

Package ‘goeveg’

May 10, 2021

Type Package

Title Functions for Community Data and Ordinations

Version 0.5.1

Date 2021-05-10

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Description A collection of functions useful in (vegetation) community analyses and ordinations. Includes automatic species selection for ordination diagrams, NMDS stress plots, species response curves and rank-abundance curves.

License GPL (>= 2)

LazyData TRUE

Depends R (>= 2.10)

Imports vegan, fields, mgcv, Hmisc

Suggests vegdata, BiodiversityR, cluster

URL <https://github.com/fgoral/goeveg/>

BugReports <https://github.com/fgoral/goeveg/issues>

RoxygenNote 7.1.1

NeedsCompilation no

Repository CRAN

Date/Publication 2021-05-10 14:50:02 UTC

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cv	<i>Coefficient of variation (CV)</i>
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Description

Compute the coefficient of variation (CV). The CV, also known as relative standard deviation (RSD), is a standardized measure of dispersion of a probability distribution or frequency distribution. It is defined as the ratio of the standard deviation to the mean and is often expressed as a percentage. In contrast to the standard deviation, it enables comparison between datasets as the CV is independent of the unit in which the measurement has been taken. If `na.rm` is TRUE then missing values are removed before computation proceeds.

Usage

```
cv(x, na.rm = FALSE)
```

Arguments

x	a numeric vector
na.rm	logical. Should missing values be removed?

Value

A numeric scalar – the sample coefficient of variation.

Details

The coefficient of variation (CV) should be computed only for data measured on a ratio scale, as these are the measurements that can only take non-negative values. The CV may not have any meaning for data on an interval scale.

According to *Dormann 2017* CV-values below 0.05 (5%) indicate very high precision of the data, values above 0.2 (20%) low precision. However, this is considered as a rule of thumb. In studies of highly variable systems (e.g. some ecological studies) CV values above 1 may occur.

The CV of a zero-length vector (after removal of NAs if `na.rm = TRUE`) is not defined and gives an error. If there is only a single value, `sd` is NA and `cv` returns NA.

References

Dormann, C. (2017). Parametrische Statistik. Verteilungen, maximum likelihood und GLM in R. Springer. doi: [10.1007/9783662546840](https://doi.org/10.1007/9783662546840)

"What is the difference between ordinal, interval and ratio variables? Why should I care?" *Graph-Pad Software Inc.* <https://www.graphpad.com/support/faqid/1089/>.

See Also[sd](#)**Examples**

```
## Calculate CV for variable soil depth
cv(schedenenv$soil_depth)
```

`dimcheckMDS`*Stress plot/Scree plot for NMDS*

Description

This function provides a simple plot of stress values for a given number of tested dimensions (default $k = 6$) in NMDS. This stress plot (or scree plot) shows the decrease in ordination stress with an increase in the number of ordination dimensions. It is based on function [metaMDS](#) (vegan package) and uses the monoMDS engine.

Usage

```
dimcheckMDS(
  matrix,
  distance = "bray",
  k = 6,
  trymax = 20,
  autotransform = TRUE
)
```

Arguments

<code>matrix</code>	Community data, a matrix-like object with samples in rows and species in columns.
<code>distance</code>	Dissimilarity index used in <code>vegdist</code> .
<code>k</code>	Number of dimensions (default $k = 6$).
<code>trymax</code>	Maximum number of random configuration for iterative search search of stable solution.
<code>autotransform</code>	Whether to use transformation (see metaMDS) or not. Default is <code>autotransform = TRUE</code> .

Value

A numeric vector of length k containing stress values for k dimensions.

Details

Goodness of Non-metric multidimensional scaling (NMDS) is measured by stress value. The lower the stress value, the better fit of original distances/dissimilarities and projected distances in ordination diagram is reached. Stress value depends on dimensionality; it is decreasing with increasing dimensionality. On the other hand, stress-reduction does not mean to maximise interpretation capability. Low-dimensional projections are often better to interpret and are so preferable for interpretation issues. The stress plot (or sometimes also called scree plot) is a diagnostic plots to explore both, dimensionality and interpretative value. It provides dimension-dependent stress reduction and curve estimate gives indices for meaningful stress reduction with increasing dimensionality. Furthermore, another diagnostic plot for detecting best dimension for projection of NMDS, the Shepard diagram ([stressplot](#)) is recommended for detecting best dimensionality in NMDS.

Clarke 1993 suggests the following guidelines for acceptable stress values: <0.05 = excellent, <0.10 = good, <0.20 = usable, >0.20 = not acceptable. The plot shows the border of the 0.20 stress value limit. Solutions with higher stress values should be interpreted with caution and those with stress above 0.30 are highly suspect.

Author(s)

Jenny Schellenberg (<jschell@gwdg.de>) and Friedemann Goral (<fgoral@gwdg.de>)

References

Clarke, K. R. (1993). Non-parametric multivariate analysis of changes in community structure. *Austral J Ecol* **18**: 117-143. doi: [10.1111/j.14429993.1993.tb00438.x](https://doi.org/10.1111/j.14429993.1993.tb00438.x)

See Also

[metaMDS stressplot](#)

Examples

```
## Use of function with default values
dimcheckMDS(schedenveg)

## Use of function for testing 10 dimensions
dimcheckMDS(schedenveg, k = 10)
```

ordiselect

Species selection for ordination plots

Description

This function simplifies the selection of relevant species in ordination diagrams. It works with result objects from vegan package. The selection can be based upon cover abundances, frequency values and/or species fit to multivariate analysis. The result is a vector of names of the selected species and can be used for the select argument in ordination plots.

Usage

```
ordiselect(
  matrix,
  ord,
  ablim = 0.9,
  fitlim = 1,
  choices = c(1, 2),
  method = "axes",
  env,
  p.max = 0.05,
  freq = FALSE,
  na.rm = FALSE
)
```

Arguments

<code>matrix</code>	Community data, a matrix-like object with samples in rows and species in columns.
<code>ord</code>	vegan ordination result object (e.g. from decorana , cca or metaMDS).
<code>ablim</code>	Proportion of species with highest abundances to be displayed. Value between 0 and 1.
<code>fitlim</code>	Proportion of species with best fit to be displayed. Value between 0 and 1.
<code>choices</code>	Axes shown.
<code>method</code>	The species fit method: "axes" or "vars". See details for methods.
<code>env</code>	Fitted environmental variables (result object of envfit). Only used if method = "vars".
<code>p.max</code>	Significance limit for variables used in method = "vars".
<code>freq</code>	Whether to use cover abundances (= default) or frequencies of <code>matrix</code> . If TRUE, frequencies of species are used.
<code>na.rm</code>	Set to TRUE if your ordination object has NAs (e.g. due to selection)

Value

A vector of variable length containing the names of selected species from `matrix`.

Details

Two methods for species fit are implemented.

- In method = "axes" species scores are used for selecting best fitting species. This is the default method. The basic assumption is that species that show high correlations to ordination axes have good fit to the gradient. High scores along ordination axes mean high correlation. In this method, all species with high correlations to ordination axes will be filtered.
- In method = "vars" environmental variables are used for selecting best fitting species. This is a distance-based approach for showing the species with best species-environment-correlation in ordination diagrams. Therefore Euclidean distances between species and environment variable centroids are calculated. Only high-responding species with very close or very far distances are considered.

If `method = "vars"` is used, the environmental variables need to be fitted with `envfit` and the result of this function must be provided to the `env` argument. The `p.max` argument allows selection of only significant variables, default is `p.max = 0.05`.

The two described methods work well both in eigenvalue-based and in distance-based ordinations. But note, that the distance-based approach for species fit is recommended for distance-based methods (e.g. NMDS), in which axes are arbitrary. If axes fit should be applied on distance-based ordination, species scores need to be calculated during the analysis, e.g. by selecting `wascores = TRUE` in `metaMDS`. On the other hand, distance calculation may be meaningless in Eigenvalue-based approaches. However, both methods provide good option of objective reduction of visible species in ordination plot for better interpretation issues.

The default for `matrix` input is a cover-abundance-matrix. This matrix should also be used for ordination.

If no limit is defined for one of the arguments `ablim`, `fitlim`, all species are displayed.

Author(s)

Friedemann Goral (<fgoral@gwdg.de>) and Jenny Schellenberg

Examples

```
## Calculate DCA
library(vegan)
scheden.dca <- decorana(schedenveg)

## Select the 30% most abundant species and call the result
limited <- ordiselect(schedenveg, scheden.dca, ablim = 0.3)
limited

## Use the result in plotting
plot(scheden.dca, display="n")
points(scheden.dca, display="sites")
points(scheden.dca, display="species",
       select = limited, pch=3, col="red", cex=0.7)
ordipointlabel(scheden.dca, display="species",
              select = limited, col="red", cex=0.7, add = TRUE)

## Select the 30% most frequent species with 50% best axis fit
limited <- ordiselect(schedenveg, scheden.dca, ablim = 0.3,
                   fitlim = 0.5, freq = TRUE)

## Select the 30% most abundant species with 60% best environmental fit
## in NDMS for axes 1 & 3
nmads <- metaMDS(schedenveg, k = 3) # run NMDS
env13 <- envfit(nmads, schedenenv[,2:10], choices=c(1,3))
limited13 <- ordiselect(schedenveg, nmads, ablim = 0.3, fitlim = 0.6,
                    choices = c(1,3), method = "vars", env = env13)
```

racurve *Rank-abundance curves*

Description

This function draws a rank-abundance curve for community data. You can optionally add labels for a selected number of species. If you wish to draw multiple rank-abundance curves for selected samples use [racurves](#).

Usage

```
racurve(  
  matrix,  
  main = "Rank-abundance diagram",  
  nlab = 0,  
  ylog = FALSE,  
  frequency = FALSE,  
  ylim = NULL,  
  xlim = NULL  
)
```

Arguments

matrix	Community data, a matrix-like object with samples in rows.
main	The main title (optional).
nlab	Number of labeled species (default = 0). Species are labeled in decreasing order beginning from the highest relative abundance.
ylog	If set on TRUE the y-axis is displayed on a log-scale.
frequency	If set on TRUE frequencies of species are calculated instead of relative abundances.
xlim, ylim	Define axis limits

Value

Returns an (invisible) list composed of:

abund	abundances of each species (in decreasing order)
rel.abund	relative abundances of each species (in decreasing order)
freq	frequency of each species (in decreasing order)

Details

Rank abundance curves or Whittaker plots (see *Whittaker 1965*) are used to display relative species abundance as biodiversity component. They are a means to visualize species richness and species evenness.

Author(s)

Friedemann Goral (<fgoral@gwdg.de>)

References

Whittaker, R. H. (1965). Dominance and Diversity in Land Plant Communities: Numerical relations of species express the importance of competition in community function and evolution. *Science* **147** : 250-260. doi: [10.1126/science.147.3655.250](https://doi.org/10.1126/science.147.3655.250)

See Also

[racurves](#) for multiple curves and [rankabundance](#) from package BiodiversityR for a more sophisticated function

Examples

```
## Draw rank-abundance curve
racurve(schedenveg)

## Draw rank-abundance curve and label first 5 species
racurve(schedenveg, nlab = 5)

## Draw rank-abundance curve with log-scaled axis
racurve(schedenveg, ylog = TRUE)

## Draw rank-abundance curve with frequencies and no main title
racurve(schedenveg, frequency = TRUE, nlab = 1, main = "")
```

racurves

Multiple rank-abundance curves

Description

This function draws multiple rank-abundance curves for selected samples into one diagram. If you wish to draw a simple rank-abundance curve see [racurve](#).

Usage

```
racurves(matrix, main = "Rank-abundance diagram", bw = TRUE)
```

Arguments

<code>matrix</code>	Community data, a matrix-like object with samples in rows and species in columns. Rank-abundance curves are drawn for all selected rows (samples).
<code>main</code>	The main title (optional).
<code>bw</code>	If set on FALSE the lines will be drawn in colors instead of black/white lines with different line types.

Value

No return value, only diagram.

Details

Rank abundance curves or Whittaker plots (see *Whittaker 1965*) are used to display relative species abundance as biodiversity component. They are a means to visualize species richness and species evenness.

The axes of the diagram will be scaled according automatically. As the line type is used to differentiate between samples, a maximum of 6 curves per diagram is feasible in black/white mode.

Author(s)

Friedemann Goral (<fgoral@gwdg.de>)

References

Whittaker, R. H. (1965). Dominance and Diversity in Land Plant Communities: Numerical relations of species express the importance of competition in community function and evolution. *Science* **147** : 250-260. doi: [10.1126/science.147.3655.250](https://doi.org/10.1126/science.147.3655.250)

See Also

[racurve](#) for a simple curve and [rankabundance](#) from package BiodiversityR for a more sophisticated function

Examples

```
## Draw multiple rank-abundance curves for selected samples
racurves(schedenveg[c(1,7,20,25), ])
```

```
## Draw multiple rank-abundance curves for selected samples with coloured lines
racurves(schedenveg[c(1,7,20,25), ], bw = FALSE)
```

schedenenv

Header data for Vegetation releves from Scheden

Description

An example vegetation dataset containing 28 grassland releves from Scheden, Niedersachsen, Germany. The releves were done May 2016 during a students field course at the University of Goettingen. Locations at the study site are based on the diploma thesis from *Eichholz (1997)*

Usage

```
schedenenv
```

Format

A data frame with 28 rows (samples) and 10 variables

- comm: Plant community as defined in 1997: *Arrhenatheretum* or *Gentiano-Koelerietum*
- altit: Altitude (m)
- north: North value as cosinus of aspect
- slope: Slope (degrees)
- cov_herb: Cover of herb layer (%)
- cov_litt: Cover of litter (%)
- cov_moss: Cover of mosses (%)
- cov_opensoil: Cover of open soil (%)
- height_herb: Average height of herb layer (cm)
- soil_depth: Soil depth (cm)

References

Eichholz, A. (1997): Wiesen und Magerrasen am Suedhang des Hohen Hagen. Diplomarbeit Biologie, University of Goettingen.

schedenveg

Vegetation releves from Scheden

Description

An example vegetation dataset containing 28 grassland releves from Scheden, Niedersachsen, Germany. The releves were done May 2016 during a students field course at the University of Goettingen. Locations at the study site are based on the diploma thesis from *Eichholz (1997)*

Usage

schedenveg

Format

A data frame with 28 rows (samples) and 155 variables (species)

References

Eichholz, A. (1997): Wiesen und Magerrasen am Suedhang des Hohen Hagen. Diplomarbeit Biologie, University of Goettingen.

sem	<i>Standard error of the mean (SEM)</i>
-----	---

Description

Compute the standard error of the mean (SEM). The SEM is the standard deviation of the sample-mean's estimate of a population mean. It therefore describes the accuracy of the calculation of a sample's mean. If `na.rm` is TRUE then missing values are removed before computation proceeds.

Usage

```
sem(x, na.rm = FALSE)
```

Arguments

<code>x</code>	a numeric vector
<code>na.rm</code>	logical. Should missing values be removed?

Value

A numeric scalar – the standard error of the mean.

Details

The SEM of a zero-length vector (after removal of NAs if `na.rm = TRUE`) is not defined and gives an error. The SEM of a length-one vector is NA.

See Also

[sd](#)

Examples

```
## Calculate mean and SEM for variable soil depth
mean(schedenenv$soil_depth)
sem(schedenenv$soil_depth)
```

specresponse

*Species response curves***Description**

This function fits species response curves to visualize species responses to environmental gradients or ordination axes. It is based on Logistic Regression using Generalized Linear Models (GLMs) or Generalized Additive Models (GAMs) with integrated smoothness estimation. The function can draw response curves for single or multiple species.

Usage

```
specresponse(
  species,
  var,
  main,
  xlab,
  model = "auto",
  method = "env",
  axis = 1,
  points = FALSE,
  bw = FALSE,
  lwd = NULL
)
```

Arguments

species	Species data (either a community matrix object with samples in rows and species in columns - response curves are drawn for all (selected) columns; or a single vector containing species abundances per plot).
var	Vector containing environmental variable (per plot) OR vegan ordination result object if method = "ord".
main	Optional: Main title.
xlab	Optional: Label of x-axis.
model	Defining the assumed species response: Default model = "auto" selects the model automatically based on AIC. Other methods are model = "linear" (linear response), model = "unimodal" (unimodal response), model = "bimodal" (bimodal response) and model = "gam" (using GAM with regression smoother).
method	Method defining the type of variable. Default method = "env" fits a response curve to environmental variables. Alternatively method = "ord" fits a response along ordination axes.
axis	Ordination axis (only if method = "ord").
points	If set on TRUE the species occurrences are shown as transparent points (the darker the point the more samples at this x-value). To avoid overlapping they are shown with vertical offset.

bw	If set on TRUE the lines will be drawn in black/white with different line types instead of colors.
lwd	Optional: Graphical parameter defining the line width.

Value

Returns character string with information on model type and parameters per species.

No return if model type predefined *linear* or *unimodal*.

Details

For response curves based on environmental gradients the argument `var` takes a single vector containing the variable corresponding to the species abundances.

For a response to ordination axis (`method = "ord"`) the argument `var` requires a vegan ordination result object (e.g. from `decorana`, `cca` or `metaMDS`). First axis is used as default.

By default the response curves are drawn with automatic GLM model selection based on AIC out of GLMs with 1 - 3 polynomial degrees (thus excluding bimodal responses which must be manually defined). The GAM model is more flexible and chooses automatically between an upper limit of 3 - 6 degrees of freedom for the regression smoother.

Available information about species is reduced to presence-absence as species abundances can contain much noise (being affected by complex factors) and the results of Logistic Regression are easier to interpret showing the "probabilities of occurrence". Be aware that response curves are only a simplification of reality (model) and their shape is strongly dependent on the available dataset.

Author(s)

Friedemann Goral (<fgoral@gwdg.de>)

Examples

```
## Draw species response curve for one species on environmental variable
## with points of occurrences
specresponse(schedenveg$ArrElat, schedenenv$soil_depth, points = TRUE)

## Draw species response curve on environmental variable with custom labels
specresponse(schedenveg$ArrElat, schedenenv$soil_depth, points = TRUE,
             main = "Arrhenatherum elatius", xlab = "Soil depth")

## Draw species response curve on ordination axes
## First calculate DCA
library(vegan)
scheden.dca <- decorana(schedenveg)

# Using a linear model on first axis
specresponse(schedenveg$ArrElat, scheden.dca, method = "ord", model = "linear")
# Using an unimodal model on second axis
specresponse(schedenveg$ArrElat, scheden.dca, method = "ord", axis = 2, model = "unimodal")

## Community data: species (columns) need to be selected; call names() to get column numbers
```

```
names(schedenveg)
## Draw multiple species response curves on variable in black/white
specresponse(schedenveg[ ,c(9,18,14,19)], schedenenv$height_herb, bw = TRUE)

## Draw the same curves based on GAM
specresponse(schedenveg[ ,c(9,18,14,19)], schedenenv$height_herb, bw = TRUE, model = "gam")

## Draw multiple species response curves on variable with
## custom x-axis label and points of occurrences
specresponse(schedenveg[ ,c(9,18,14,19)], schedenenv$height_herb,
             xlab = "Height of herb layer (cm)", points = TRUE)

## Draw multiple species response curves on ordination axes
specresponse(schedenveg[ ,c(9,18,14,19)], scheden.dca, method = "ord")
specresponse(schedenveg[ ,c(9,18,14,19)], scheden.dca, method = "ord", axis = 2)
```

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