "Circle" or "Square".
Counts The number of organisms counted per species, always 1 for this example.

## Source

Lozupone C, Knight R. UniFrac: a new phylogenetic method for comparing microbial communities. Applied and environmental microbiology. 2005;71(12):8228-8235. 10.1128/AEM.71.12.82288235.2005
manhattan Manhattan and related distances

## Description

The Manhattan or city block distance is the sum of absolute differences between the elements of two vectors. The mean character difference is a closely related measure.

## Usage

manhattan ( $\mathrm{x}, \mathrm{y}$ )
mean_character_difference(x, y)
modified_mean_character_difference(x, y)

## Arguments

$\mathrm{x}, \mathrm{y} \quad$ Numeric vectors

## Details

For vectors $x$ and $y$, the Manhattan distance is given by

$$
d(x, y)=\sum_{i}\left|x_{i}-y_{i}\right|
$$

Relation of manhattan() to other definitions:

- Equivalent to R's built-in dist() function with method = "manhattan".
- Equivalent to vegdist() with method = "manhattan".
- Equivalent to the cityblock() function in scipy.spatial. distance.
- Equivalent to the manhattan calculator in Mothur.
- Equivalent to $D_{7}$ in Legendre \& Legendre.
- Whittaker's index of association ( $D_{9}$ in Legendre \& Legendre) is the Manhattan distance computed after transforming to proportions and dividing by 2.

The mean character difference is the Manhattan distance divided by the length of the vectors. It was proposed by Cain and Harrison in 1958. Relation of mean_character_difference() to other definitions:

- Equivalent to $D_{8}$ in Legendre \& Legendre.
- For binary data, equivalent to $1-S_{1}$ in Legendre \& Legendre, where $S_{1}$ is the simple matching coefficient.

The modified mean character difference is the Manhattan distance divided by the number elements where either $x$ or $y$ (or both) are nonzero. Relation of modified_mean_character_difference() to other definitions:

- Equivalent to $D_{19}$ in Legendre \& Legendre.
- Equivalent to vegdist() with method = "altGower".
- For binary data, it is equivalent to the Jaccard distance.


## Value

The distance between $x$ and $y$. The modified mean character difference is undefined if all elements in $x$ and $y$ are zero, in which case we return NaN .

## References

Cain AJ, Harrison GA. An analysis of the taxonomist's judgment of affinity. Proceedings of the Zoological Society of London 1958;131:85-98.

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
y<- c(10, 2, 0, 1, 1, 0)
manhattan(x, y)
# Whittaker's index of association
manhattan(x / sum(x), y / sum(y)) / 2
mean_character_difference(x, y)
# Simple matching coefficient for presence/absence data
# Should be 2 / 6
mean_character_difference(x > 0, y > 0)
modified_mean_character_difference(x, y)
# Jaccard distance for presence/absence data
modified_mean_character_difference(x > 0, y > 0)
jaccard(x, y)
```

margalef Margalef's richness index

## Description

Margalef's richness index

## Usage

margalef(x)

## Arguments

$x \quad$ A numeric vector of species counts.

## Details

For a vector x of species counts, Margalef's index is

$$
D=\frac{S-1}{\log N}
$$

where $S$ is the total number of species observed and $N$ is the total number of counts.
This index is appropriate only for raw counts, not transformed counts or proportions.
Equivalent to margalef() in skbio.diversity.alpha.

## Value

The value of Margalef's index, $D \geq 0$. This index is undefined when the total number of counts is 1 or 0 , in which case we return NaN .

## References

Margalef R. Information theory in ecology. General Systems 3. 1958;36-71.

## Examples

```
x<-c(15, 6, 4, 0, 3, 0)
margalef(x)
```

match_to_tree Match vector of counts to phylogenetic tree

## Description

Match vector of counts to phylogenetic tree

## Usage

match_to_tree (x, tree, x_labels = NULL)

## Arguments

x
tree
x_labels

A vector of species counts.
A phylogenetic tree of class "phylo".
A vector of species labels for $x$.

## Details

This function applies a couple of different methods to arrange the data in $x$ to match a phylogenetic tree.

- If x_labels is provided, we use this vector to match the elements of $x$ with the tip labels in the tree.
- If $x \_l a b e l s$ is not provided and $x$ is a named vector, we use the names to match the tip labels in the tree.
- If $x$ _labels is not provided and $x$ is not named, we assume that $x$ is already in the correct order, check that the length of $x$ matches the number of tips in the tree, and return $x$.


## Value

The vector x , re-arranged to match the tree.

```
mcintosh_d McIntosh dominance index D
```


## Description

McIntosh dominance index D

## Usage

mcintosh_d(x)

## Arguments

$x \quad$ A numeric vector of species counts.

## Details

For a vector x of raw species counts, the McIntosh dominance index is defined as

$$
D=\frac{N-U}{N-\sqrt{N}}
$$

where $N$ is the total number of counts and $U=\sqrt{\sum_{i} x_{i}^{2}}$.
This index is appropriate only for raw counts, not transformed counts or proportions.
Equivalent to mcintosh_d() in skbio.diversity.alpha.

## Value

The McIntosh dominance index, $0 \leq D<1$. The index is undefined when the total number of counts is 1 or 0 , in which case we return NaN .
mcintosh_e

## References

McIntosh RP. An index of diversity and the relation of certain concepts to diversity. Ecology. 1967;48:1115-1126.

## Examples

$x<-c(15,6,4,0,3,0)$
mcintosh_d(x)
mcintosh_e McIntosh's evenness measure E

## Description

McIntosh's evenness measure E

## Usage

mcintosh_e(x)

## Arguments

$x \quad$ A numeric vector of species counts.

## Details

For a vector x of raw species counts, the McIntosh evenness measure is

$$
E=\frac{\sqrt{\sum_{i} x_{i}^{2}}}{\sqrt{(N-S+1)^{2}+S-1}}
$$

where $N$ is the total number of counts and $S$ is the total number of species observed. This index is appropriate only for raw counts, not transformed counts or proportions.
Equivalent to mcintosh_e() in skbio.diversity.alpha.

## Value

McIntosh's evenness measure, $0<E \leq 1$. The index is undefined when the total number of counts is 0 , in which case we return NaN .

## References

Heip C, Engels P. Comparing Species Diversity and Evenness Indices. J. Mar. Bioi. Ass. U.K. 1974;54:559-563.

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
mcintosh_e(x)
```

```
    menhinick Menhinick's richness index
```


## Description

Menhinick's richness index

## Usage

menhinick(x)

## Arguments

$x \quad$ A numeric vector of species counts.

## Details

For a vector x of raw species counts, the Menhinick's richness index is $\frac{S}{\sqrt{N}}$, where $N$ is the total number of counts and $S$ is the total number of species observed.
This index is appropriate only for raw counts, not transformed counts or proportions.
Equivalent to menhinick() in skbio. diversity. alpha.

## Value

Menhinick's richness index, $R>0$. The index is undefined when the total number of counts is 0 , in which case we return NaN .

## Examples

```
x<-c(15, 6, 4, 0, 3, 0)
menhinick(x)
```

minkowski Minkowski distance

## Description

The Minkowski metric is a generalized form of Euclidean ( $\mathrm{p}=2$ ) and Manhattan ( $\mathrm{p}=1$ ) distance.

## Usage

minkowski(x, y, $p=1$ )

## Arguments

$x, y$
Numeric vectors.
p
Exponent parameter, a single number greater than zero.

## Details

For vectors $x$ and $y$, the Minkowski distance is defined as

$$
d(x, y)=\left(\sum_{i}\left|x_{i}-y_{i}\right|^{p}\right)^{1 / p}
$$

Relation to other definitions:

- Equivalent to R's built-in dist() function with method = "minkowski".
- Equivalent to the minkowski() function in scipy. spatial. distance.
- Equivalent to $D_{6}$ in Legendre \& Legendre.

The default value of $p=1$ makes this distance equal to the Manhattan distance.

## Value

The Minkowski distance between $x$ and $y$.

The Morisita index and Horn-Morisita index

## Description

The Morisita and the Horn-Morisita indices measure the probability that individuals drawn one from each vector will belong to different species, relative to drawing from each vector separately. The Morisita index is formulated for count data only, whereas the Horn-Morisita index can be used with transformed counts or proportions.

## Usage

morisita(x, y)
horn_morisita(x, y)

## Arguments

$x, y \quad$ Numeric vectors

## Details

For two vectors $x$ and $y$, the Morisita index of dissimilarity is

$$
d(x, y)=1-\frac{2 \sum_{i} x_{i} y_{i}}{\left(\lambda_{x}+\lambda_{y}\right) N_{x} N_{y}}
$$

where

$$
\lambda_{x}=\frac{\sum_{i} x_{i}\left(x_{i}-1\right)}{N_{x}\left(N_{x}-1\right)}
$$

and $N_{x}=\sum_{i} x_{i}$ The formula for $\lambda_{x}$ is the unbiased estimate for the probability of drawing two individuals of the same species from $x$, without replacement. The correction for sampling without replacement only makes sense for species count data.
Relation of morisita() to other definitions:

- Equivalent to vegdist() with method = "morisita".

Horn (1966) reformulated the index to use the equations for sampling with replacement in $\lambda_{x}$ and $\lambda_{y}$ :

$$
\lambda_{x}=\frac{\sum_{i} x_{i}^{2}}{N_{x}^{2}}
$$

With this modification, the index is valid for proportions or transformed count data.
Relation of horn_morisita() to other definitions:

- Equivalent to vegdist() with method = "horn".
- Equivalent to the morisitahorn calculator in Mothur.


## Value

The Morisita or Horn-Morisita index between x and y . Both are undefined if x or y have no nonzero elements, in which case we return NaN .

## References

Mosrisita M. Measuring of interspecific association and similarity between communities. Memoirs of the Faculty of Science, Kyushu Univ., Series E (Biology). 1959;3:65-80.
Horn HS. Measurement of "Overlap" in Comparative Ecological Studies. The American Naturalist, 1966;100(914):419-424.
richness Richness or number of observed species

## Description

Richness or number of observed species

## Usage

richness(x)

## Arguments

x
A numeric vector of species counts or proportions.

## Details

The richness is simply the number of nonzero elements in $x$. Relation to other definitions:

- Equivalent to observed_otus() in skbio.diversity.alpha.
- Equivalent to specnumber in vegan.
- Equivalent to the sobs calculator in Mothur.


## Value

The number of species observed, $R \geq 0$.

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
richness(x) # 4
```

ruzicka Ruzicka or weighted Jaccard distance

## Description

Ruzicka or weighted Jaccard distance

## Usage

ruzicka(x, y)

## Arguments

$x, y \quad$ Numeric vectors.

## Details

For vectors $x$ and $y$, the Ruzicka distance is defined as

$$
d(x, y)=1-\frac{\sum_{i} \min (x, y)}{\sum_{i} \max (x, y)}
$$

Relation to other definitions:

- Equivalent to vegdist() with method = "jaccard".
- Related to the Bray-Curtis distance, $d_{r}=2 d_{b c} /\left(1+d_{b c}\right)$.


## Value

The Ruzicka distance between $x$ and $y$. The distance is not defined if all elements in $x$ and $y$ are zero, and we return NaN in this case.

## Description

The Shannon index of diversity

## Usage

shannon(x, base $=\exp (1))$
brillouin_d(x)
heip_e(x)
pielou_e(x)

## Arguments

$x \quad$ A numeric vector of species counts or proportions.

## base

Base of the logarithm to use in the calculation.

## Details

The Shannon index of diversity or Shannon information entropy has deep roots in information theory. It is defined as

$$
H=-\sum_{i} p_{i} \log p_{i}
$$

where $p_{i}$ is the species proportion. Relation to other definitions:

- Equivalent to diversity () in vegan with index = "shannon".
- Equivalent to shannon() in skbio.diversity.alpha.

The Brillouin index (Brillouin 1956) is similar to Shannon's index, but accounts for sampling without replacement. For a vector of species counts $x$, the Brillouin index is

$$
\frac{1}{N} \log \frac{N!}{\prod_{i} x_{i}!}=\frac{\log N!-\sum_{i} \log x_{i}!}{N}
$$

where $N$ is the total number of counts. Relation to other definitions:

- Equivalent to brillouin_d() in skbio.diversity.alpha.
- Equivalent to the shannon calculator in Mothur.

The Brillouin index accounts for the total number of individuals sampled, and should be used on raw count data, not proportions.

Heip's evenness measure is

$$
\frac{e^{H}-1}{S-1}
$$

where $S$ is the total number of species observed. Relation to other definitions:

- Equivalent to heip_e() in skbio.diversity.alpha.

Pielou's Evenness index $J=H / \log S$. Relation to other definitions:

- Equivalent to peilou_e() in skbio.diversity.alpha.


## Value

The Shannon diversity, $H \geq 0$, or related quantity. The value of $H$ is undefined if x sums to zero, and we return NaN in this case. Heip's evenness measure and Pielou's Evenness index are undefined if only one element of x is nonzero, and again we return NaN if this is the case.

## References

Brillouin L. Science and Information Theory. 1956;Academic Press, New York.
Pielou EC. The Measurement of Diversity in Different Types of Biological Collections. Journal of Theoretical Biology. 1966;13:131-144.

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
shannon(x)
# Using a different base is the same as dividing by the log of that base
shannon(x, base = 10)
shannon(x) / log(10)
brillouin_d(x)
# Brillouin index should be almost identical to Shannon index for large N
brillouin_d(10000 * x)
shannon(10000 * x)
heip_e(x)
(exp(shannon(x)) - 1) / (richness(x) - 1)
pielou_e(x)
shannon(x) / log(richness(x))
```


## Description

These measures are based on the sum of squared species proportions. The function dominance() gives this quantity, simpson() gives one minus this quantity, invsimpson() gives the reciprocal of the quantity, and simpson_e gives the reciprocal divided by the number of species.

## Usage

simpson(x)
dominance(x)
invsimpson(x)
simpson_e(x)

## Arguments

$x \quad$ A numeric vector of species counts or proportions.

## Details

For a vector of species counts $x$, the dominance index is defined as

$$
D=\sum_{i} p_{i}^{2}
$$

where $p_{i}$ is the species proportion, $p_{i}=x_{i} / N$, and $N$ is the total number of counts. This is equal to the probability of selecting two individuals from the same species, with replacement. Relation to other definitions:

- Equivalent to dominance() in skbio.diversity.alpha.
- Similar to the simpson calculator in Mothur. They use the unbiased estimate $p_{i}=x_{i}\left(x_{i}-\right.$ 1) $/(N(N-1))$.

Simpson's index is defined here as $1-D$, or the probability of selecting two individuals from different species, with replacement. Relation to other definitions:

- Equivalent to diversity() in vegan with index = "simpson".
- Equivalent to simpson() in skbio.diversity.alpha.

The inverse Simpson index is $1 / D$. Relation to other definitions:

- Equivalent to diversity() in vegan with index = "invsimpson".
- Equivalent to enspie() in skbio.diversity.alpha.
- Similar to the invsimpson calculator in Mothur. They use the unbiased estimate $p_{i}=x_{i}\left(x_{i}-\right.$ 1) $/(N(N-1))$.

Simpson's evenness index is the inverse Simpson index divided by the number of species observed, $1 /(D S)$. Relation to other definitions:

- Equivalent to simpson_e() in skbio.diversity.alpha.

Please be warned that the naming conventions vary between sources. For example Wikipedia calls $D$ the Simpson index and $1-D$ the Gini-Simpson index. We have followed the convention from vegan, to avoid confusion within the R ecosystem.

## Value

The value of the dominance $(0<D \leq 1)$, Simpson index, or inverse Simpson index. The dominance is undefined if the vector sums to zero, in which case we return NaN .

## Examples

```
x <- c(15, 6, 4, 0, 3, 0)
dominance(x)
# Simpson is 1 - D
simpson(x)
1 - dominance(x)
# Inverse Simpson is 1/D
invsimpson(x)
1 / dominance(x)
# Simpson's evenness is 1/(D * S)
simpson_e(x)
1 / (dominance(x) * richness(x))
```

strong Strong's dominance index

## Description

Strong's dominance index measures the maximum departure between the observed proportions and a perfectly even community.

## Usage

strong( $x$ )

## Arguments

## Details

Strong's dominance index is defined as

$$
D_{W}=\max _{i}\left[\frac{b_{i}}{N}-\frac{i}{S}\right]
$$

where $b_{i}$ is the abundance of the $i$ th species, ordered from smallest to largest, $N$ is the total number of counts, and $S$ is the number of species observed.
Equivalent to strong() in skbio.diversity.alpha.

## Value

Strong's dominance index, $0 \leq D_{W}<1$. The index is undefined if $x$ sums to 0 , and we return NaN in this case.

## References

Strong WL. Assessing species abundance uneveness within and between plant communities. Community Ecology. 2002;3:237-246.

## Examples

```
x<- c(9, 0, 1, 2, 5, 2, 1, 1, 0, 7, 2, 1, 0, 1, 1)
strong(x)
```

```
unifrac
```

UniFrac distance

## Description

The UniFrac distance is a phylogenetically-weighted distance between two communities of organisms. The measure has been extended a number of times to include abundance-weighted and variance-adjusted versions.

## Usage

unweighted_unifrac ( $x, y$, tree, $x y \_l a b e l s=N U L L$ )
weighted_unifrac(x, y, tree, $\left.x y \_l a b e l s=N U L L\right)$
weighted_normalized_unifrac(x, y, tree, xy_labels = NULL)
variance_adjusted_unifrac(x, y, tree, xy_labels = NULL)
generalized_unifrac(x, y, tree, alpha = 0.5, xy_labels = NULL)
information_unifrac(x, y, tree, xy_labels = NULL)
phylosor (x, y, tree, $\left.x y \_l a b e l s=N U L L\right)$

## Arguments

$x, y \quad$ Numeric vectors of species counts or proportions.
tree A phylogenetic tree object.
$x y \_l a b e l s \quad$ A character vector of species labels for $x$ and $y$.
alpha Generalized UniFrac parameter.

## Details

These functions compute different variations of the UniFrac distance between communities described by the vectors $x$ and $y$. If the vectors are named, the names will be automatically used to match the vectors with the tree. Missing names are filled in with zero counts. If the vectors are not named and $x y$ _labels is provided, these labels will be used to match the vectors with the tree. If the vectors are not named and $x y$ _labels is not provided, it is assumed that the vectors are already in the correct order, and we simply check that their length matches the number of tips in the tree.
unweighted_unifrac gives the original UniFrac distance from Lozupone and Knight (2005), which is the fraction of total branch length leading to community $x$ or community $y$, but not both. It is based on species presence/absence.
weighted_unifrac gives the abundance-weighted version of UniFrac proposed by Lozupone et al. (2007). In this measure, the branch lengths of the tree are multiplied by the absolute difference in species abundances below each branch.
weighted_normalized_unifrac provides a normalized version of weighted_unifrac, so the distance is between 0 and 1 .
variance_adjusted_unifrac was proposed by Chang et al. (2011) to adjust for the variation of weights in weighted UniFrac under random sampling.
generalized_unifrac was proposed by Chen et al. (2012) to provide a unified mathematical framework for weighted and unweighted UniFrac distance. It includes a parameter, $\alpha$, which can be used to adjust the abundance-weighting in the distance. A value of $\alpha=1$ corresponds to weighted UniFrac. A value of $\alpha=0$ corresponds to unweighted UniFrac if presence/absence vectors are provided. The authors suggest a value of $\alpha=0.5$ as a compromise between weighted and unweighted distances.
information_unifrac was proposed by Wong et al. (2016) to connect UniFrac distance with compositional data analysis. They also proposed a "ratio UniFrac" distance, which is not yet implemented.
phylosor, proposed by Bryant et al. (2008), is closely related to unweighted UniFrac distance. If unweighted UniFrac distance is the analogue of Jaccard distance using branches on a phylogenetic tree, PhyloSor is the analogue of Sorenson dissimilarity.

## Value

The UniFrac distance between communities $x$ and $y$. The distance is not defined if either $x$ or $y$ have all zero elements. We return NaN if this is the case.

## References

Lozupone C, Knight R. UniFrac: a new phylogenetic method for comparing microbial communities. Applied and environmental microbiology. 2005;71(12):8228-8235. 10.1128/AEM.71.12.82288235.2005

Lozupone CA, Hamady M, Kelley ST, Knight R. Quantitative and qualitative $\beta$ diversity measures lead to different insights into factors that structure microbial communities. Applied and environmental microbiology. 2007;73(5):1576-1585. 10.1128/AEM.01996-06
Chang Q., et al. Variance adjusted weighted UniFrac: a powerful beta diversity measure for comparing communities based on phylogeny. BMC Bioinformatics. 2011;12:118. 10.1186/1471-2105-12-118
Chen J, Bittinger K, Charlson ES, Hoffmann C, Lewis J, Wu GD, et al. Associating microbiome composition with environmental covariates using generalized UniFrac distances. Bioinformatics. 2012;28(16):2106-2113. 10.1093/bioinformatics/bts342

Wong RG, Wu JR, Gloor GB. Expanding the UniFrac Toolbox. PLOS ONE. 2016;11(9):1-20. 10.1371/journal.pone. 0161196

Bryant JA, Lamanna C, Morlon H, Kerkhoff AJ, Enquist BJ, Green JL. Microbes on mountainsides: contrasting elevational patterns of bacterial and plant diversity. Proc Natl Acad Sci U S A. 2008;105 Suppl 1:11505-11. 10.1073/pnas. 0801920105

## Examples

```
# From Lozupone and Knight (2005), Figure 1.
# Panel A
x1 <- c(1, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1)
x2 <- c(0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0)
unweighted_unifrac(x1, x2, lozupone_tree)
# Panel B
x3 <- c(0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1)
x4 <- c(1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0)
unweighted_unifrac(x3, x4, lozupone_tree)
# Can use named vectors to specify species
weighted_normalized_unifrac(
    C(A=1, C=1, D=1, F=1, I=1, L=1, N=1),
    C(B=1, E=1, G=1, H=1, J=1, K=1, M=1),
    lozupone_tree)
weighted_normalized_unifrac(x1, x2, lozupone_tree)
# Generalized UniFrac is equal to weighted normalized UniFrac when alpha = 1
generalized_unifrac(x1, x2, lozupone_tree, alpha=1)
generalized_unifrac(x1, x2, lozupone_tree, alpha=0.5)
```

unifrac_components

## Description

Nestedness and turnover components of unweighted UniFrac distance

## Usage

```
unweighted_unifrac_turnover(x, y, tree, xy_labels = NULL)
unweighted_unifrac_nestedness(x, y, tree, xy_labels = NULL)
phylosor_turnover(x, y, tree, xy_labels = NULL)
phylosor_nestedness(x, y, tree, xy_labels = NULL)
```


## Arguments

$x, y \quad$ Numeric vectors of species counts or proportions.
tree A phylogenetic tree object.
$x y \_l a b e l s \quad$ A character vector of species labels for $x$ and $y$.

## Details

Leprieur et al. (2012) showed that measures of phylogenetic beta diversity could be partitioned into nestedness and turnover components, following the approach of Baselga (2010) for Sorenson dissimilarity.

## Value

The nestedness or turnover component of the UniFrac distance between communities $x$ and $y$. This quantity is undefined when either $x$ or $y$ have all elements equal to zero, in which case we return NaN .

## References

Baselga A. Partitioning the turnover and nestedness components of beta diversity. Global Ecol. Biogeogr. 2010;19:134-143.
Leprieur F, Albouy C, De Bortoli J, Cowman PF, Bellwood DR, Mouillot D. Quantifying phylogenetic beta diversity: distinguishing between "true" turnover of lineages and phylogenetic diversity gradients. PLoS One. 2012;7(8):e42760. 10.1371/journal.pone. 0042760

## Examples

```
# Vectors x and y have turnover but no nestedness
x <- c(1, 1, 1, 0, 0, 0, 0, 0)
y <- c(0, 1, 1, 1, 0, 0, 0, 0)
unweighted_unifrac(x, y, leprieur_tree)
unweighted_unifrac_turnover(x, y, leprieur_tree)
unweighted_unifrac_nestedness(x, y, leprieur_tree)
```

```
phylosor(x, y, leprieur_tree)
phylosor_turnover(x, y, leprieur_tree)
phylosor_nestedness(x, y, leprieur_tree)
# Vectors y and z have nestedness but no turnover
z <- c(0, 1, 1, 1, 1, 1, 1, 1)
unweighted_unifrac(y, z, leprieur_tree)
unweighted_unifrac_turnover(y, z, leprieur_tree)
unweighted_unifrac_nestedness(y, z, leprieur_tree)
phylosor(y, z, leprieur_tree)
phylosor_turnover(y, z, leprieur_tree)
phylosor_nestedness(y, z, leprieur_tree)
```

weighted_kulczynski_second
Weighted Kulczynski distance

## Description

The quantitative version of the second Kulczynski index

## Usage

weighted_kulczynski_second(x, y)

## Arguments

$x, y$
Numeric vectors

## Details

The quantitative version of the second Kulczynski index is defined as

$$
d(x, y)=1-\frac{1}{2}\left(\frac{\sum_{i} \min \left(x_{i}, y_{i}\right)}{\sum_{i} x_{i}}+\frac{\sum_{i} \min \left(x_{i}, y_{i}\right)}{\sum_{i} y_{i}}\right) .
$$

Relation of weighted_kulczynski_second() to other definitions:

- Equivalent to vegdist() with method = "kulczynski".
- Equivalent to structkulczynski in Mothur.
- Equivalent to $1-S_{18}$ in Legendre \& Legendre.


## Value

The weighted Kulczynski distance between $x$ and $y$. The distance is undefined if the sum of $x$ or the sum of y is zero, in which case we return NaN .

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```


[^0]:    x
    A numeric vector of species counts or proportions.

