

Package ‘breakaway’

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Title Species Richness Estimation and Modeling

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Description Understanding the drivers of microbial diversity is an important frontier of microbial ecology, and investigating the diversity of samples from microbial ecosystems is a common step in any microbiome analysis. 'breakaway' is the premier package for statistical analysis of microbial diversity. 'breakaway' implements the latest and greatest estimates of species richness, described in Willis and Bunge (2015) <[doi:10.1111/biom.12332](https://doi.org/10.1111/biom.12332)>, Willis et al. (2017) <[doi:10.1111/rssc.12206](https://doi.org/10.1111/rssc.12206)>, and Willis et al. (2018) <[doi:10.1111/biom.12700](https://doi.org/10.1111/biom.12700)>. 'breakaway' also implements commonly used estimates, including the objective Bayes approach described in Barger and Bunge (2010) <[doi:10.1214/10-BA527](https://doi.org/10.1214/10-BA527)>.

License GPL-2

BugReports <https://github.com/adw96/breakaway/issues>

LazyData true

RoxygenNote 7.2.3

Depends R (>= 3.5.0)

Imports ggplot2,
graphics,
lme4,
magrittr,
MASS,
phyloseq,
stats,
tibble,
utils

Suggests devtools,
knitr,
plyr,
RCurl,
rmarkdown,
testthat,
dplyr

VignetteBuilder knitr

URL <https://adw96.github.io/breakaway/>

Encoding UTF-8

R topics documented:

alpha_estimate	3
alpha_estimates	4
apples	4
betta	5
betta_lincom	8
betta_pic	9
betta_random	10
breakaway	12
breakaway_nofl	13
build_frequency_count_tables	15
chao1	15
chao1_bc	16
chao_bunge	17
chao_shen	17
check_format	18
convert	18
cutoff_wrap	19
F_test	19
get_F_stat	20
good_turing	21
hawaii	21
kemp	22
make_design_matrix	23
make_frequency_count_table	24
objective_bayes_geometric	24
objective_bayes_mixedgeo	26
objective_bayes_negbin	27
objective_bayes_poisson	29
pasolli_et_al	30
physeq_wrap	31
plot.alpha_estimates	31
poisson_model	32
poisson_model_nofl	33
proportions_instead	33
rnbinomtable	34
rztnbinomtable	34
sample_inverse_simpson	35
sample_richness	35
sample_shannon	36
sample_shannon_e	36
sample_simpson	37
sample_size_estimate	37
sample_size_figure	38
simulate_betta	39
simulate_betta_random	39
soil_phylo	40
test_submodel	40
toy_metadata	42
toy_otu_table	43
toy_taxonomy	43

<i>alpha_estimate</i>	3
-----------------------	---

true_gini	43
true_hill	44
true_inverse_simpson	44
true_shannon	45
true_shannon_e	45
true_simpson	46
wlrm_transformed	46
wlrm_untransformed	47

Index	49
--------------	-----------

<i>alpha_estimate</i>	<i>alpha_estimate</i>
-----------------------	-----------------------

Description

Build objects of class *alpha_estimate* from their components. *alpha_estimate()* is a constructor method

Usage

```
alpha_estimate(
  estimate = NULL,
  error = NULL,
  estimand = NULL,
  name = NULL,
  interval = NULL,
  interval_type = NULL,
  type = NULL,
  model = NULL,
  warnings = NULL,
  frequentist = NULL,
  parametric = NULL,
  plot = NULL,
  reasonable = NULL,
  other = NULL,
  ...
)
```

Arguments

<i>estimate</i>	The estimate
<i>error</i>	The standard error in the estimate
<i>estimand</i>	What is the estimate trying to estimate? (richness, Shannon...)
<i>name</i>	The name of the method
<i>interval</i>	An interval estimate
<i>interval_type</i>	Type of interval estimate
<i>type</i>	TODO(Amy): Deprecate?
<i>model</i>	What model is fit
<i>warnings</i>	Any warnings?

frequentist	Logical. Frequentist or Bayesian?
parametric	Logical. Parametric or not?
plot	A ggplot associated with the estimate
reasonable	Is the estimate likely to be reasonable?
other	Any other relevant objects
...	Any other objects

Value

An object of class `alpha_estimate`

<code>alpha_estimates</code>	<i>alpha_estimates</i>
------------------------------	------------------------

Description

Build objects of class `alpha_estimates` from their components. `alpha_estimates()` is a constructor method

Usage

```
alpha_estimates(...)
```

Arguments

... Objects of class `alpha_estimate`, or a list of objects of class `alpha_estimate`

Value

An object of class `alpha_estimates`

<code>apples</code>	<i>(Data) Frequency count table of soil microbes in an apples orchard.</i>
---------------------	--

Description

(Data) Frequency count table of soil microbes in an apples orchard.

Usage

```
apples
```

Format

A data frame with 88 rows and 2 variables:

index an index variable

frequency number of taxa that were observed with this frequency ...

References

- Willis, A. and Bunge, J. (2015). Estimating diversity via frequency ratios. *Biometrics*, **71**(4), 1042–1049. doi:[10.1111/biom.12332](https://doi.org/10.1111/biom.12332)
- Walsh, F. et al. (2014). (2014). Restricted streptomycin use in apple orchards did not adversely alter the soil bacteria communities. *Frontiers in Microbiology* **4**, 383.

beta	<i>Modelling total diversity with beta</i>
------	--

Description

This function tests for heterogeneity of total diversity (observed plus unobserved) across multiple sites. It can account or test for fixed effects that may explain diversity. It returns the significance of the covariates in explaining diversity and a hypothesis test for heterogeneity.

Usage

```
beta(
  chats = NULL,
  ses,
  X = NULL,
  initial_est = NULL,
  formula = NULL,
  data = NULL,
  p.digits = 3
)
```

Arguments

chats	A vector of estimates of total diversity at different sampling locations. ‘breakaway’ estimates are suggested in the high-diversity case but not enforced.
ses	The standard errors in chats, the diversity estimates. This can either be a vector of standard errors (with the arguments chats and X), or the name of the variable in the dataframe data that contains the standard errors (with the arguments formula and data).
X	A numeric matrix of covariates. If not supplied, an intercept-only model will be fit. This is optional with the chats argument.
initial_est	(Optional) A vector of length $1 + \text{ncol}(X)$ giving the starting values for the likelihood maximisation search. The first element is the starting estimate for σ^2_u , and the remaining elements are the starting elements for beta. Defaults to NULL, in which case the starting values outlined in the paper are used.
formula	A formula object of the form $y \sim x group$. Required with the data argument.
data	A dataframe containing the response, response standard errors, covariates, and grouping variable. Required with the formula argument.
p.digits	(Optional) A number that specifies the number of digits to which p-values will be rounded. The default value is 3 digits.

Value

table	A coefficient table for the model parameters. The columns give the parameter estimates, standard errors, and p-values, respectively. This model is only as effective as your diversity estimation procedure; for this reason please confirm that your estimates are appropriate and that your model is not misspecified. <code>beta_pic</code> may be useful for this purpose.
cov	Estimated covariance matrix of the parameter estimates.
ssq_u	The estimate of the heterogeneity variance.
homogeneity	The test statistic and p-value for the test of homogeneity.
global	The test statistic and p-value for the test of model explanatory power.
blups	The conditional expected values of the diversity estimates (conditional on the random effects). The authors propose that if the practitioner believes that information from one diversity estimator may inform the others, then using the ‘condfits’ as estimators of total diversity rather than ‘Chats’ may reduce variance of diversity estimates by “sharing strength” across the samples.
blupses	The estimated standard deviation (standard errors) in the blups.
loglikelihood	The log likelihood of the fitted model.
aic	The Akaike information criterion for the fitted model.
aicc	The finite sample correction of the Akaike information criterion for the fitted model.
r_squared_wls	The weighted R^2 statistic, appropriate for heteroskedastic linear models.
function.args	A list containing values initially passed to <code>beta_random</code> .

Note

Ecologists who are interested in the way species richness varies with covariate information often run a regression-type analysis on the observed diversity using their covariate information as predictors. However, in many settings (especially microbial), rare and unobserved taxa play a hugely important role in explaining the subtleties of the ecosystem, however, a regression analysis on the observed diversity level fails to account for these unobserved taxa. By predicting the total level of diversity (for example, via [breakaway](#)) and estimating the standard error in the estimate, one can take account of these unobserved, but important, taxa. In order to account for the estimated nature of the response, a mixed model approach is taken, whereby the varying levels of confidence in the estimates contributes to a diagonal but heteroscedastic covariance matrix. Given covariates constitute the fixed effects in the mixed model, and significance of the random effect term “`sigsq_u`” reflects heterogeneity in the sample, that is, variability that cannot be explained by only the covariates. The authors believe this to be the first attempt at modelling total diversity in a way that accounts for its estimated nature.

Author(s)

Amy Willis

References

- Willis, A., Bunge, J., and Whitman, T. (2015). Inference for changes in biodiversity. *arXiv preprint*.
- Willis, A. and Bunge, J. (2015). Estimating diversity via frequency ratios. *Biometrics*.

See Also

[breakaway](#); [breakaway_nof1](#); [apples](#)

Examples

```
df <- data.frame(chats = c(2000, 3000, 4000, 3000), ses = c(100, 200, 150, 180),
                 Cont_var = c(100, 150, 100, 50))

# formula notation
beta(formula = chats ~ Cont_var, ses = ses, data = df)

# direct input
beta(c(2000, 3000, 4000, 3000), c(100, 200, 150, 180), cbind(1, c(100, 150, 100,
50)))

## handles missing data
beta(c(2000, 3000, 4000, 3000), c(100, 200, 150, NA))

## A test for heterogeneity of apples diversity estimates vs butterfly estimates
beta(c(1552, 1500, 884), c(305, 675, 205), cbind(1, c(0, 0, 1)))
```

betta_lincom	<i>Confidence intervals and testing for linear combinations of fixed effects estimated via betta() or betta_random()</i>
--------------	--

Description

This function provides point estimates, standard errors, and equal-tailed confidence intervals for linear combinations of fixed effects estimated via `betta()` or `betta_random()`. A p-value for a Wald test of the null that the linear combination of effects is equal to zero (against a general alternative) is also returned.

Usage

```
betta_lincom(fitted_betta, linear_com, signif_cutoff = 0.05)
```

Arguments

<code>fitted_betta</code>	A fitted betta object – i.e., the output of either <code>betta()</code> or <code>betta_random()</code> – containing fixed effect estimates of interest.
<code>linear_com</code>	The linear combination of fixed effects for which a point estimate, confidence interval, and hypothesis test are to be produced.
<code>signif_cutoff</code>	The type-I significance threshold for confidence intervals. Defaults to 0.05.

Value

<code>table</code>	A table containing a point estimate, standard error, lower and upper confidence bounds, and a p-value for the linear combination of fixed effects specified in input. The p-value is generated via a two-sided Wald test of the null that the linear combination of fixed effects is equal to zero.
--------------------	---

Author(s)

David Clausen

References

Willis, A., Bunge, J., and Whitman, T. (2015). Inference for changes in biodiversity. *arXiv preprint*.

See Also

[betta](#);

Examples

```
# generate example data
df <- data.frame(chats = c(2000, 3000, 4000, 3000), ses = c(100, 200, 150, 180),
                 Cont_var = c(100, 150, 100, 50))

# fit betta()
example_fit <- betta(formula = chats ~ Cont_var, ses = ses, data = df)

# generate point estimate and 95% CI for mean richness at Cont_var = 125

betta_lincom(fitted_betta = example_fit,
             linear_com = c(1, 125)) # this tells betta_lincom to estimate value of beta_0 + 125*beta_1,
# where beta_0 is the intercept, and beta_1 is the (true value of the) coefficient on Cont_var
```

betta_pic	<i>function for plotting total diversity</i>
-----------	--

Description

A simple plotting interface for comparing total diversity across samples or a covariate gradient.

Usage

```
betta_pic(
  y,
  se,
  x = 1:length(y),
  ylimu = NULL,
  myy = NULL,
  mymain = NULL,
  mycol = NULL,
  labs = NULL,
  mypch = NULL,
  myxlim = NULL
)
```

Arguments

y	A vector of estimates of total diversity. Other parameter estimates are accessible; this method may be used for plotting any parameter estimates..
se	The standard errors in 'y', the diversity (or other parameter's) estimates.
x	A vector of covariates to form the x-coordinates of the intervals. If no argument is given, defaults to the order.
ylimu	The upper endpoint of the y-axis.
myy	Deprecated, for backwards compatibility
mymain	Deprecated, for backwards compatibility
mycol	Deprecated, for backwards compatibility
labs	Deprecated, for backwards compatibility
mypch	Deprecated, for backwards compatibility
myxlim	Deprecated, for backwards compatibility

Value

A ggplot object.

Author(s)

Amy Willis

References

Willis, A., Bunge, J., and Whitman, T. (2015). Inference for changes in biodiversity. arXiv preprint.

See Also

[beta](#)

Examples

```
beta_pic(c(1552, 1500, 884), c(305, 675, 205), mymain = "Example title")
```

beta_random	<i>modelling total diversity with random effects</i>
-------------	--

Description

This function extends beta() to permit random effects modelling.

Usage

```
beta_random(
  chats = NULL,
  ses,
  X = NULL,
  groups = NULL,
  formula = NULL,
  data = NULL,
  p.digits = 3
)
```

Arguments

chats	A vector of estimates of total diversity at different sampling locations. Required with the groups argument, and optionally with the X argument.
ses	The standard errors in chats, the diversity estimates. This can either be a vector of standard errors (with the arguments chats and X), or the name of the variable in the dataframe data that contains the standard errors (with the arguments formula and data).
X	A numeric matrix of covariates corresponding to fixed effects. If not supplied, an intercept-only model will be fit. Optional with the chats and groups arguments.

groups	A categorical variable representing the random-effects groups that each of the estimates belong to. Required with the chats argument and optionally with the X argument.
formula	A formula object of the form $y \sim x group$. Required with the data argument.
data	A dataframe containing the response, response standard errors, covariates, and grouping variable. Required with the formula argument.
p.digits	(Optional) A number that specifies the number of digits to which p-values will be rounded. The default value is 3 digits.

Value

table	A coefficient table for the model parameters. The columns give the parameter estimates, standard errors, and p-values, respectively. This model is only as effective as your diversity estimation procedure; for this reason please confirm that your estimates are appropriate and that your model is not misspecified. beta_pic may be useful for this purpose.
cov	Estimated covariance matrix of the parameter estimates.
ssq_u	The estimate of the heterogeneity variance.
ssq_g	Estimates of within-group variance. The estimate will be zero for groups with only one observation.
homogeneity	The test statistic and p-value for the test of homogeneity.
global	The test statistic and p-value for the test of model explanatory power.
blups	The conditional expected values of the diversity estimates (conditional on the random effects). Estimates of variability for the random effects case are unavailable at this time; please contact the maintainer if needed.
loglikelihood	The log likelihood of the fitted model.
aic	The Akaike information criterion for the fitted model.
aicc	The finite sample correction of the Akaike information criterion for the fitted model.
function.args	A list containing values initially passed to beta_random.

Author(s)

Amy Willis

References

Willis, A., Bunge, J., and Whitman, T. (2015). Inference for changes in biodiversity. *arXiv preprint*.

See Also

[beta](#);

Examples

```
df <- data.frame(chats = c(2000, 3000, 4000, 3000),
  ses = c(100, 200, 150, 180),
  Cont_var = c(100, 150, 100, 50),
  groups = c("a", "a", "a", "b"))
```

```
# formula notation
betta_random(formula = chats ~ Cont_var | groups,
ses = ses,
data = df)

# direct input
betta_random(c(2000, 3000, 4000, 3000), c(100, 200, 150, 180),
X = cbind(Int = 1, Cont_var = c(100, 150, 100, 50)),
groups = c("a", "a", "a", "b"))

## handles missing data
betta_random(c(2000, 3000, 4000, 3000), c(100, 200, 150, NA),
groups = c("a", NA,
"b", "b"))
```

breakaway

*Species richness estimation with breakaway***Description**

breakaway is a wrapper for modern species richness estimation for modern datasets

Usage

```
breakaway(
  input_data,
  cutoff = NA,
  output = NULL,
  plot = NULL,
  answers = NULL,
  print = NULL,
  ...
)
```

Arguments

input_data	An input type that can be processed by <code>convert()</code>
cutoff	The maximum frequency count to use for model fitting
output	Deprecated; only for backwards compatibility
plot	Deprecated; only for backwards compatibility
answers	Deprecated; only for backwards compatibility
print	Deprecated; only for backwards compatibility
...	Additional arguments will be ignored; this is for backward compatibility

Value

An object of class `alpha_estimate`

Note

‘breakaway’ presents an estimator of species richness that is well-suited to the high-diversity/microbial setting. However, many microbial datasets display more diversity than the Kemp-type models can permit. In this case, the log-transformed WLRM diversity estimator of Rocchetti et. al. (2011) is returned. The authors’ experience suggests that some datasets that require the log-transformed WLRM contain “false” diversity, that is, diversity attributable to sequencing errors (via an inflated singleton count). The authors encourage judicious use of diversity estimators when the dataset may contain these errors, and recommend the use of [breakaway_nof1](#) as an exploratory tool in this case.

Author(s)

Amy Willis

References

Willis, A. and Bunge, J. (2015). Estimating diversity via frequency ratios. *Biometrics*, **71**(4), 1042–1049.

See Also

[breakaway_nof1](#); [kemp](#); [apples](#)

Examples

```
breakaway(apples)
breakaway(apples, plot = FALSE, output = FALSE, answers = TRUE)
```

breakaway_nof1

species richness estimation without singletons

Description

This function permits estimation of total diversity based on a sample frequency count table. Unlike [breakaway](#), it does not require an input for the number of species observed once, making it an excellent exploratory tool for microbial ecologists who believe that their sample may contain spurious singletons. The underlying estimation procedure is similar to that of [breakaway](#) and is outlined in Willis & Bunge (2014). The diversity estimate, standard error, estimated model coefficients and plot of the fitted model are returned.

Usage

```
breakaway_nof1(
  input_data,
  output = NULL,
  plot = NULL,
  answers = NULL,
  print = NULL
)
```

Arguments

input_data	An input type that can be processed by <code>convert()</code>
output	Deprecated; only for backwards compatibility
plot	Deprecated; only for backwards compatibility
answers	Deprecated; only for backwards compatibility
print	Deprecated; only for backwards compatibility

Value

An object of class `alpha_estimate`

code	A category representing algorithm behaviour. ‘code=1’ indicates no nonlinear models converged and the transformed WLRM diversity estimate of Rocchetti et. al. (2011) is returned. ‘code=2’ indicates that the iteratively reweighted model converged and was returned. ‘code=3’ indicates that iterative reweighting did not converge but a model based on a simplified variance structure was returned (in this case, the variance of the frequency ratios is assumed to be proportional to the denominator frequency index). Please peruse your fitted model before using your diversity estimate.
name	The “name” of the selected model. The first integer represents the numerator polynomial degree and the second integer represents the denominator polynomial degree. See Willis & Bunge (2014) for details.
para	Estimated model parameters and standard errors.
est	The estimate of total (observed plus unobserved) diversity.
seest	The standard error in the diversity estimate.
full	The chosen nonlinear model for frequency ratios.

Note

It is common for microbial ecologists to believe that their dataset contains false diversity. This often arises because sequencing errors result in classifying already observed organisms as new organisms. ‘breakaway_nof1’ was developed as an exploratory tool in this case. Practitioners can run ‘breakaway’ on their dataset including the singletons, and ‘breakaway_nof1’ on their dataset excluding the singletons, and assess if the estimated levels of diversity are very different. Great disparity may provide evidence of an inflated singleton count, or at the very least, that ‘breakaway’ is especially sensitive to the number of rare species observed. Note that ‘breakaway_nof1’ may be less stable than ‘breakaway’ due to predicting based on a reduced dataset, and have greater standard errors.

Author(s)

Amy Willis

References

- Willis, A. (2015). Species richness estimation with high diversity but spurious singletons. *arXiv*.
 Willis, A. and Bunge, J. (2015). Estimating diversity via frequency ratios. *Biometrics*.

See Also

[breakaway](#); [apples](#)

Examples

```
breakaway_nof1(apples[-1, ])
breakaway_nof1(apples[-1, ], plot = FALSE, output = FALSE, answers = TRUE)
```

```
build_frequency_count_tables
```

Build frequency count tables from an OTU table

Description

Build frequency count tables from an OTU table

Usage

```
build_frequency_count_tables(the_table)
```

Arguments

the_table	An OTU table as a data frame or a matrix. Columns are the samples and rows give the taxa.
-----------	---

Value

A list of frequency count tables corresponding to the columns.

```
chao1
```

Chao1 species richness estimator

Description

This function implements the Chao1 richness estimate, which is often mistakenly referred to as an index.

Usage

```
chao1(input_data, output = NULL, answers = NULL)
```

Arguments

input_data	An input type that can be processed by <code>convert()</code> or a phyloseq object
output	Deprecated; only for backwards compatibility
answers	Deprecated; only for backwards compatibility

Value

An object of class `alpha_estimate`, or `alpha_estimates` for phyloseq objects

Note

The authors of this package strongly discourage the use of this estimator. It is only valid when you wish to assume that every taxa has equal probability of being observed. You don't really think that's possible, do you?

Author(s)

Amy Willis

Examples

```
chao1(apples)
```

chao1_bc

Bias-corrected Chao1 species richness estimator

Description

This function implements the bias-corrected Chao1 richness estimate.

Usage

```
chao1_bc(input_data, output = NULL, answers = NULL)
```

Arguments

input_data	An input type that can be processed by <code>convert()</code> or a phyloseq object
output	Deprecated; only for backwards compatibility
answers	Deprecated; only for backwards compatibility

Value

An object of class `alpha_estimate`, or `alpha_estimates` for phyloseq objects

Note

The authors of this package strongly discourage the use of this estimator. It is only valid when you wish to assume that every taxa has equal probability of being observed. You don't really think that's possible, do you?

Author(s)

Amy Willis

Examples

```
chao1_bc(apples)
```

chao_bunge	<i>Chao-Bunge species richness estimator</i>
------------	--

Description

This function implements the species richness estimation procedure outlined in Chao & Bunge (2002).

Usage

```
chao_bunge(input_data, cutoff = 10, output = NULL, answers = NULL)
```

Arguments

input_data	An input type that can be processed by <code>convert()</code> or a <code>phyloseq</code> object
cutoff	The maximum frequency to use in fitting.
output	Deprecated; only for backwards compatibility
answers	Deprecated; only for backwards compatibility

Value

An object of class `alpha_estimate`, or `alpha_estimates` for `phyloseq` objects

Author(s)

Amy Willis

Examples

```
chao_bunge(apples)
```

chao_shen	<i>The Chao-Shen estimate of Shannon diversity</i>
-----------	--

Description

The Chao-Shen estimate of Shannon diversity

Usage

```
chao_shen(input_data)
```

Arguments

input_data	An input type that can be processed by <code>convert()</code>
------------	---

Value

An object of class `alpha_estimate`

check_format	<i>Run some basic checks on a possible frequency count table</i>
--------------	--

Description

Run some basic checks on a possible frequency count table

Usage

```
check_format(output_data)
```

Arguments

output_data A matrix to test

Value

A checked frequency count table

convert	<i>convert between different inputs for alpha-diversity estimates</i>
---------	---

Description

Inputs slated for development include phyloseq and otu_table

Usage

```
convert(input_data)
```

Arguments

input_data Supported types include filenames, data frames, matrices, vectors...

Value

Frequency count able

cutoff_wrap	<i>Find a cut-off for estimates relying on contiguous counts</i>
-------------	--

Description

Find a cut-off for estimates relying on contiguous counts

Usage

```
cutoff_wrap(my_data, requested = NA)
```

Arguments

my_data	Frequency count table
requested	The user-requested cutoff

Value

Cutoff value

F_test	<i>Conduct F test of null hypothesis $LB = 0$ using output from <code>betta()</code> or <code>betta_random()</code></i>
--------	--

Description

This function performs an F-test of a null hypothesis $LB = 0$ where B is a vector of p fixed effects returned by `betta()` or `betta_random()` and L is an m x p matrix with linearly independent rows.

Usage

```
F_test(fitted_betta, L, method = "bootstrap", nboot = 1000)
```

Arguments

fitted_betta	A fitted betta object – i.e., the output of either <code>betta()</code> or <code>betta_random()</code> – containing fixed effect estimates of interest.
L	An m x p matrix defining the null $LB = 0$. L must have full row rank.
method	A character variable indicating which method should be used to estimate the distribution of the test statistic under the null.
nboot	Number of bootstrap samples to use if <code>method = "bootstrap"</code> . Ignored if <code>method = "asymptotic"</code> .

Value

A list containing

pval	The p-value
F_stat	The calculated F statistic
boot_F	A vector of bootstrapped F statistics if bootstrap has been used. Otherwise NULL.

Author(s)

David Clausen

ReferencesWillis, A., Bunge, J., and Whitman, T. (2015). Inference for changes in biodiversity. *arXiv preprint*.**See Also**

[betta](#);
[betta_random](#);
[betta](#);
[get_F_stat](#)

Examples

```

# generate example data
df <- data.frame(chats = c(2000, 3000, 4000, 3000,
2000, 3000, 4000, 3000), ses = c(100, 200, 150, 180,
100, 200, 150, 180),
                Cont_var = c(100, 150, 100, 50,
100, 150, 100, 50),
                Cont_var_2 = c(50, 200, 25, 125,
50, 200, 25, 125))

# fit betta()
example_fit <- betta(formula = chats ~ Cont_var + Cont_var_2, ses = ses, data = df)

# construct L for hypothesis that B_cont_var = B_cont_var_2 = 0
L <- rbind(c(0,1,0),
           c(0,0,1))

F_test_results <- F_test(example_fit, L, nboot = 100)

```

get_F_stat

Calculate F statistic under null hypothesis $LB = 0$ using output from `betta()` or `betta_random()`

Description

This function calculates an F statistic for a test of null hypothesis $LB = 0$ (against an unrestricted alternative) where B is a vector of p fixed effects returned by `betta()` or `betta_random()` and L is an $m \times p$ matrix with linearly independent rows.

Usage

```
get_F_stat(fitted_betta, L)
```

Arguments

- fitted_betta** A fitted betta object – i.e., the output of either `betta()` or `betta_random()` – containing fixed effect estimates of interest.
- L** An $m \times p$ matrix defining the null $LB = 0$. **L** must have full row rank.

Value

- A list containing
- F_stat** The calculated F statistic

<code>good_turing</code>	<i>The Good-Turing estimate of species richness</i>
--------------------------	---

Description

The Good-Turing estimate of species richness

Usage

```
good_turing(input_data)
```

Arguments

- input_data** An input type that can be processed by `convert()` or a phyloseq object

Value

An object of class `alpha_estimate`, or `alpha_estimates` for phyloseq objects

<code>hawaii</code>	<i>(Data) Frequency count table of soil microbes in Hawaii.</i>
---------------------	---

Description

(Data) Frequency count table of soil microbes in Hawaii.

Usage

```
hawaii
```

Format

A data frame with 198 rows and 2 variables:

index an index variable

frequency number of taxa that were observed with this frequency ...

References

Willis, A. and Bunge, J. (2015). Estimating diversity via frequency ratios. *Biometrics*, **71**(4), 1042–1049. doi:[10.1111/biom.12332](https://doi.org/10.1111/biom.12332)

kemp

*Species richness estimation with Kemp-type models***Description**

This function implements the species richness estimation procedure outlined in Willis & Bunge (2015). The diversity estimate, standard error, estimated model coefficients, model details and plot of the fitted model are returned.

Usage

```
kemp(
  input_data,
  cutoff = NA,
  output = NULL,
  plot = NULL,
  answers = NULL,
  print = NULL,
  ...
)
```

Arguments

input_data	An input type that can be processed by <code>convert()</code>
cutoff	The maximum frequency count to use for model fitting
output	Deprecated; only for backwards compatibility
plot	Deprecated; only for backwards compatibility
answers	Deprecated; only for backwards compatibility
print	Deprecated; only for backwards compatibility
...	Additional arguments will be ignored; this is for backward compatibility

Value

An object of class `alpha_estimate`

code	A category representing algorithm behaviour. ‘code=1’ indicates no nonlinear models converged and the transformed WLRM diversity estimate of Rocchetti et. al. (2011) is returned. ‘code=2’ indicates that the iteratively reweighted model converged and was returned. ‘code=3’ indicates that iterative reweighting did not converge but a model based on a simplified variance structure was returned (in this case, the variance of the frequency ratios is assumed to be proportional to the denominator frequency index). Please peruse your fitted model before using your diversity estimate.
name	The “name” of the selected model. The first integer represents the numerator polynomial degree and the second integer represents the denominator polynomial degree of the model for the frequency ratios. See Willis & Bunge (2015) for details.
para	Estimated model parameters and standard errors.
est	The estimate of total (observed plus unobserved) diversity.

seest	The standard error in the diversity estimate.
full	The chosen nonlinear model for frequency ratios.
ci	An asymmetric 95% confidence interval for diversity.

Author(s)

Amy Willis

References

- Willis, A. and Bunge, J. (2015). Estimating diversity via frequency ratios. *Biometrics*, **71**(4), 1042–1049.
- Rocchetti, I., Bunge, J. and Bohning, D. (2011). Population size estimation based upon ratios of recapture probabilities. *Annals of Applied Statistics*, **5**.

See Also

[breakaway](#); [breakaway_nof1](#); [apples](#)

Examples

```
kemp(apples)
kemp(apples, plot = FALSE, output = FALSE, answers = TRUE)
```

make_design_matrix	<i>Make design matrix</i>
--------------------	---------------------------

Description

Make design matrix

Usage

```
make_design_matrix(phyloseq_object, variables)
```

Arguments

phyloseq_object	A phyloseq object
variables	variable names

Value

A matrix object giving the design matrix from the desired variables.

```
make_frequency_count_table
```

Draw frequency count subtables from an OTU table

Description

Draw frequency count subtables from an OTU table

Usage

```
make_frequency_count_table(labels)
```

Arguments

labels	A vector of counts of the taxa; i.e. a vector giving the number of times each taxon was observed.
--------	---

Value

A frequency count table.

```
objective_bayes_geometric
```

Estimate species richness with an objective Bayes method using a geometric model

Description

Estimate species richness with an objective Bayes method using a geometric model

Usage

```
objective_bayes_geometric(
  data,
  output = TRUE,
  plot = TRUE,
  answers = FALSE,
  tau = 10,
  burn.in = 100,
  iterations = 2500,
  Metropolis.stdev.N = 75,
  Metropolis.start.theta = 1,
  Metropolis.stdev.theta = 0.3
)
```


Arguments

data	TODO(Kathryn)(Kathryn)
output	TODO(Kathryn)(Kathryn)
plot	TODO(Kathryn)(Kathryn)
answers	TODO(Kathryn)(Kathryn)
tau	TODO(Kathryn)
burn.in	TODO(Kathryn)
iterations	TODO(Kathryn)
Metropolis.stdev.N	TODO(Kathryn)
Metropolis.start.theta	TODO(Kathryn)
Metropolis.stdev.theta	TODO(Kathryn)

Value

A list of results, including

est	the median of estimates of N
,	
ci	a confidence interval for N
,	
mean	the mean of estimates of N
,	
semeanest	the standard error of mean estimates
,	
dic	the DIC of the model
,	
fits	fitted values
, and	
diagnostics	model diagnostics
.	

```
objective_bayes_mixedgeo
```

Objective Bayes species richness estimate with the mixed-geometric model

Description

Objective Bayes species richness estimate with the mixed-geometric model

Usage

```
objective_bayes_mixedgeo(
  data,
  output = TRUE,
  plot = TRUE,
  answers = FALSE,
  tau = 10,
  burn.in = 100,
  iterations = 2500,
  Metropolis.stdev.N = 100,
  Metropolis.start.T1 = 1,
  Metropolis.stdev.T1 = 2,
  Metropolis.start.T2 = 3,
  Metropolis.stdev.T2 = 2,
  bars = 3
)
```

Arguments

data	TODO(Kathryn)
output	TODO(Kathryn)
plot	TODO(Kathryn)
answers	TODO(Kathryn)
tau	TODO(Kathryn)
burn.in	TODO(Kathryn)
iterations	TODO(Kathryn)
Metropolis.stdev.N	TODO(Kathryn)
Metropolis.start.T1	TODO(Kathryn)
Metropolis.stdev.T1	TODO(Kathryn)
Metropolis.start.T2	TODO(Kathryn)
Metropolis.stdev.T2	TODO(Kathryn)
bars	TODO(Kathryn)

Value

A list of results, including

```
est          the median of estimates of N
,
ci           a confidence interval for N
,
mean         the mean of estimates of N
,
semeanest    the standard error of mean estimates
,
dic          the DIC of the model
,
fits         fitted values
, and
diagnostics  model diagnostics
.
```

objective_bayes_negbin

Objective Bayes species richness estimate with the Negative Binomial model

Description

Objective Bayes species richness estimate with the Negative Binomial model

Usage

```
objective_bayes_negbin(
  data,
  output = TRUE,
  plot = TRUE,
  answers = FALSE,
  tau = 10,
  burn.in = 1000,
  iterations = 5000,
  Metropolis.stdev.N = 100,
  Metropolis.start.T1 = -0.8,
  Metropolis.stdev.T1 = 0.01,
  Metropolis.start.T2 = 0.8,
  Metropolis.stdev.T2 = 0.01,
  bars = 5
)
```

Arguments

data	TODO(Kathryn)
output	TODO(Kathryn)
plot	TODO(Kathryn)
answers	TODO(Kathryn)
tau	TODO(Kathryn)
burn.in	TODO(Kathryn)
iterations	TODO(Kathryn)
Metropolis.stdev.N	TODO(Kathryn)
Metropolis.start.T1	TODO(Kathryn)
Metropolis.stdev.T1	TODO(Kathryn)
Metropolis.start.T2	TODO(Kathryn)
Metropolis.stdev.T2	TODO(Kathryn)
bars	TODO(Kathryn)

Value

A list of results, including

est	the median of estimates of N
,	
ci	a confidence interval for N
,	
mean	the mean of estimates of N
,	
semeanest	the standard error of mean estimates
,	
dic	the DIC of the model
,	
fits	fitted values
, and	
diagnostics	model diagnostics
.	

objective_bayes_poisson

Objective Bayes species richness estimate with the Poisson model

Description

Objective Bayes species richness estimate with the Poisson model

Usage

```
objective_bayes_poisson(
  data,
  output = TRUE,
  plot = TRUE,
  answers = FALSE,
  tau = 10,
  burn.in = 100,
  iterations = 2500,
  Metropolis.stdev.N = 75,
  Metropolis.start.lambda = 1,
  Metropolis.stdev.lambda = 0.3,
  bars = 5
)
```

Arguments

data	TODO(Kathryn)
output	TODO(Kathryn)
plot	TODO(Kathryn)
answers	TODO(Kathryn)
tau	TODO(Kathryn)
burn.in	TODO(Kathryn)
iterations	TODO(Kathryn)
Metropolis.stdev.N	TODO(Kathryn)
Metropolis.start.lambda	TODO(Kathryn)
Metropolis.stdev.lambda	TODO(Kathryn)
bars	TODO(Kathryn)

Value

A list of results, including

est	the median of estimates of N
,	
ci	a confidence interval for N

```
,
mean          the mean of estimates of N
,
semeanest     the standard error of mean estimates
,
dic           the DIC of the model
,
fits          fitted values
, and
diagnostics   model diagnostics
.
```

`pasolli_et_al` *(Data) Data frame of covariate information about pasolli_et_al.*

Description

(Data) Data frame of covariate information about pasolli_et_al.

Usage

```
pasolli_et_al
```

Format

A data frame with 4930 rows and 9 variables:

SGB.ID sample ID

AverageAbundance average abundance in the sample

std standard deviation

q1 first quartile

MedianAbundance median abundance

q3 third quartile

min minimum

max maximum

#.Samples number of samples ...

physeq_wrap	<i>Wrapper for phyloseq objects</i>
-------------	-------------------------------------

Description

Wrapper for phyloseq objects

Usage

```
physeq_wrap(fn, physeq, ...)
```

Arguments

fn	alpha diversity estimator function with breakaway to be applied to physeq
physeq	A phyloseq object, or an object of class <code>otu_table</code>
...	Additional arguments for fn

Value

Object of class `alpha_estimates`

plot.alpha_estimates	<i>Plot function for alpha_estimates class</i>
----------------------	--

Description

Plot function for `alpha_estimates` class

Usage

```
## S3 method for class 'alpha_estimates'
plot(
  x,
  physeq = NULL,
  measure = NULL,
  color = NULL,
  shape = NULL,
  title = NULL,
  trim_plot = FALSE,
  ...
)
```

Arguments

x	Object of class alpha_estimates.
physeq	(Optional). Default NULL. Required object of class phyloseq if including a sample_data variable for color or shape.
measure	(Optional). If there are multiple richness measures included in x, this can be set to the the desired measure to be plotted. Defaults to the measure of the first alpha diversity estimate.
color	(Optional). Default NULL. The sample variable to map to different colors. Can be a single character string of the variable name in sample_data or a custom supplied vector with length equal to the number of samples.
shape	(Optional). Default NULL. The sample variable to map to different shapes. Can be a single character string of the variable name in sample_data or a custom supplied vector with length equal to the number of samples.
title	(Optional). Default NULL. Character string. The main title for the graphic.
trim_plot	(Optional). Default FALSE. Boolean indicator for whether you want the plot to include the full confidence intervals.
...	See details

Details

... does not currently have any implemented options. Optional arguments currently include "trim_plot", a Optional

Value

A ggplot object.

Examples

```
library(phyloseq)
data(GlobalPatterns)
alphas <- breakaway(GlobalPatterns)
plot(alphas)
```

poisson_model

PoissonModel

Description

A model to estimate the number of missing taxa under a Poisson Model

Usage

```
poisson_model(input_data, cutoff = 10)
```

Arguments

input_data	A frequency count table
cutoff	The largest frequency to use for predicting f0. Defaults to 10.

Value

An object of class `alpha_estimate`.

<code>poisson_model_nof1</code>	<i>PoissonModelNof1</i>
---------------------------------	-------------------------

Description

A model to estimate the number of missing taxa under a zero- and one-truncated Poisson Model

Usage

```
poisson_model_nof1(input_data, cutoff = 10)
```

Arguments

<code>input_data</code>	A frequency count table
<code>cutoff</code>	The largest frequency to use for predicting f_0 and f_1 . Defaults to 10.

Value

An object of class `alpha_estimate`.

<code>proportions_instead</code>	<i>OTU table to relative abundances</i>
----------------------------------	---

Description

OTU table to relative abundances

Usage

```
proportions_instead(the_table)
```

Arguments

<code>the_table</code>	An OTU table
------------------------	--------------

Value

A proportion table or vector.

rnbinomtable	<i>Negative binomially distributed frequency count tables.</i>
--------------	--

Description

Simulate a frequency count table based on a negative binomial model. Zero-truncated, obviously.

Usage

```
rnbinomtable(C, size, probability)
```

Arguments

C	species richness
size	size parameter for the negative binomial distribution
probability	probability parameter for the negative binomial distribution

Value

A simulated frequency count table.

Author(s)

Amy Willis

rztnbinomtable	<i>beta version: Zero-truncated negative binomially distributed frequency count tables.</i>
----------------	---

Description

Simulate a frequency count table based on a negative binomial model. Zero-truncated, obviously.

Usage

```
rztnbinomtable(C, size, probability)
```

Arguments

C	species richness
size	size parameter for the negative binomial distribution
probability	probability parameter for the negative binomial distribution

Value

A simulated frequency count table.

Author(s)

Amy Willis

`sample_inverse_simpson`*Plug-in Inverse Simpson diversity*

Description

This function implements the plug-in Inverse Simpson diversity

Usage

```
sample_inverse_simpson(input_data)
```

Arguments

`input_data` An input type that can be processed by `convert()` or a `phyloseq` object

Value

An object of class `alpha_estimate`, or `alpha_estimates` for `phyloseq` objects

Examples

```
sample_inverse_simpson(apples)
```

`sample_richness`*Sample richness estimator*

Description

This function implements the sample richness estimate, which is the number of non-zero taxa per sample.

Usage

```
sample_richness(input_data)
```

Arguments

`input_data` An input type that can be processed by `convert()` or a `phyloseq` object

Value

An object of class `alpha_estimate`, or `alpha_estimates` for `phyloseq` objects

Examples

```
sample_richness(apples)
```

sample_shannon	<i>Plug-in Shannon diversity</i>
----------------	----------------------------------

Description

This function implements the plug-in Shannon diversity

Usage

```
sample_shannon(input_data)
```

Arguments

input_data An input type that can be processed by `convert()` or a `phyloseq` object

Value

An object of class `alpha_estimate`, or `alpha_estimates` for `phyloseq` objects

Examples

```
sample_shannon(apples)
```

sample_shannon_e	<i>Plug-in Shannon's E ("Equitability")</i>
------------------	---

Description

This function implements the plug-in Shannon's E

Usage

```
sample_shannon_e(input_data)
```

Arguments

input_data An input type that can be processed by `convert()` or a `phyloseq` object

Value

An object of class `alpha_estimate`, or `alpha_estimates` for `phyloseq` objects

Examples

```
sample_shannon_e(apples)
```

sample_simpson	<i>Plug-in Simpson diversity</i>
----------------	----------------------------------

Description

This function implements the plug-in Simpson diversity

Usage

```
sample_simpson(input_data)
```

Arguments

input_data An input type that can be processed by `convert()` or a phyloseq object

Value

An object of class `alpha_estimate`, or `alpha_estimates` for phyloseq objects

Examples

```
sample_simpson(apples)
```

sample_size_estimate	<i>Estimate the sample size needed to do an unpaired one-way test using betta</i>
----------------------	---

Description

Estimate the sample size needed to do an unpaired one-way test using betta

Usage

```
sample_size_estimate(
  control_group_est,
  se_est,
  diff = 5,
  alpha = 0.05,
  prop = 0.8,
  samples = 20,
  precision = 5
)
```

Arguments

control_group_est	An estimate of the alpha diversity parameter for the control group
se_est	An estimate of the (common) standard deviation
diff	An estimate of the difference between the control and non-control groups
alpha	Minimum significance level desired
prop	Desired power
samples	Number of bootstrap resamples used to estimate the sample size. Increase for a more accurate estimate.
precision	How much to increment the sample size as we try to increase the power

Value

An estimate of the necessary sample size and some details

sample_size_figure	<i>Plot the power obtained with sample size</i>
--------------------	---

Description

Plot the power obtained with sample size

Usage

```
sample_size_figure(control_group_est, se_est, diff = 5, samples = 20)
```

Arguments

control_group_est	An estimate of the alpha diversity parameter for the control group
se_est	An estimate of the (common) standard deviation
diff	An estimate of the difference between the control and non-control groups
samples	Number of bootstrap resamples used to estimate the sample size. Increase for a more accurate estimate.

Value

A plot of the power with the sample size

simulate_betta	<i>Simulate from a fitted betta model</i>
----------------	---

Description

Simulate from a fitted betta model

Usage

```
simulate_betta(fitted_betta, nsim)
```

Arguments

fitted_betta	A fitted betta object
nsim	Number of times to simulate

Value

A list of length nsim, each element of which is a vector of simulated Y-values under the fitted betta model

simulate_betta_random	<i>Simulate from a fitted betta_random model</i>
-----------------------	--

Description

Simulate from a fitted betta_random model

Usage

```
simulate_betta_random(fitted_betta, nsim)
```

Arguments

fitted_betta	A fitted betta_random object
nsim	Number of times to simulate

Value

A list of length nsim, each element of which is a vector of simulated Y-values under the fitted betta model

soil_phylo	(Data) Data frame of soil data from whitman_et_al.
------------	--

Description

A phyloseq object with an OTU table and sample data from a soil microbiome study.

Usage

```
soil_phylo
```

Format

A phyloseq-class experiment-level object with an OTU table and sample data.

otu_table OTU table with 7,770 taxa and 119 samples

tax_table taxonomy table

sam_data sample data with the following covariates:

- Plants, values 0 and 1. Index for different plants
- Day, values 0 (initial sampling point), 1 (12 days after treatment additions), and 2 (82 days after treatment additions). Index for different days of measurement
- Amdmt, values 0 (no additions), 1 (biochar additions), and 2 (fresh biomass additions). Index for different soil additives.
- DayAmdmt, values 00, 01, 02, 10, 11, 12, 20, 21, and 22. A single index for the combination of Day and Amdmt with Day as the first digit and Amdmt as the second digit.
- ID, values A, B, C, D, and F. Index for different soil plots.

...

References

Whitman, T., Pepe-Ranne, C., Enders, A., Koechli, C., Campbell, A., Buckley, D. H., Lehmann, J. (2016). *Dynamics of microbial community composition and soil organic carbon mineralization in soil following addition of pyrogenic and fresh organic matter*. The ISME journal, 10(12):2918. <doi: 10.1038/ismej.2016.68>.

test_submodel	Conduct F test of null hypothesis $LB = 0$ using output from <code>betta()</code> or <code>betta_random()</code>
---------------	--

Description

This function performs an F-test of a null hypothesis $LB = 0$ where B is a vector of p fixed effects returned by `betta()` or `betta_random()` and L is an m x p matrix with linearly independent rows.

Usage

```
test_submodel(
  fitted_beta,
  submodel_formula,
  method = "bootstrap",
  nboot = 1000
)
```

Arguments

fitted_beta	A fitted beta object – i.e., the output of either <code>beta()</code> or <code>beta_random()</code> – containing fixed effect estimates of interest.
submodel_formula	A formula defining which submodel to treat as the null. It is not necessary to include random effects in this formula (they will be ignored if included – the submodel will be fit with the same random effect structure as the full model regardless of input.)
method	A character variable indicating which method should be used to estimate the distribution of the test statistic under the null.
nboot	Number of bootstrap samples to use if <code>method = "bootstrap"</code> . Ignored if <code>method = "asymptotic"</code> .

Value

A list containing	
pval	The p-value
F_stat	The calculated F statistic
boot_F	A vector of bootstrapped F statistics if bootstrap has been used. Otherwise NULL.

Author(s)

David Clausen

References

Willis, A., Bunge, J., and Whitman, T. (2015). Inference for changes in biodiversity. *arXiv preprint*.

See Also

[beta](#);
[beta_random](#);
[beta](#);
[F_test](#)

Examples

```
# generate example data
df <- data.frame(chats = c(2000, 3000, 4000, 3000,
2000, 3000, 4000, 3000), ses = c(100, 200, 150, 180,
100, 200, 150, 180),
                Cont_var = c(100, 150, 100, 50,
100, 150, 100, 50),
                Cont_var_2 = c(50,200,25,125,
50,200,25,125))

# fit betta()
example_fit <- betta(formula = chats ~ Cont_var + Cont_var_2, ses = ses, data = df)

# construct L for hypothesis that B_cont_var = B_cont_var_2 = 0
L <- rbind(c(0,1,0),
           c(0,0,1))

F_test_results <- F_test(example_fit, L, nboot = 100)
```

toy_metadata	<i>(Data) Data frame of covariate information about toy_otu_table.</i>
--------------	--

Description

(Data) Data frame of covariate information about toy_otu_table.

Usage

```
toy_metadata
```

Format

A data frame with 143 rows and 4 variables:

Years Year of sampling

bloom2 Did the sample correspond to a bloom event?

Period What season was sampled?

Site Where was the sample taken from? ...

toy_otu_table	(Data) A toy OTU table.
---------------	-------------------------

Description

Covariate info available in ‘toy_metadata’. A data frame with 448 rows and 143 columns. Rows give the abundance of each taxon; columns give the samples

Usage

```
toy_otu_table
```

Format

An object of class `data.frame` with 448 rows and 143 columns.

toy_taxonomy	(Data) The taxonomy of the OTUs in ‘toy_otu_table’.
--------------	---

Description

(Data) The taxonomy of the OTUs in ‘toy_otu_table’.

Usage

```
toy_taxonomy
```

Format

An object of class `factor` of length 448.

true_gini	Calculate the true Gini-Simpson index
-----------	---------------------------------------

Description

Calculate the true Gini-Simpson index

Usage

```
true_gini(input)
```

Arguments

input	A vector of proportions.
-------	--------------------------

Value

The Gini-Simpson index of the population given by input.

Note

This function is intended for population-level data. If you are dealing with a microbial sample, use DivNet instead.

true_hill	<i>Calculate the true Hill numbers</i>
-----------	--

Description

Calculate the true Hill numbers

Usage

```
true_hill(input, q)
```

Arguments

input	A vector of proportions.
q	The Hill number of interest. $q = 0$ corresponds to species richness, $q = 2$ corresponds to inverse Simpson, etc.

Value

The Hill number of the population given by input.

true_inverse_simpson	<i>Calculate the true Inverse Simpson index</i>
----------------------	---

Description

Calculate the true Inverse Simpson index

Usage

```
true_inverse_simpson(input)
```

Arguments

input	A vector of proportions.
-------	--------------------------

Value

The inverse-Simpson index of the population given by input.

Note

This function is intended for population-level data. If you are dealing with a microbial sample, use DivNet instead.

true_shannon	<i>Calculate the true Shannon index based on proportions</i>
--------------	--

Description

Calculate the true Shannon index based on proportions

Usage

```
true_shannon(input)
```

Arguments

input	A vector of proportions.
-------	--------------------------

Value

The Shannon index of the population given by input.

Note

This function is intended for population-level data. If you are dealing with a microbial sample, use DivNet instead.

true_shannon_e	<i>Calculate the true Shannon's equitability index</i>
----------------	--

Description

Calculate the true Shannon's equitability index

Usage

```
true_shannon_e(input)
```

Arguments

input	A vector of proportions.
-------	--------------------------

Value

The Shannon E's of the population given by input.

Note

This function is intended for population-level data. If you are dealing with a microbial sample, use DivNet instead.

true_simpson	<i>Calculate the true Simpson index</i>
--------------	---

Description

Calculate the true Simpson index

Usage

```
true_simpson(input)
```

Arguments

input	A vector of proportions.
-------	--------------------------

Value

The Simpson index of the population given by input.

Note

This function is intended for population-level data. If you are dealing with a microbial sample, use DivNet instead.

wlrn_transformed	<i>The transformed weighted linear regression estimator for species richness estimation</i>
------------------	---

Description

This function implements the transformed version of the species richness estimation procedure outlined in Rocchetti, Bunge and Bohning (2011).

Usage

```
wlrn_transformed(
  input_data,
  cutoff = NA,
  print = NULL,
  plot = NULL,
  answers = NULL
)
```

Arguments

input_data	An input type that can be processed by <code>convert()</code> or a <code>phyloseq</code> object
cutoff	Maximum frequency count to use
print	Deprecated; only for backwards compatibility
plot	Deprecated; only for backwards compatibility
answers	Deprecated; only for backwards compatibility

Value

An object of class `alpha_estimate`, or `alpha_estimates` for phyloseq objects

Note

While robust to many different structures, model is almost always misspecified. The result is usually implausible diversity estimates with artificially low standard errors. Extreme caution is advised.

Author(s)

Amy Willis

References

Rocchetti, I., Bunge, J. and Bohning, D. (2011). Population size estimation based upon ratios of recapture probabilities. *Annals of Applied Statistics*, **5**.

See Also

[breakaway](#); [apples](#); [wlrn_untransformed](#)

Examples

```
wlrn_transformed(apples)
wlrn_transformed(apples, plot = FALSE, print = FALSE, answers = TRUE)
```

wlrn_untransformed	<i>The untransformed weighted linear regression estimator for species richness estimation</i>
--------------------	---

Description

This function implements the untransformed version of the species richness estimation procedure outlined in Rocchetti, Bunge and Bohning (2011).

Usage

```
wlrn_untransformed(
  input_data,
  cutoff = NA,
  print = NULL,
  plot = NULL,
  answers = NULL
)
```

Arguments

<code>input_data</code>	An input type that can be processed by <code>convert()</code> or a phyloseq object
<code>cutoff</code>	Maximum frequency count to use
<code>print</code>	Deprecated; only for backwards compatibility
<code>plot</code>	Deprecated; only for backwards compatibility
<code>answers</code>	Deprecated; only for backwards compatibility

Value

An object of class `alpha_estimate`, or `alpha_estimates` for phyloseq objects

Note

This estimator is based on the negative binomial model and for that reason generally produces poor fits to microbial data. The result is usually artificially low standard errors. Caution is advised.

Author(s)

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References

Rocchetti, I., Bunge, J. and Bohning, D. (2011). Population size estimation based upon ratios of recapture probabilities. *Annals of Applied Statistics*, **5**.

See Also

[breakaway](#); [apples](#); [wlrn_transformed](#)

Examples

```
wlrn_untransformed(apples)
```


Index

- * **datasets**
 - apples, [4](#)
 - hawaii, [21](#)
 - pasolli_et_al, [30](#)
 - soil_phylo, [40](#)
 - toy_metadata, [42](#)
 - toy_otu_table, [43](#)
 - toy_taxonomy, [43](#)
- * **diversity**
 - betta, [5](#)
 - betta_lincom, [8](#)
 - betta_pic, [9](#)
 - betta_random, [10](#)
 - breakaway, [12](#)
 - breakaway_nof1, [13](#)
 - F_test, [19](#)
 - kemp, [22](#)
 - test_submodel, [40](#)
 - wlrm_transformed, [46](#)
 - wlrm_untransformed, [47](#)
- * **error**
 - breakaway_nof1, [13](#)
- * **microbial**
 - breakaway, [12](#)
 - breakaway_nof1, [13](#)
 - kemp, [22](#)
- * **models**
 - breakaway, [12](#)
 - breakaway_nof1, [13](#)
 - kemp, [22](#)
 - wlrm_transformed, [46](#)
 - wlrm_untransformed, [47](#)
- * **nonlinear**
 - breakaway, [12](#)
 - breakaway_nof1, [13](#)
 - kemp, [22](#)
- alpha_estimate, [3](#)
- alpha_estimates, [4](#)
- apples, [4](#), [7](#), [13](#), [14](#), [23](#), [47](#), [48](#)
- betta, [5](#), [8](#), [10](#), [11](#), [20](#), [41](#)
- betta_lincom, [8](#)
- betta_pic, [9](#)
- betta_random, [10](#), [20](#), [41](#)
- breakaway, [6](#), [7](#), [12](#), [13](#), [14](#), [23](#), [47](#), [48](#)
- breakaway_nof1, [7](#), [13](#), [13](#), [23](#)
- build_frequency_count_tables, [15](#)
- chao1, [15](#)
- chao1_bc, [16](#)
- chao_bunge, [17](#)
- chao_shen, [17](#)
- check_format, [18](#)
- convert, [18](#)
- cutoff_wrap, [19](#)
- F_test, [19](#), [41](#)
- get_F_stat, [20](#), [20](#)
- good_turing, [21](#)
- hawaii, [21](#)
- kemp, [13](#), [22](#)
- make_design_matrix, [23](#)
- make_frequency_count_table, [24](#)
- objective_bayes_geometric, [24](#)
- objective_bayes_mixedgeo, [26](#)
- objective_bayes_negbin, [27](#)
- objective_bayes_poisson, [29](#)
- pasolli_et_al, [30](#)
- physeq_wrap, [31](#)
- plot.alpha_estimates, [31](#)
- poisson_model, [32](#)
- poisson_model_nof1, [33](#)
- proportions_instead, [33](#)
- rnbinomtable, [34](#)
- rztnbinomtable, [34](#)
- sample_inverse_simpson, [35](#)
- sample_richness, [35](#)
- sample_shannon, [36](#)
- sample_shannon_e, [36](#)
- sample_simpson, [37](#)

`sample_size_estimate`, [37](#)
`sample_size_figure`, [38](#)
`simulate_betta`, [39](#)
`simulate_betta_random`, [39](#)
`soil_phylo`, [40](#)

`test_submodel`, [40](#)
`toy_metadata`, [42](#)
`toy_otu_table`, [43](#)
`toy_taxonomy`, [43](#)
`true_gini`, [43](#)
`true_hill`, [44](#)
`true_inverse_simpson`, [44](#)
`true_shannon`, [45](#)
`true_shannon_e`, [45](#)
`true_simpson`, [46](#)

`wlrm_transformed`, [46](#), [48](#)
`wlrm_untransformed`, [47](#), [47](#)