

# Package ‘ecospat’

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**Depends** ade4, ape, gam, gbm, sp

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**Suggests** MigClim, rgdal, rJava, XML

**LazyData** true

**URL** <http://www.unil.ch/ecospat/home/menuinst/tools--data/tools.html>

**Description** R package for the support of spatial analyses and modelling of species niches and distributions, written by current and former members and collaborators of the ecospat group of Antoine Guisan, Department of Ecology and Evolution (DEE) & Institute of Earth Surface Dynamics (IDYST), University of Lausanne, Switzerland.

**License** GPL

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## Description

R package for the support of spatial analyses and modelling of species niches and distributions, written by current and former members and collaborators of the ecospat group of Antoine Guisan, Department of Ecology and Evolution (DEE) & Institute of Earth Surface Dynamics (IDYST), University of Lausanne, Switzerland.

**ecospat** offers the possibility to perform Pre-modelling Analysis, such as Spatial autocorrelation analysis, MESS (Multivariate Environmental Similarity Surfaces) analyses, Phylogenetic diversity Measures, Biotic Interactions. It also provides functions to complement **biomod2** in preparing the data, calibrating and evaluating (e.g. boyce index) and projecting the models. Complementary analysis based on model predictions (e.g. co-occurrences analyses) are also provided.

In addition, the **ecospat** package includes Niche Quantification and Overlap functions that were used in Broennimann et al. 2012 and Petitpierre et al. 2012 to quantify climatic niche shifts between the native and invaded ranges of invasive species.

**ecospat** can also be complemented with the **MigClim** package, developed by members of the same group.

## Details

Package: ecospat  
 Type: Package  
 Version: 1.1  
 Date: 2015-03-06  
 License: GPL

## Author(s)

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---

ecospat.binary.model    *Generate Binary Models*

---

## Description

Generate a binary map from a continuous model prediction (i.e., values from 0 to 1000).

## Usage

ecospat.binary.model (Pred, Sp.occ.xy, Percentage)

## Arguments

Pred	Predicted suitability values (from 0 to 1000). A RasterStack object containing models predictions ( Output from biomod2 in raster format).
Sp.occ.xy	Ocurrences of the species. A dataframe object with two columns: longitude and latitude. Coordinate systems other than longitude and latitude can be used, for example "x" and "y".
Percentage	The percentage of omission error used to generate the binary model.

## Details

This function generates a binary model prediction (presence/absence) from an original model (continuous values from 0 to 1000) applying a threshold of maximum acceptable error of false negatives (i.e. percentage of the presence predicted as absences, omission error).

## Value

The binary model prediction (presence/absence) generated.

## Author(s)

Ruben G. Mateo <rubeng.mateo@gmail.com>

## References

- Fielding, A.H. and J.F. Bell. 1997. A review of methods for the assessment of prediction errors in conservation presence/absence models. *Environmental Conservation*, **24**: 38-49.
- Engler, R., A. Guisan and L. Rechsteiner. 2004. An improved approach for predicting the distribution of rare and endangered species from occurrence and pseudo-absence data. *Journal of Applied Ecology*, **41**, 263-274.
- Guisan, A., O. Broennimann, R. Engler, M. Vust, N.G. Yoccoz, A. Lehmann and N.E. Zimmermann. 2006. Using niche-based models to improve the sampling of rare species. *Conservation Biology*, **20**, 501-511.

## Examples

```
## Not run:

#Run biomod2 to produce a model prediction
DataSpecies <- read.csv(system.file("external/species/mammals_table.csv", package="biomod2"))

myRespName <- GuloGulo
# the presence/absences data for our species
myResp <- as.numeric(DataSpecies[,myRespName])
# the XY coordinates of species data
myRespXY <- DataSpecies[,c("X_WGS84", "Y_WGS84")]
# load the environmental raster layers (could be .img, ArcGIS
# rasters or any supported format by the raster package)
# Environmental variables extracted from Worldclim (bio_3, bio_4,
# bio_7, bio_11 & bio_12)
myExpl = stack( system.file( "external/bioclim/current/bio3.grd",
                             package="biomod2"),
                 system.file( "external/bioclim/current/bio4.grd",
```

```

                                package="biomod2"),
  system.file( "external/bioclim/current/bio7.grd",
                                package="biomod2"),
  system.file( "external/bioclim/current/bio11.grd",
                                package="biomod2"),
  system.file( "external/bioclim/current/bio12.grd",
                                package="biomod2"))

myBiomodData <- BIOMOD_FormatingData(resp.var = myResp,
                                    expl.var = myExpl,
                                    resp.xy = myRespXY,
                                    resp.name = myRespName)

myBiomodData
myBiomodOption <- BIOMOD_ModelingOptions()

myBiomodModelOut <- BIOMOD_Modeling(
  myBiomodData,
  models = c(GLM),
  models.options = myBiomodOption,
  NbRunEval=1,
  DataSplit=80,
  Prevalence=0.5,
  VarImport=3,
  models.eval.meth = c(TSS,ROC),
  SaveObj = TRUE,
  rescal.all.models = TRUE,
  do.full.models = FALSE,
  modeling.id = paste(myRespName,"FirstModeling",sep=""))

myBiomodModelOut
myBiomodModelEval <- get_evaluations(myBiomodModelOut)

myBiomodEM <- BIOMOD_EnsembleModeling(
  modeling.output = myBiomodModelOut,
  chosen.models = all,
  em.by=all,
  eval.metric = c(TSS),
  eval.metric.quality.threshold = c(0.7),
  prob.mean = TRUE,
  prob.cv = TRUE,
  prob.ci = TRUE,
  prob.ci.alpha = 0.05,
  prob.median = TRUE,
  committee.averaging = TRUE,
  prob.mean.weight = TRUE,
  prob.mean.weight.decay = proportional )

myBiomodEM

myBiomodProj <- BIOMOD_Projection(
  modeling.output = myBiomodModelOut,
  new.env = myExpl,
  proj.name = current,
  selected.models = all,
  binary.meth = TSS,
  compress = xz,

```

```

    clamping.mask = FALSE,
    output.format = .grd)

myBiomodProj
plot(myBiomodProj, str.grep = GLM)
#
Pred <-get_predictions(myBiomodProj)
Sp.occ.xy <- DataSpecies[DataSpecies[,5]==1,2:3]
Percentage <- 7

binary.model<-ecospat.binary.model (Pred, Sp.occ.xy, Percentage)
plot(binary.model)

## End(Not run)

```

---

ecospat.boyce

*Calculate Boyce Index*


---

## Description

Calculate the Boyce index as in Hirzel et al. (2006). The Boyce index is used to assess model performance.

## Usage

```
ecospat.boyce (fit, obs, nclass=0, window.w="default", res=100, PEplot=T)
```

## Arguments

fit	A vector containing the predicted suitability values
obs	A vector containing the predicted suitability values of the validation points (presence records)
nclass	The number of classes or vector with class thresholds. If nclass=0, the Boyce index is calculated with a moving window (see next parameters)
window.w	The width of the moving window (by default 1/10 of the suitability range)
res	The resolution of the moving window (by default 100 focals)
PEplot	If true, plot the predicted to expected ratio along the suitability class

## Details

The Boyce index only requires presences and measures how much model predictions differ from random distribution of the observed presences across the prediction gradients (Boyce et al. 2002). It is thus the most appropriate metric in the case of presence-only models. It is continuous and varies between -1 and +1. Positive values indicate a model which present predictions are consistent with the distribution of presences in the evaluation dataset, values close to zero mean that the model is not different from a random model, negative values indicate counter predictions, i.e., predicting poor quality areas where presences are more frequent (Hirzel et al. 2006).

**Value**

Returns the predicted-to-expected ratio for each class-interval: F.ratio

Returns the Boyce index value: Spearman.cor

Creates a graphical plot of the predicted to expected ratio along the suitability class

**Author(s)**

Blaise Petitpierre <bpetitpierre@gmail.com> and Frank Breiner <frank.breiner@unil.ch>

**References**

Boyce, M.S., P.R. Vernier, S.E. Nielsen and F.K.A. Schmiegelow. 2002. Evaluating resource selection functions. *Ecol. Model.*, **157**, 281-300.

Hirzel, A.H., G. Le Lay, V. Helfer, C. Randin and A. Guisan. 2006. Evaluating the ability of habitat suitability models to predict species presences. *Ecol. Model.*, **199**, 142-152.

**Examples**

```
obs <- (ecospat.testData$glm_Saxifraga_oppositifolia
[which(ecospat.testData$Saxifraga_oppositifolia==1)])

ecospat.boyce (fit = ecospat.testData$glm_Saxifraga_oppositifolia , obs, nclass=0,
window.w="default", res=100, PEplot=TRUE)
```

---

ecospat.calculate.pd    *Calculate Phylogenetic Diversity Measures*

---

**Description**

Calculate all phylogenetic diversity measures listed in Schweiger et al., 2008 (see full reference below).

**Usage**

```
ecospat.calculate.pd (tree, data, method="spanning", type="clade", root=FALSE,
average=FALSE, verbose=TRUE)
```

**Arguments**

tree	The phylogenetic tree
data	A presence or absence (binary) matrix for each species (columns) in each location or grid cell (rows)
method	The method to use. Options are "pairwise", "topology", and "spanning". Default is "spanning".
type	Phylogenetic measure from those listed in Schweiger et al 2008. Options are "Q", "P", "W", "clade", "species", "J", "F", "AvTD", "TTD", "Dd". Default is "clade".
root	Phylogenetic diversity can either be rooted or unrooted. Details in Schweiger et al 2008. Default is FALSE.

average	Phylogenetic diversity can either be averaged or not averaged. Details in Schweiger et al 2008. Default is FALSE.
verbose	Boolean indicating whether to print progress output during calculation. Default is TRUE.

### Details

Given a phylogenetic tree and a presence/absence matrix this script calculates phylogenetic diversity of a group of species across a given set of grid cells or locations. The library "ape" is required to read the tree in R. Command is "read.tree" or "read.nexus". Options of type: "P" is a normalized measure of "Q". "clade" is "PDnode" when root= FALSE, and is "PDroot" when root =TRUE. "species" is "AvPD".

### Value

This function returns a list of phylogenetic diversity values for each of the grid cells in the presence/absence matrix

### Author(s)

Nicolas Salamin <nicolas.salamin@unil.ch> and Dorothea Pio <Dorothea.Pio@fauna-flora.org>

### References

- Schweiger, O., S. Klotz, W. Durka and I. Kuhn. 2008. A comparative test of phylogenetic diversity indices. *Oecologia*, **157**, 485-495.
- Pio, D.V., O. Broennimann, T.G. Barraclough, G. Reeves, A.G. Rebelo, W. Thuiller, A. Guisan and N. Salamin. 2011. Spatial predictions of phylogenetic diversity in conservation decision making. *Conservation Biology*, **25**, 1229-1239.
- Pio, D.V., R. Engler, H.P. Linder, A. Monadjem, F.P.D. Cotterill, P.J. Taylor, M.C. Schoeman, B.W. Price, M.H. Villet, G. Eick, N. Salamin and A. Guisan. 2014. Climate change effects on animal and plant phylogenetic diversity in southern Africa. *Global Change Biology*, **20**, 1538-1549.

### Examples

```
fpath <- system.file("extdata", "ecospat.testTree.tre", package="ecospat")
tree<-read.tree(fpath)
data <- ecospat.testData[9:52]

pd<- ecospat.calculate.pd(tree, data, method = "spanning", type = "species", root = FALSE,
average = FALSE, verbose = TRUE )

plot(pd)
```

---

ecospat.caleval

*Calibration And Evaluation Dataset*

---

### Description

Generate an evaluation and calibration dataset with a desired ratio of disaggregation.



**Usage**

```
ecospat.caleval (data, xy, row.num=1:nrow(data), nrep=1, ratio=0.7,
disaggregate=0, pseudoabs=0, npres=0, replace=F)
```

**Arguments**

data	A vector with presence-absence (0-1) data for one species.
xy	The x and y coordinates of the projection dataset.
row.num	Row original number
nrep	Number of repetitions
ratio	Ratio of disaggregation
disaggregate	Minimum distance of disaggregation (has to be in the same scale as xy)
pseudoabs	Number of pseudoabsences
npres	To select a smaller number of presences from the dataset to be subsetted. The maximum number is the total number of presences
replace	F to replace de pseudoabsences

**Details**

This functions generates two list, one with the calibration or training dataset and other list with the evaluation or testing dataset disaggregated with a minimum distance.

**Value**

```
list("eval"=eval,"cal"=cal))
```

**Author(s)**

Blaise Petitpierre <bpetitpierre@gmail.com>

**Examples**

```
data<-ecospat.testData
caleval<-ecospat.caleval (data = ecospat.testData[53], xy = data[2:3], row.num = 1:nrow(data),
nrep = 2, ratio = 0.7, disaggregate = 0.2, pseudoabs = 100, npres = 10, replace = FALSE)
caleval
```

---

ecospat.cons\_Cscore      *Constrained co-occurrence analysis.*

---

**Description**

Co-occurrence Analysis & Environmentally Constrained Null Models. The function tests for non-random patterns of species co-occurrence in a presence-absence matrix. It calculates the C-score index for the whole community and for each species pair. An environmental constraint is applied during the generation of the null communities.

**Usage**

```
ecospat.cons_Cscore(presence,pred,nbpermut,outputpath)
```

## Arguments

presence	A presence-absence dataframe for each species (columns) in each location or grid cell (rows) Column names (species names) and row names (sampling plots).
pred	A dataframe object with SDM predictions. Column names (species names SDM) and row names (sampling plots).
nbpermut	The number of permutation in the null model.
outpath	Path to specify where to save the results.

## Details

An environmentally constrained approach to null models will provide a more robust evaluation of species associations by facilitating the distinction between mutually exclusive processes that may shape species distributions and community assembly. The format required for input databases: a plots (rows) x species (columns) matrix. Input matrices should have column names (species names) and row names (sampling plots). NOTE: a SES that is greater than 2 or less than -2 is statistically significant with a tail probability of less than 0.05 (Gotelli & McCabe 2002 - Ecology)

## Value

Returns the C-score index for the observed community (ObsCscoreTot), the mean of C-score for the simulated communities (SimCscoreTot), p.value (PValTot) and standardized effect size (SES.Tot). It also saves a table in the specified path where the same metrics are calculated for each species pair (only the table with species pairs with significant p.values is saved in this version).

## Author(s)

Anne Dubuis <anne.dubuis@gmail.com> and Manuela D'Amen <manuela.damen@unil.ch>

## References

- Gotelli, N.J. and D.J. McCabe. 2002. Species co-occurrence: a meta-analysis of JM Diamond's assembly rules model. *Ecology*, **83**, 2091-2096.
- Peres-Neto, P.R., J.D. Olden and D.A. Jackson. 2001. Environmentally constrained null models: site suitability as occupancy criterion. *Oikos*, **93**, 110-120.

## Examples

```
## Not run:
presence<-ecospat.testData[c(53,62,58,70,61,66,65,71,69,43,63,56,68,57,55,60,54,67,59,64)]
pred<-ecospat.testData[c(73:92)]
nbpermut<-10000
outpath<-getwd()
ecospat.cons_Cscore(presence, pred, nbpermut, outpath)

## End(Not run)
```

---

`ecospat.co_occurrences`*Species Co-occurrences*

---

## Description

Calculate an index of species co-occurrences.

## Usage

```
ecospat.co_occurrences (data)
```

## Arguments

data	A presence-absence matrix for each species (columns) in each location or grid cell (rows) or a matrix with predicted suitability values.
------	------------------------------------------------------------------------------------------------------------------------------------------

## Details

Computes an index of co-occurrences ranging from 0 (never co-occurring) to 1 (always co-occurring).

## Value

The species co-occurrence matrix and box-plot of the co-occurrence indices

## Author(s)

Loic Pellissier <loic.pellissier@unifr.ch>

## References

Pellissier, L., K.A. Brathen, J. Pottier, C.F. Randin, P. Vittoz, A. Dubuis, N.G. Yoccoz, T. Alm, N.E. Zimmermann and A. Guisan. 2010. Species distribution models reveal apparent competitive and facilitative effects of a dominant species on the distribution of tundra plants. *Ecography*, **33**, 1004-1014.

Guisan, A. and N. Zimmermann. 2000. Predictive habitat distribution models in ecology. *Ecological Modelling*, **135**:147-186

## Examples

```
## Not run:
matrix <- ecospat.testData[c(9:16,54:57)]
ecospat.co_occurrences (data=matrix)

## End(Not run)
```

---

ecospat.cv.example	<i>Cross Validation Example Function</i>
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---

### Description

Run the cross validation functions on an example data set.

### Usage

```
ecospat.cv.example ()
```

### Details

This function takes an example data set, calibrates it for various models, and then runs the cross validation functions on the results. Mainly to show how to use the cross validation functions.

### Author(s)

Christophe Randin <christophe.randin@unibas.ch> and Antoine Guisan <antoine.guisan@unil.ch>

### Examples

```
## Not run:
ecospat.cv.example ()

## End(Not run)
```

---

ecospat.cv.gam	<i>GAM Cross Validation</i>
----------------	-----------------------------

---

### Description

K-fold and leave-one-out cross validation for GAM.

### Usage

```
ecospat.cv.gam (gam.obj, K=10, cv.lim=10, jack.knife=F)
```

### Arguments

gam.obj	Any calibrated GAM object with a binomial error distribution.
K	Number of folds. 10 is recommended; 5 for small data sets.
cv.lim	Minimum number of presences required to perform the K-fold cross-validation.
jack.knife	If TRUE, then the leave-one-out / jackknife cross-validation is performed instead of the 10-fold cross-validation.

## Details

This function takes a calibrated GAM object with a binomial error distribution and returns predictions from a stratified 10-fold cross-validation or a leave-one-out / jack-knived cross-validation. Stratified means that the original prevalence of the presences and absences in the full dataset is conserved in each fold.

## Value

Returns a dataframe with the observations (obs) and the corresponding predictions by cross-validation or jackknife.

## Author(s)

Christophe Randin <christophe.randin@unibas.ch> and Antoine Guisan <antoine.guisan@unil.ch>

## References

- Randin, C.F., T. Dirnbock, S. Dullinger, N.E. Zimmermann, M. Zappa and A. Guisan. 2006. Are niche-based species distribution models transferable in space? *Journal of Biogeography*, **33**, 1689-1703.
- Pearman, P.B., C.F. Randin, O. Broennimann, P. Vittoz, W.O. van der Knaap, R. Engler, G. Le Lay, N.E. Zimmermann and A. Guisan. 2008. Prediction of plant species distributions across six millennia. *Ecology Letters*, **11**, 357-369.

## Examples

```
## Not run:
gam<-ecospat.cv.gam (gam.obj= get ("gam.Agrostis_capillaris", envir=ecospat.env),
K=10, cv.lim=10, jack.knife=FALSE)

## End(Not run)
```

---

ecospat.cv.gbm

*GBM Cross Validation*

---

## Description

K-fold and leave-one-out cross validation for GBM.

## Usage

```
ecospat.cv.gbm (gbm.obj, data.cv, K=10, cv.lim=10, jack.knife=F)
```

## Arguments

- |            |                                                                                                                                            |
|------------|--------------------------------------------------------------------------------------------------------------------------------------------|
| gbm.obj    | A calibrated GBM object with a binomial error distribution. Attention: users have to tune model input parameters according to their study! |
| data.cv    | A dataframe object containing the calibration data set with the same names for response and predictor variables.                           |
| K          | Number of folds. 10 is recommended; 5 for small data sets.                                                                                 |
| cv.lim     | Minimum number of presences required to perform the K-fold cross-validation.                                                               |
| jack.knife | If TRUE, then the leave-one-out / jackknife cross-validation is performed instead of the 10-fold cross-validation.                         |

## Details

This function takes a calibrated GBM object with a binomial error distribution and returns predictions from a stratified 10-fold cross-validation or a leave-one-out / jack-knived cross-validation. Stratified means that the original prevalence of the presences and absences in the full dataset is conserved in each fold.

## Value

Returns a dataframe with the observations (obs) and the corresponding predictions by cross-validation or jackknife.

## Author(s)

Christophe Randin <christophe.randin@unibas.ch> and Antoine Guisan <antoine.guisan@unil.ch>

## References

- Randin, C.F., T. Dirnbock, S. Dullinger, N.E. Zimmermann, M. Zappa and A. Guisan. 2006. Are niche-based species distribution models transferable in space? *Journal of Biogeography*, **33**, 1689-1703.
- Pearman, P.B., C.F. Randin, O. Broennimann, P. Vittoz, W.O. van der Knaap, R. Engler, G. Le Lay, N.E. Zimmermann and A. Guisan. 2008. Prediction of plant species distributions across six millennia. *Ecology Letters*, **11**, 357-369.

## Examples

```
## Not run:
gbm<-ecospat.cv.gbm (gbm.obj= get ("gbm.Agrostis_capillaris", envir=ecospat.env),
ecospat.testData, K=10, cv.lim=10, jack.knife=FALSE)

## End(Not run)
```

---

ecospat.cv.glm

*GLM Cross Validation*

---

## Description

K-fold and leave-one-out cross validation for GLM.

## Usage

```
ecospat.cv.glm (glm.obj, K=10, cv.lim=10, jack.knife=F)
```

## Arguments

glm.obj	Any calibrated GLM object with a binomial error distribution.
K	Number of folds. 10 is recommended; 5 for small data sets.
cv.lim	Minimum number of presences required to perform the K-fold cross-validation.
jack.knife	If TRUE, then the leave-one-out / jackknife cross-validation is performed instead of the 10-fold cross-validation.

## Details

This function takes a calibrated GLM object with a binomial error distribution and returns predictions from a stratified 10-fold cross-validation or a leave-one-out / jack-knived cross-validation. Stratified means that the original prevalence of the presences and absences in the full dataset is conserved in each fold.

## Value

Returns a dataframe with the observations (obs) and the corresponding predictions by cross-validation or jackknife.

## Author(s)

Christophe Randin <christophe.randin@unibas.ch> and Antoine Guisan <antoine.guisan@unil.ch>

## References

- Randin, C.F., T. Dirnbock, S. Dullinger, N.E. Zimmermann, M. Zappa and A. Guisan. 2006. Are niche-based species distribution models transferable in space? *Journal of Biogeography*, **33**, 1689-1703.
- Pearman, P.B., C.F. Randin, O. Broennimann, P. Vittoz, W.O. van der Knaap, R. Engler, G. Le Lay, N.E. Zimmermann and A. Guisan. 2008. Prediction of plant species distributions across six millennia. *Ecology Letters*, **11**, 357-369.

## Examples

```
## Not run:
glm<-ecospat.cv.glm (glm.obj = get ("glm.Agrostis_capillaris", envir=ecospat.env),
K=10, cv.lim=10, jack.knife=FALSE)

## End(Not run)
```

---

ecospat.cv.me

*Maxent Cross Validation*

---

## Description

K-fold and leave-one-out cross validation for Maxent.

## Usage

```
ecospat.cv.me (data.cv.me, name.sp, names.pred, K=10, cv.lim=10, jack.knife=F)
```

## Arguments

data.cv.me	A dataframe object containing the calibration data set of a Maxent object to validate with the same names for response and predictor variables.
name.sp	Name of the species / response variable.
names.pred	Names of the predicting variables.
K	Number of folds. 10 is recommended; 5 for small data sets.

cv.lim	Minimum number of presences required to perform the K-fold cross-validation.
jack.knife	If TRUE, then the leave-one-out / jackknife cross-validation is performed instead of the 10-fold cross-validation.

### Details

This function takes a calibrated Maxent object with a binomial error distribution and returns predictions from a stratified 10-fold cross-validation or a leave-one-out / jack-knived cross-validation. Stratified means that the original prevalence of the presences and absences in the full dataset is conserved in each fold.

### Value

Returns a dataframe with the observations (obs) and the corresponding predictions by cross-validation or jackknife.

### Author(s)

Christophe Randin <christophe.randin@unibas.ch> and Antoine Guisan <antoine.guisan@unil.ch>

### References

Randin, C.F., T. Dirnbock, S. Dullinger, N.E. Zimmermann, M. Zappa and A. Guisan. 2006. Are niche-based species distribution models transferable in space? *Journal of Biogeography*, **33**, 1689-1703.

Pearman, P.B., C.F. Randin, O. Broennimann, P. Vittoz, W.O. van der Knaap, R. Engler, G. Le Lay, N.E. Zimmermann and A. Guisan. 2008. Prediction of plant species distributions across six millennia. *Ecology Letters*, **11**, 357-369.

### Examples

```
## Not run:
me <- ecospat.cv.me(ecospat.testData, names(ecospat.testData)[53],
names(ecospat.testData)[4:8], K = 10, cv.lim = 10, jack.knife = FALSE)

## End(Not run)
```

---

ecospat.cv.rf

*RandomForest Cross Validation*

---

### Description

K-fold and leave-one-out cross validation for randomForest.

### Usage

```
ecospat.cv.rf (rf.obj, data.cv, K=10, cv.lim=10, jack.knife=F)
```



## Arguments

<code>rf.obj</code>	Any calibrated randomForest object with a binomial error distribution.
<code>data.cv</code>	A dataframe object containing the calibration data set with the same names for response and predictor variables.
<code>K</code>	Number of folds. 10 is recommended; 5 for small data sets.
<code>cv.lim</code>	Minimum number of presences required to perform the K-fold cross-validation.
<code>jack.knife</code>	If TRUE, then the leave-one-out / jackknife cross-validation is performed instead of the 10-fold cross-validation.

## Details

This function takes a calibrated randomForest object with a binomial error distribution and returns predictions from a stratified 10-fold cross-validation or a leave-one-out / jack-knived cross-validation. Stratified means that the original prevalence of the presences and absences in the full dataset is conserved in each fold.

## Value

Returns a dataframe with the observations (obs) and the corresponding predictions by cross-validation or jackknife.

## Author(s)

Christophe Randin <christophe.randin@unibas.ch> and Antoine Guisan <antoine.guisan@unil.ch>

## References

Randin, C.F., T. Dirnbock, S. Dullinger, N.E. Zimmermann, M. Zappa and A. Guisan. 2006. Are niche-based species distribution models transferable in space? *Journal of Biogeography*, **33**, 1689-1703.

Pearman, P.B., C.F. Randin, O. Broennimann, P. Vittoz, W.O. van der Knaap, R. Engler, G. Le Lay, N.E. Zimmermann and A. Guisan. 2008. Prediction of plant species distributions across six millennia. *Ecology Letters*, **11**, 357-369.

## Examples

```
## Not run:
rf <- ecospat.cv.rf(get("rf.Agrostis_capillaris", envir = ecospat.env),
  ecospat.testData[, c(53, 4:8)], K = 10, cv.lim = 10, jack.knife = FALSE)

## End(Not run)
```

---

 ecospat.env

*Package Environment*


---

### Description

A package environment that is used to contain certain (local) variables and results, especially those in example functions and data sets.

### Examples

```
ls(envir=ecospat.env)
```

---

ecospat.Epred

*Prediction Mean*


---

### Description

Calculate the mean (or weighted mean) of several predictions.

### Usage

```
ecospat.Epred (x, w=rep(1,ncol(x)), th=0)
```

### Arguments

x	A dataframe object with SDM predictions.
w	Weight of the model, e.g. AUC. The default is 1.
th	Threshold used to binarize.

### Details

The Weighted Average consensus method utilizes pre-evaluation of the predictive performance of the single-models. In this approach, half (i.e. four) of the eight single-models with highest accuracy are selected first, and then a WA is calculated based on the pre-evaluated AUC of the single-models

### Value

A weighted mean binary transformation of the models.

### Author(s)

Blaise Petitpierre <bpetitpierre@gmail.com>

### References

Boyce, M.S., P.R. Vernier, S.E. Nielsen and F.K.A. Schmiegelow. 2002. Evaluating resource selection functions. *Ecol. Model.*, **157**, 281-300.

Marmion, M., M. Parviainen, M. Luoto, R.K. Heikkinen and W. Thuiller. 2009. Evaluation of consensus methods in predictive species distribution modelling. *Diversity and Distributions*, **15**, 59-69.

**Examples**

```
x <- ecospat.testData[c(92,96)]  
mean<-ecospat.Epred (x, w=rep(1,ncol(x)), th=0.5)
```

---

ecospat.fun.arrows	<i>Draw Centroid Arrows</i>
--------------------	-----------------------------

---

**Description**

Draw arrows linking the centroid of the native and exotic (non-native) distribution (continuous line) and between native and invaded extent (dashed line).

**Usage**

```
ecospat.fun.arrows (sp1, sp2, clim1, clim2)
```

**Arguments**

sp1	The scores of the species native distribution along the the two first axes of the PCA.
sp2	The scores of the species invasive distribution along the the two first axes of the PCA.
clim1	The scores of the entire native extent along the the two first axes of the PCA.
clim2	The scores of the entire invaded extent along the the two first axes of the PCA.

**Details**

To compare invasive species niche, the arrow links the centroid of the native and invasive distribution (continuous line) and between native and invaded extent (dashed line).

**Value**

Arrow on the overlap test plot

**Author(s)**

Blaise Petitpierre <bpetitpierre@gmail.com>

---

ecospat.grid.clim.dyn *Dynamic Occurrence Densities Grid*


---

**Description**

Create a grid with occurrence densities along one or two gridded environmental gradients.

**Usage**

```
ecospat.grid.clim.dyn (glob, glob1, sp, R, th.sp, th.env)
```

**Arguments**

glob	A two-column dataframe (or a vector) of the environmental values (in column) for background pixels of the whole study area (in row).
glob1	A two-column dataframe (or a vector) of the environmental values (in column) for the background pixels of the species (in row).
sp	A two-column dataframe (or a vector) of the environmental values (in column) for the occurrences of the species (in row).
R	The resolution of the grid.
th.sp	The quantile of species densities at species occurrences used as a threshold to exclude low species density values.
th.env	The quantile of environmental densities at all study sites used as a threshold to exclude low environmental density values.

**Details**

Using the scores of an ordination (or SDM prediction), create a grid  $z$  of  $R \times R$  pixels (or a vector of  $R$  pixels when using scores of dimension 1 or SDM predictions) with occurrence densities. Only scores of one, or two dimensions can be used.

**Value**

A grid  $z$  of  $R \times R$  pixels (or a vector of  $R$  pixels) with  $z.uncor$  being the density of occurrence of the species, and  $z.cor$  the occupancy of the environment by the species (density of occurrences divided by the density of environment in the study area).

**Author(s)**

Olivier Broennimann <olivier.broennimann@unil.ch> and Blaise Petitpierre <bpetitpierre@gmail.com>

**References**

- Broennimann, O., M.C. Fitzpatrick, P.B. Pearman, B. Petitpierre, L. Pellissier, N.G. Yoccoz, W. Thuiller, M.J. Fortin, C. Randin, N.E. Zimmermann, C.H. Graham and A. Guisan. 2012. Measuring ecological niche overlap from occurrence and spatial environmental data. *Global Ecology and Biogeography*, **21**:481-497.
- Petitpierre, B., C. Kueffer, O. Broennimann, C. Randin, C. Daehler and A. Guisan. 2012. Climatic niche shifts are rare among terrestrial plant invaders. *Science*, **335**:1344-1348.

**See Also**[ecospat.plot.niche.dyn](#)**Examples**

```
## Not run:
spp<-ecospat.testNiche
clim<-ecospat.testData[2:8]

occ.sp_test<-na.exclude(ecospat.sample.envvar(dfsp=spp,colspxy=2:3,colspkept=1:3,dfvar=clim,
colvarxy=1:2,colvar="all",resolution=25))

occ.sp<-cbind(occ.sp_test,spp[,4]) #add species names

# list of species
sp.list<-levels(occ.sp[,1])
sp.nbocc<-c()

for (i in 1:length(sp.list)){sp.nbocc<-c(sp.nbocc,length(which(occ.sp[,1] == sp.list[i])))}
#calculate the nb of occurrences per species

sp.list<-sp.list[sp.nbocc>4] # remove species with less than 5 occurrences
nb.sp<-length(sp.list) #nb of species
ls()
# selection of variables to include in the analyses
# try with all and then try only worldclim Variables
Xvar<-c(3:7)
nvar<-length(Xvar)

#number of iteration for the tests of equivalency and similarity
iterations<-100
#resolution of the gridding of the climate space
R=100
##### PCA-ENVIRONMENT #####
data<-rbind(occ.sp[,Xvar+1],clim[,Xvar])
w<-c(rep(0,nrow(occ.sp)),rep(1,nrow(clim)))
pca.cal <-dudi.pca(data, row.w = w, center = TRUE, scale = TRUE, scannf = FALSE, nf = 2)

##### selection of species #####
sp.list
sp.combn<-combn(1:2,2)

for(i in 1:ncol(sp.combn)) {
  row.sp1<-which(occ.sp[,1] == sp.list[sp.combn[1,i]]) # rows in data corresponding to sp1
  row.sp2<-which(occ.sp[,1] == sp.list[sp.combn[2,i]]) # rows in data corresponding to sp2
  name.sp1<-sp.list[sp.combn[1,i]]
  name.sp2<-sp.list[sp.combn[2,i]]
  # predict the scores on the axes
  scores.clim<- pca.cal$li[(nrow(occ.sp)+1):nrow(data),] #scores for global climate
  scores.sp1<- pca.cal$li[row.sp1,] #scores for sp1
  scores.sp2<- pca.cal$li[row.sp2,] #scores for sp2
}
# calculation of occurrence density and test of niche equivalency and similarity
z1<- ecospat.grid.clim.dyn(scores.clim, scores.clim, scores.sp1,R=100)
z2<- ecospat.grid.clim.dyn(scores.clim, scores.clim, scores.sp2,R=100)
```

```
## End(Not run)
```

---

```
ecospat.makeDataFrame Make Data Frame
```

---

## Description

Create a biomod2-compatible dataframe. The function also enables to remove duplicate presences within a pixel and to set a minimum distance between presence points to avoid autocorrelation. Data from GBIF can be added.

## Usage

```
ecospat.makeDataFrame (spec.list, expl.var, use.gbif=FALSE, precision=NULL,
year=NULL, remdups=TRUE, mindist=NULL, n=1000, type=random, PApoint=NULL,
ext=expl.var, tryf=5)
```

## Arguments

<code>spec.list</code>	Data.frame or Character. The species occurrence information must be a data.frame in the form: <code>\`x-coordinates\`</code> , <code>\`y-coordinates\`</code> and <code>\`species name\`</code> (in the same projection/coordinate system as <code>expl.var</code> !).
<code>expl.var</code>	a RasterStack object of the environmental layers.
<code>use.gbif</code>	Logical. If TRUE presence data from GBIF will be added. It is also possible to use GBIF data only. Default: FALSE. See <code>?gbif dismo</code> for more information. Settings: <code>geo=TRUE</code> , <code>removeZeros=T</code> , all sub-taxa will be used. <code>\`species name\`</code> in <code>spec.list</code> must be in the form: <code>\`genus species\`</code> , <code>\`genus_species\`</code> or <code>\`genus.species\`</code> . If there is no species information available on GBIF an error is returned. Try to change species name (maybe there is a synonym) or switch <code>use.gbif</code> off.
<code>precision</code>	Numeric. Use precision if <code>use.gbif = T</code> to set a minimum precision of the presences which should be added. For <code>precision = 1000</code> e.g. only presences with precision of at least 1000 meter will be added from GBIF. When <code>precision = NULL</code> all presences from GBIF will be used, also presences where precision information is NA.
<code>year</code>	Numeric. Latest year of the collected gbif occurrences. If <code>year=1960</code> only occurrences which were collected since 1960 are used.
<code>remdups</code>	Logical. If TRUE (Default) duplicated presences within a raster pixel will be removed. You will get only one presence per pixel.
<code>mindist</code>	Numeric. You can set a minimum distance between presence points to avoid autocorrelation. <code>mindist spatstat</code> is used to calculate the nearest neighbour (nn) for each point. From the pair of the minimum nn, the point is removed of which the second neighbour is closer. Unit is the same as <code>expl.var</code> .
<code>n</code>	number of Pseudo-Absences. Default 1000.
<code>type</code>	sampling design for selecting Pseudo-Absences. If <code>\`random\`</code> (default) background points are selected with the function <code>randomPoints dismo</code> . When selecting another sampling type ( <code>\`regular\`</code> , <code>\`stratified\`</code> , <code>\`nonaligned\`</code> , <code>\`hexagonal\`</code> , <code>\`clustered\`</code> or <code>\`Fibonacci\`</code> ) <code>spsample sp</code> will be used. This can immensely increase computation time and RAM usage if <code>ext</code> is a raster, especially for big raster layers because it must be converted into a <code>\`SpatialPolygonsDataFrame\`</code> first.

PApoint	data.frame or SpatialPoints. You can use your own set of Pseudo-Absences instead of generating new PAs. Two columns with 'x' and 'y' in the same projection/coordinate system as expl.var!
ext	a Spatial Object or Raster object. Extent from which PAs should be selected from (Default is expl.var).
tryf	numeric > 1. Number of trials to create the requested Pseudo-Absences after removing NA points (if type='random'). See ?randomPoints dismo

## Details

If you use a raster stack as explanatory variable and you want to model many species in a loop with Biomod, forming data will last very long as presences and PA's have to be extracted over and over again from the raster stack. To save computation time, it is better to convert the presences and PAs to a data.frame first.

## Value

A data.frame object which can be used for modeling with the Biomod package.

## Author(s)

Frank Breiner <frank.breiner@unil.ch>

## Examples

```
## Not run:
files <- list.files(path=paste(system.file(package="dismo"),
                              /ex, sep=), pattern=grd, full.names=TRUE )
predictors <- raster::stack(files[c(9,1:8)]) #file 9 has more NA values than
# the other files, this is why we choose it as the first layer (see ?randomPoints)

file <- paste(system.file(package="dismo"), "/ex/bradypus.csv", sep="")
bradypus <- read.table(file, header=TRUE, sep=,)[,c(2,3,1)]
head(bradypus)

random.spec <- cbind(as.data.frame(randomPoints(predictors,50,extf=1)),species="randomSpec")
colnames(random.spec)[1:2] <- c("lon","lat")

spec.list <- rbind(bradypus, random.spec)

df <- ecospat.makeDataFrame(spec.list, expl.var=predictors, n=5000)
head(df)

plot(predictors[[1]])
points(df[df$Bradypus.variegatus==1, c(x,y)])
points(df[df$randomSpec==1, c(x,y)], col="red")

## End(Not run)
```

---

ecospat.mantel.correlogram

*Mantel Correlogram*


---

## Description

Investigate spatial autocorrelation of environmental covariables within a set of occurrences as a function of distance.

## Usage

```
ecospat.mantel.correlogram (dfvar, colxy, n, colvar, max, nclass, nperm)
```

## Arguments

dfvar	A dataframe object with the environmental variables.
colxy	The range of columns for x and y in df.
n	The number of random occurrences used for the test.
colvar	The range of columns for variables in df.
max	The maximum distance to be computed in the correlogram.
nclass	The number of classes of distances to be computed in the correlogram.
nperm	The number of permutations in the randomization process.

## Details

Requires ecodist library. Note that computation time increase tremendously when using more than 500 occurrences (n>500)

## Value

Draws a plot with distance vs. the mantel r value. Black circles indicate that the values are significant different from zero. White circles indicate non significant autocorrelation. The selected distance is at the first white circle where values are non significant different from zero.

## Author(s)

Olivier Broennimann <olivier.broennimann@unil.ch>

## References

Legendre, P. and M.J. Fortin. 1989. Spatial pattern and ecological analysis. *Vegetatio*, **80**, 107-138.

## See Also

[mgram](#)

## Examples

```
ecospat.mantel.correlogram(dfvar=ecospat.testData[c(2:16)],colxy=1:2, n=100, colvar=3:7,
max=1000, nclass=10, nperm=100)
```



---

`ecospat.maxentvarimport`*Maxent Variable Importance*

---

## Description

Calculate the importance of variables for Maxent in the same way Biomod does, by randomly permuting each predictor variable independently, and computing the associated reduction in predictive performance.

## Usage

```
ecospat.maxentvarimport (model, dfvar, nperm)
```

## Arguments

<code>model</code>	The name of the maxent model.
<code>dfvar</code>	A dataframe object with the environmental variables.
<code>nperm</code>	The number of permutations in the randomization process. The default is 5.

## Details

The calculation is made as biomod2 "variables\_importance" function. It's more or less base on the same principle than randomForest variables importance algorithm. The principle is to shuffle a single variable of the given data. Make model prediction with this 'shuffled' data.set. Then we compute a simple correlation (Pearson's by default) between references predictions and the 'shuffled' one. The return score is  $1 - \text{cor}(\text{pred\_ref}, \text{pred\_shuffled})$ . The highest the value, the more influence the variable has on the model. A value of this 0 assumes no influence of that variable on the model. Note that this technique does not account for interactions between the variables.

## Value

a list which contains a data.frame containing variables importance scores for each permutation run.

## Author(s)

Blaise Petitpierre <bpetitpierre@gmail.com>

## Examples

```
## Not run:
model <- get ("me.Achillea_millefolium", envir=ecospat.env)
dfvar <- ecospat.testData[4:8]
nperm <- 5
ecospat.maxentvarimport (model, cal, nperm)
## End(Not run)
```

---

ecospat.mess	<i>MESS</i>
--------------	-------------

---

## Description

Calculate the MESS (i.e. extrapolation) as in Maxent.

## Usage

```
ecospat.mess (proj, cal, w="default")
```

## Arguments

proj	A dataframe object with x, y and environmental variables, used as projection dataset.
cal	A dataframe object with x, y and environmental variables, used as calibration dataset.
w	The weight for each predictor (e.g. variables importance in SDM).

## Details

Shows the variable that drives the multivariate environmental similarity surface (MESS) value in each grid cell.

## Value

MESS	The mess as calculated in Maxent, i.e. the minimal extrapolation values.
MESSw	The sum of negative MESS values corrected by the total number of predictors. If there are no negative values, MESSw is the mean MESS.
MESSneg	The number of predictors on which there is extrapolation.

## Author(s)

Blaise Petitpierre <bpetitpierre@gmail.com>

## References

Elith, J., M. Kearney and S. Phillips. 2010. The art of modelling range-shifting species. *Methods in ecology and evolution*, **1**, 330-342.

## See Also

[ecospat.plot.mess](#)

**Examples**

```
x <- ecospat.testData[c(2,3,4:8)]
proj<- x[1:90,] #A projection dataset.
cal<- x[91:300,] #A calibration dataset

#Create a MESS object
mess.object<-ecospat.mess (proj, cal, w="default")

#Plot MESS
ecospat.plot.mess (xy=proj[c(1:2)], mess.object, cex=1, pch=15)
```

---

ecospat.migclim	<i>Implementing dispersal into species distribution models</i>
-----------------	----------------------------------------------------------------

---

**Description**

Enables the implementation of species-specific dispersal constraints into projections of species distribution models under environmental change and/or landscape fragmentation scenarios.

**Usage**

```
ecospat.migclim ()
```

**Details**

The MigClim model is a cellular automaton originally designed to implement dispersal constraints into projections of species distributions under environmental change and landscape fragmentation scenarios.

**Author(s)**

Robin Engler <robin.engler@gmail.com>, Wim Hordijk <wim@WorldWideWanderings.net> and Loic Pellissier <loic.pellissier@unifr.ch>

**References**

- Engler, R., W. Hordijk and A. Guisan. 2012. The MIGCLIM R package – seamless integration of dispersal constraints into projections of species distribution models. *Ecography*, **35**, 872-878.
- Engler, R. and A. Guisan. 2009. MIGCLIM: predicting plant distribution and dispersal in a changing climate. *Diversity and Distributions*, **15**, 590-601.
- Engler, R., C.F. Randin, P. Vittoz, T. Czaka, M. Beniston, N.E. Zimmermann and A. Guisan. 2009. Predicting future distributions of mountain plants under climate change: does dispersal capacity matter? *Ecography*, **32**, 34-45.

**Examples**

```
ecospat.migclim ()
```

---

ecospat.niche.dyn.index

*Niche Expansion, Stability, and Unfilling*

---

### Description

Calculate niche expansion, stability and unfilling.

### Usage

ecospat.niche.dyn.index (z1, z2, intersection=NA)

### Arguments

z1	A gridclim object for the native distribution.
z2	A gridclim object for the invaded range.
intersection	The quantile of the environmental density used to remove marginal climates. If intersection=NA, the analysis is performed on the whole environmental extent (native and invaded). If intersection=0, the analysis is performed at the intersection between native and invaded range. If intersection=0.05, the analysis is performed at the intersection of the 5th quantile of both native and invaded environmental densities.

### Value

A list of dynamic indices: dynamic.index.w [expansion.index.w, stability.index.w, restriction.index.w]

### Author(s)

Blaise Petitpierre <bpetitpierre@gmail.com>

### See Also

[ecospat.grid.clim.dyn](#)

---

ecospat.niche.equivalency.test

*Niche Equivalency Test*

---

### Description

Run a niche equivalency test (see Warren et al 2008) based on two species occurrence density grids.

### Usage

ecospat.niche.equivalency.test (z1, z2, rep)

**Arguments**

z1	Species 1 occurrence density grid created by <code>ecospat.grid.clim</code> .
z2	Species 2 occurrence density grid created by <code>ecospat.grid.clim</code> .
rep	The number of replications to perform.

**Details**

Compares the observed niche overlap between z1 and z2 to overlaps between random niches z1.sim and z2.sim, which are built from random reallocations of occurrences of z1 and z2.

**Value**

a list with \$obs = observed overlaps, \$sim = simulated overlaps, \$p.D = p-value of the test on D, \$p.I = p-value of the test on I.

**Author(s)**

Olivier Broennimann <olivier.broennimann@unil.ch>

**References**

Broennimann, O., M.C. Fitzpatrick, P.B. Pearman, B. Petitpierre, L. Pellissier, N.G. Yoccoz, W. Thuiller, M.J. Fortin, C. Randin, N.E. Zimmermann, C.H. Graham and A. Guisan. 2012. Measuring ecological niche overlap from occurrence and spatial environmental data. *Global Ecology and Biogeography*, **21**, 481-497.

Warren, D.L., R.E. Glor and M. Turelli. 2008. Environmental niche equivalency versus conservatism: quantitative approaches to niche evolution. *Evolution*, **62**, 2868-2883.

**See Also**

[ecospat.grid.clim.dyn](#), [ecospat.niche.similarity.test](#)

---

`ecospat.niche.overlap` *Calculate Niche Overlap*

---

**Description**

Calculate the overlap metrics D and I based on two species occurrence density grids z1 and z2 created by `ecospat.grid.clim`.

**Usage**

```
ecospat.niche.overlap (z1, z2, cor)
```

**Arguments**

z1	Species 1 occurrence density grid created by <code>ecospat.grid.clim</code> .
z2	Species 2 occurrence density grid created by <code>ecospat.grid.clim</code> .
cor	Correct the occurrence densities of each species by the prevalence of the environments in their range (T = yes, F = no).

**Details**

if cor=F, the z\$uncor objects created by `ecospat.grid.clim` are used to calculate the overlap. if cor=T, the z\$cor objects are used.

**Value**

Overlap values D and I. D is Schoener's overlap metric (Schoener 1970). I is a modified Hellinger metric (Warren et al. 2008)

**Author(s)**

Olivier Broennimann <olivier.broennimann@unil.ch>

**References**

Broennimann, O., M.C. Fitzpatrick, P.B. Pearman, B. Petitpierre, L. Pellissier, N.G. Yoccoz, W. Thuiller, M.J. Fortin, C. Randin, N.E. Zimmermann, C.H. Graham and A. Guisan. 2012. Measuring ecological niche overlap from occurrence and spatial environmental data. *Global Ecology and Biogeography*, **21**, 481-497.

Schoener, T.W. 1968. Anolis lizards of Bimini: resource partitioning in a complex fauna. *Ecology*, **49**, 704-726.

Warren, D.L., R.E. Glor and M. Turelli. 2008. Environmental niche equivalency versus conservatism: quantitative approaches to niche evolution. *Evolution*, **62**, 2868-2883.

**See Also**

[ecospat.grid.clim.dyn](#)

---

ecospat.niche.similarity.test

*Niche Similarity Test*

---

**Description**

Run a niche similarity test (see Warren et al 2008) based on two species occurrence density grids.

**Usage**

```
ecospat.niche.similarity.test (z1, z2, rep, one.sided=T)
```

**Arguments**

z1	Species 1 occurrence density grid created by <code>ecospat.grid.clim</code> .
z2	Species 2 occurrence density grid created by <code>ecospat.grid.clim</code> .
rep	The number of replications to perform.
one.sided	If true, perform a one-sided test, otherwise a two-sided test.

**Details**

Compares the observed niche overlap between  $z_1$  and  $z_2$  to overlaps between  $z_1$  and random niches ( $z_2.sim$ ) as available in the range of  $z_2$  ( $z_2.Z$ ).  $z_2.sim$  has the same pattern as  $z_2$  but the center is randomly translated in the available  $z_2.Z$  space and weighted by  $z_2.Z$  densities.

**Value**

a list with  $\$obs$  = observed overlaps,  $\$sim$  = simulated overlaps,  $\$p.D$  = p-value of the test on D,  $\$p.I$  = p-value of the test on I.

**Author(s)**

Olivier Broennimann <olivier.broennimann@unil.ch>

**References**

Broennimann, O., M.C. Fitzpatrick, P.B. Pearman, B. Petitpierre, L. Pellissier, N.G. Yoccoz, W. Thuiller, M.J. Fortin, C. Randin, N.E. Zimmermann, C.H. Graham and A. Guisan. 2012. Measuring ecological niche overlap from occurrence and spatial environmental data. *Global Ecology and Biogeography*, **21**, 481-497.

Warren, D.L., R.E. Glor and M. Turelli. 2008. Environmental niche equivalency versus conservatism: quantitative approaches to niche evolution. *Evolution*, **62**, 2868-2883.

**See Also**

[ecospat.grid.clim.dyn](#), [ecospat.niche.equivalency.test](#)

---

ecospat.npred	<i>Number Of Predictors</i>
---------------	-----------------------------

---

**Description**

Calculate the maximum number of predictors to include in the model with a desired correlation between predictors.

**Usage**

```
ecospat.npred (x, th)
```

**Arguments**

x	Correlation matrix of the predictors.
th	Desired threshold of correlation between predictors.

**Value**

Returns the number of predictors to use.

**Author(s)**

Blaise Petitpierre <bpetitpierre@gmail.com>

**Examples**

```
colvar <- ecospat.testData[c(4:8)]
x <- cor(colvar, method="pearson")
ecospat.npred (x, th=0.75)
```

---

ecospat.occ.desaggregation

*Species Occurrences Desaggregation*

---

**Description**

Remove species occurrences in a dataframe that are closer to each other than a specified distance threshold.

**Usage**

```
ecospat.occ.desaggregation (dfvar, colxy, colvar=NULL, min.dist, plot=T)
```

**Arguments**

dfvar	A dataframe with x, y, and variables.
colxy	The range of columns for x and y in df.
colvar	The range of columns for variables in df.
min.dist	The minimum distance threshold in the sub-dataframe.
plot	A boolean indicating whether to plot the resulting distribution of occurrences.

**Details**

The number of occurrences kept and excluded is shown in a dynamic plot. At the end of the selection process, a plot showing the resulting distribution of occurrences will be plotted if argument plot=F

**Value**

A subset of df with the columns specified in colvar.

**Author(s)**

Olivier Broennimann <olivier.broennimann@unil.ch>

**Examples**

```
## Not run:
spp<-ecospat.testNiche
sp1<- spp[1:32,1:3]

occ.sp1<-ecospat.occ.desaggregation(dfvar=sp1,colxy=2:3,colvar=NULL, min.dist=500,plot=TRUE)

## End(Not run)
```



---

ecospat.permut.glm	<i>GLM Permutation Function</i>
--------------------	---------------------------------

---

## Description

A permutation function to get p-values on GLM coefficients and deviance.

## Usage

```
ecospat.permut.glm (glm.obj, nperm)
```

## Arguments

glm.obj	Any calibrated GLM or GAM object with a binomial error distribution.
nperm	The number of permutations in the randomization process.

## Details

Rows of the response variable are permuted and a new GLM is calibrated as well as deviance, adjusted deviance and coefficients are calculated. These random parameters are compared to the true parameters in order to derive p-values.

## Value

Return p-values that are how the true parameters of the original model deviate from the distribution of the random parameters. A p-value of zero means that the true parameter is completely outside the random distribution.

## Author(s)

Christophe Randin <christophe.randin@unibas.ch>, Antoine Guisan <antoine.guisan@unil.ch> and Trevor Hastie

## References

Hastie, T., R. Tibshirani and J. Friedman. 2001. *Elements of Statistical Learning: Data Mining, Inference, and Prediction*, Springer-Verlag, New York.

Legendre, P. and L. Legendre. 1998. *Numerical Ecology*, 2nd English edition. Elsevier Science BV, Amsterdam.

## Examples

```
## Not run:  
ecospat.permut.glm (get ("glm.Achillea_atrata", envir=ecospat.env), 1000)  
  
## End(Not run)
```

---

## ecospat.plot.contrib *Plot Variables Contribution*

---

### Description

Plot the contribution of the initial variables to the analysis (i.e. correlation circle). Typically these are the eigen vectors and eigen values in ordinations.

### Usage

```
ecospat.plot.contrib (contrib, eigen)
```

### Arguments

contrib	A dataframe of the contribution of each original variable on each axis of the analysis, i.e. the eigen vectors.
eigen	A vector of the importance of the axes in the ordination, i.e. a vector of eigen values.

### Details

Requires ade4 library. If using [princomp](#), use \$loadings and \$sdev of the princomp object. if using [dudi.pca](#), use \$li and \$eig of the dudi.pca object.

### Author(s)

Olivier Broennimann <[olivier.broennimann@unil.ch](mailto:olivier.broennimann@unil.ch)>

### References

Broennimann, O., M.C. Fitzpatrick, P.B. Pearman, B. Petitpierre, L. Pellissier, N.G. Yoccoz, W. Thuiller, M.J. Fortin, C. Randin, N.E. Zimmermann, C.H. Graham and A. Guisan. 2012. Measuring ecological niche overlap from occurrence and spatial environmental data. *Global Ecology and Biogeography*, **21**, 481-497.

### See Also

[ecospat.plot.niche.dyn](#), [ecospat.plot.overlap.test](#), [ecospat.niche.similarity.test](#), [princomp](#)

---

## ecospat.plot.mess *Plot MESS*

---

### Description

Plot the MESS extrapolation index onto the geographical space.

### Usage

```
ecospat.plot.mess (xy, mess.object, cex=1, pch=15)
```

**Arguments**

xy	The x and y coordinates of the projection dataset.
mess.object	A dataframe as returned by the <code>ecospat.mess</code> function.
cex	Specify the size of the symbol.
pch	Specify the point symbols.

**Value**

Returns a plot of the the MESS extrapolation index onto the geographical space.

**Author(s)**

Blaise Petitpierre <bpetitpierre@gmail.com>

**References**

Elith, J., M. Kearney and S. Phillips. 2010. The art of modelling range-shifting species. *Methods in ecology and evolution*, **1**, 330-342.

**See Also**

[ecospat.mess](#)

**Examples**

```
## Not run:
x <- ecospat.testData[c(2,3,4:8)]
proj<- x[1:90,] #A projection dataset.
cal<- x[91:300,] #A calibration dataset

#Create a MESS object
mess.object<-ecospat.mess (proj, cal, w="default")

#Plot MESS
ecospat.plot.mess (xy=proj[c(1:2)], mess.object, cex=1, pch=15)

## End(Not run)
```

---

ecospat.plot.niche      *Plot Niche*

---

**Description**

Plot a niche z created by `ecospat.grid.clim.dyn`.

**Usage**

```
ecospat.plot.niche (z, title, name.axis1, name.axis2, cor=F)
```

**Arguments**

<code>z</code>	A gridclim object for the species distribution created by <code>ecospat.grid.clim.dyn</code> .
<code>title</code>	A title for the plot.
<code>name.axis1</code>	A label for the first axis.
<code>name.axis2</code>	A label for the second axis.
<code>cor</code>	Correct the occurrence densities of the species by the prevalence of the environments in its range (T = yes, F = no).

**Details**

if `z` is bivariate, a bivariate plot of the niche of the species. if `z` is univariate, a histogram of the niche of the species

**Author(s)**

Olivier Broennimann <olivier.broennimann@unil.ch>

**References**

Broennimann, O., M.C. Fitzpatrick, P.B. Pearman, B. Petitpierre, L. Pellissier, N.G. Yoccoz, W. Thuiller, M.J. Fortin, C. Randin, N.E. Zimmermann, C.H. Graham and A. Guisan. 2012. Measuring ecological niche overlap from occurrence and spatial environmental data. *Global Ecology and Biogeography*, **21**, 481-497.

**See Also**

[ecospat.grid.clim.dyn](#)

---

ecospat.plot.niche.dyn

*Niche Categories and Species Density*

---

**Description**

Plot niche categories and species density created by `ecospat.grid.clim.dyn`.

**Usage**

```
ecospat.plot.niche.dyn (z1,z2,quant,title,interest,
colz1,colz2,colinter,colZ1,colZ2)
```

**Arguments**

<code>z1</code>	A gridclim object for the native distribution.
<code>z2</code>	A gridclim object for the invaded range.
<code>quant</code>	The quantile of the environmental density used to remove marginal climates.
<code>title</code>	The title of the plot.
<code>interest</code>	Choose which density to plot: if <code>interest=1</code> , plot native density, if <code>interest=2</code> , plot invasive density.

colz1	The color used to depict unfilling area.
colz2	The color used to depict expansion area.
colinter	The color used to depict overlap area.
colZ1	The color used to delimit the native extent.
colZ2	The color used to delimit the invaded extent.

**Author(s)**

Blaise Petitpierre <bpetitpierre@gmail.com>

---

ecospat.plot.overlap.test  
*Plot Overlap Test*

---

**Description**

Plot a histogram of observed and randomly simulated overlaps, with p-values of equivalency or similarity tests.

**Usage**

ecospat.plot.overlap.test (x, type, title)

**Arguments**

x	Object created by <code>ecospat.niche.similarity.test</code> or <code>ecospat.niche.equivalency.test</code> .
type	Must be either "D" or "I".
title	The title for the plot.

**Author(s)**

Olivier Broennimann <olivier.broennimann@unil.ch>

**References**

Broennimann, O., M.C. Fitzpatrick, P.B. Pearman, B. Petitpierre, L. Pellissier, N.G. Yoccoz, W. Thuiller, M.J. Fortin, C. Randin, N.E. Zimmermann, C.H. Graham and A. Guisan. 2012. Measuring ecological niche overlap from occurrence and spatial environmental data. *Global Ecology and Biogeography*, **21**, 481-497.

**See Also**

[ecospat.niche.similarity.test](#), [ecospat.niche.equivalency.test](#)

---

ecospat.rand.pseudoabsences

*Sample Pseudo-Absences*


---

## Description

Randomly sample pseudoabsences from an environmental dataframe covering the study area.

## Usage

```
ecospat.rand.pseudoabsences (nbabsences, glob, colxyglob, colvar="all",
presence, colxypresence, mindist)
```

## Arguments

nbabsences	The number of pseudoabsences desired.
glob	A two-column dataframe (or a vector) of the environmental values (in column) for background pixels of the whole study area (in row).
colxyglob	The range of columns for x and y in glob.
colvar	The range of columns for the environmental variables in glob. colvar="all" keeps all the variables in glob in the final dataframe. colvar=NULL keeps only x and y.
presence	A presence-absence dataframe for each species (columns) in each location or grid cell (rows).
colxypresence	The range of columns for x and y in presence.
mindist	The minimum distance from presences within which pseudoabsences should not be drawn (buffer distance around presences).

## Value

A dataframe of random absences.

## Author(s)

Olivier Broennimann <olivier.broennimann@unil.ch>

## Examples

```
glob<-ecospat.testData[2:8]
presence<-ecospat.testData[c(2:3,9)]
presence<-presence[presence[,3]==1,1:2]
ecospat.rand.pseudoabsences (nbabsences=10, glob=glob, colxyglob=1:2, colvar = "all",
presence= presence, colxypresence=1:2, mindist=20)
```

---

ecospat.sample.envar    *Sample Environmental Variables*


---

**Description**

Add environmental values to a species dataframe.

**Usage**

```
ecospat.sample.envar (dfsp, colspxy, colspkept = "xy", dfvar,
  colvarxy, colvar = "all", resolution)
```

**Arguments**

dfsp	A species dataframe with x (long), y (lat) and optional other variables.
colspxy	The range of columns for x (long) and y (lat) in dfsp.
colspkept	The columns of dfsp that should be kept in the final dataframe (default: xy).
dfvar	A dataframe object with x, y and environmental variables.
colvarxy	The range of columns for x and y in dfvar.
colvar	The range of enviromental variable columns in dfvar (default: all except xy).
resolution	The distance between x,y of species and environmental datafreme beyond which values shouldn't be added.

**Details**

The xy (lat/long) coordinates of the species occurrences are compared to those of the environment dataframe and the value of the closest pixel is added to the species dataframe. When the closest environment pixel is more distant than the given resolution, NA is added instead of the value. This function is similar to sample() in ArcGIS.

**Value**

A Dataframe with the same rows as dfsp, with environmental values from dfvar in column.

**Author(s)**

Olivier Broennimann <olivier.broennimann@unil.ch>

**Examples**

```
## Not run:
spp<-ecospat.testNiche
sp1<- spp[1:32,1:3]
occ.sp1<-ecospat.occ.desaggregation(dfvar=sp1,colxy=2:3,colvar=NULL, min.dist=500,plot=TRUE)
clim<- ecospat.testData[2:8]

occ_sp1<-na.exclude(ecospat.sample.envar(dfsp=occ.sp1,colspxy=1:2,colspkept=1:2,dfvar=clim,
colvarxy=1:2,colvar="all",resolution=25))

## End(Not run)
```

---

ecospat.SSDMeval      *SSDM evaluation*

---

## Description

Calculate several indices of accuracy of community predictions.

## Usage

```
ecospat.SSDMeval (eval, pred, proba, ntir)
```

## Arguments

eval	A matrix of observed presence-absence (ideally independent from the dataset used to fit species distribution models) of the species with n rows for the sites and s columns for the species.
pred	A matrix of predictions for the s species in the n sites. Should have the same dimension as eval.
proba	Logical variable indicating whether the prediction matrix contains presences-absences (FALSE) or probabilities (TRUE).
ntir	Number of trials of presence-absence predictions if pred is a probability matrix.

## Details

This function calculates several indices of accuracy of community predictions based on stacked predictions of species ditribution models. In case proba is set to FALSE the function returns one value per index and per site. In case proba is set to TRUE the function generates presences-absences based on the predicted probabilities and returns one value per index, per site and per trial.

## Value

A list of evaluation metrics calculated for each site (+ each trial if proba is set to TRUE):

deviance.rich.pred: the deviation of the predicted species richness to the observed

overprediction: the proportion of species predicted as present but not observed among the species predicted as present

underprediction: the proportion of species predicted as absent but observed among the species observed as present

prediction.success: the proportion of species correctly predicted as present or absent

sensitivity: the proportion of species correctly predicted as present among the species observed as present

specificity : the proportion of species correctly predicted as absent among the species observed as absent

kappa: the proportion of specific agreement

TSS: sensitivity+specificity-1

similarity: the similarity of community composition between the observation and the prediction. The calculation is based on the Sorenses index.

Jaccard: this index is a widely used metric of community similarity.



**Author(s)**

Julien Pottier <julien.pottier@clermont.inra.fr> with contribution of Daniel Scherrer <daniel.scherrer@unil.ch>  
 Anne Dubuis <anne.dubuis@gmail.com> and Manuela D'Amen <manuela.damen@unil.ch>

**References**

Pottier, J., A. Dubuis, L. Pellissier, L. Maiorano, L. Rossier, C.F. Randin, P. Vittoz and A. Guisan.  
 2013. The accuracy of plant assemblage prediction from species distribution models varies along  
 environmental gradients. *Global Ecology and Biogeography*, **22**, 52-63.

**Examples**

```
## Not run:
eval<-Data[c(53,62,58,70,61,66,65,71,69,43,63,56,68,57,55,60,54,67,59,64)]
pred<-Data[c(73:92)]

ecospat.SSDMeval (eval, pred, proba=TRUE, ntir=10)

## End(Not run)
```

---

ecospat.testData	<i>Test data for the ecospat library</i>
------------------	------------------------------------------

---

**Description**

Data frame that contains vegetation plots data: presence records of 50 species, a set of environmental variables (topo-climatic) and SDM predictions for some species in the Western Swiss Alps (Canton de Vaud, Switzerland).

**Usage**

```
data(ecospat.testData)
```

**Format**

A data frame with 300 observations on the following 96 variables.

numplots Number of the vegetation plot.

long Longitude, in Swiss plane coordinate system of the vegetation plot.

lat Latitude, in Swiss plane coordinate system of the vegetation plot.

ddeg Growing degree days (with a 0 degrees Celsius threshold).

mind Moisture index over the growing season (average values for June to August in mm day<sup>-1</sup>).

srad The annual sum of radiation (in kJ m<sup>-2</sup> year<sup>-1</sup>).

slp Slope (in degrees) calculated from the DEM25.

topo Topographic position (an integrated and unitless measure of topographic exposure).

Achillea\_atrata

Achillea\_millefolium

Acinos\_alpinus

Adenostyles\_glabra

Aposeris\_foetida  
Arnica\_montana  
Aster\_bellidiastrum  
Bartsia\_alpina  
Bellis\_perennis  
Campanula\_rotundifolia  
Centaurea\_montana  
Cerastium\_latifolium  
Cruciata\_laevipes  
Doronicum\_grandiflorum  
Galium\_album  
Galium\_anisophyllum  
Galium\_megalospermum  
Gentiana\_bavarica  
Gentiana\_lutea  
Gentiana\_purpurea  
Gentiana\_verna  
Globularia\_cordifolia  
Globularia\_nudicaulis  
Gypsophila\_repens  
Hieracium\_lactucella  
Homogyne\_alpina  
Hypochaeris\_radicata  
Leontodon\_autumnalis  
Leontodon\_helveticus  
Myosotis\_alpestris  
Myosotis\_arvensis  
Phyteuma\_orbiculare  
Phyteuma\_spicatum  
Plantago\_alpina  
Plantago\_lanceolata  
Polygonum\_bistorta  
Polygonum\_viviparum  
Prunella\_grandiflora  
Rhinanthus\_alectorolophus  
Rumex\_acetosa  
Rumex\_crispus  
Vaccinium\_gaultherioides  
Veronica\_alpina  
Veronica\_aphylla

Agrostis\_capillaris  
 Bromus\_erectus\_sstr  
 Campanula\_scheuchzeri  
 Carex sempervirens  
 Cynosurus\_cristatus  
 Dactylis\_glomerata  
 Daucus\_carota  
 Festuca\_pratensis\_sl  
 Geranium\_sylvaticum  
 Leontodon\_hispidus\_sl  
 Potentilla\_erecta  
 Pritzelago\_alpina\_sstr  
 Prunella\_vulgaris  
 Ranunculus\_acris\_sl  
 Saxifraga\_oppositifolia  
 Soldanella\_alpina  
 Taraxacum\_officinale\_aggr  
 Trifolium\_repens\_sstr  
 Veronica\_chamaedrys  
 Parnassia\_palustris  
 glm\_Agrostis\_capillaris GLM model for the species Agrostis\_capillaris.  
 glm\_Leontodon\_hispidus\_sl GLM model for the species Leontodon\_hispidus\_sl.  
 glm\_Dactylis\_glomerata GLM model for the species Dactylis\_glomerata.  
 glm\_Trifolium\_repens\_sstr GLM model for the species Trifolium\_repens\_sstr.  
 glm\_Geranium\_sylvaticum GLM model for the species Geranium\_sylvaticum.  
 glm\_Ranunculus\_acris\_sl GLM model for the species Ranunculus\_acris\_sl.  
 glm\_Prunella\_vulgaris GLM model for the species Prunella\_vulgaris.  
 glm\_Veronica\_chamaedrys GLM model for the species Veronica\_chamaedrys.  
 glm\_Taraxacum\_officinale\_aggr GLM model for the species Taraxacum\_officinale\_aggr.  
 glm\_Plantago\_lanceolata GLM model for the species Plantago\_lanceolata.  
 glm\_Potentilla\_erecta GLM model for the species Potentilla\_erecta.  
 glm\_Carex sempervirens GLM model for the species Carex sempervirens.  
 glm\_Soldanella\_alpina GLM model for the species Soldanella\_alpina.  
 glm\_Cynosurus\_cristatus GLM model for the species Cynosurus\_cristatus.  
 glm\_Campanula\_scheuchzeri GLM model for the species Campanula\_scheuchzeri.  
 glm\_Festuca\_pratensis\_sl GLM model for the species Festuca\_pratensis\_sl.  
 gbm\_Bromus\_erectus\_sstr GBM model for the species Bromus\_erectus\_sstr.  
 glm\_Saxifraga\_oppositifolia GLM model for the species Saxifraga\_oppositifolia.  
 glm\_Daucus\_carota GLM model for the species Daucus\_carota.  
 glm\_Pritzelago\_alpina\_sstr GLM model for the species Pritzelago\_alpina\_sstr.  
 glm\_Bromus\_erectus\_sstr GLM model for the species Bromus\_erectus\_sstr.  
 gbm\_Saxifraga\_oppositifolia GBM model for the species Saxifraga\_oppositifolia.  
 gbm\_Daucus\_carota GBM model for the species Daucus\_carota.  
 gbm\_Pritzelago\_alpina\_sstr GBM model for the species Pritzelago\_alpina\_sstr.

## Details

The study area is the Western Swiss Alps of Canton de Vaud, Switzerland.

Five topo-climatic explanatory variables to calibrate the SDMs: growing degree days (with a 0 degrees Celsius threshold); moisture index over the growing season (average values for June to August in mm day<sup>-1</sup>); slope (in degrees); topographic position (an integrated and unitless measure of topographic exposure; Zimmermann et al., 2007); and the annual sum of radiation (in kJ m<sup>-2</sup> year<sup>-1</sup>). The spatial resolution of the predictor is 25 m x 25 m so that the models could capture most of the small-scale variations of the climatic factors in the mountainous areas.

Two modelling techniques were used to produce the SDMs: generalized linear models (GLM; McCullagh & Nelder, 1989; R library 'glm') and generalized boosted models (GBM; Friedman, 2001; R library 'gbm'). The SDMs correspond to 20 species: *Agrostis\_capillaris*, *Leontodon\_hispidus\_sl*, *Dactylis\_glomerata*, *Trifolium\_repens\_sstr*, *Geranium\_sylvaticum*, *Ranunculus\_acris\_sl*, *Prunella\_vulgaris*, *Veronica\_chamaedrys*, *Taraxacum\_officinale\_aggr*, *Plantago\_lanceolata*, *Potentilla\_erecta*, *Carex sempervirens*, *Soldanella\_alpina*, *Cynosurus\_cristatus*, *Campanula\_scheuchzeri*, *Festuca\_pratensis\_sl*, *Daucus\_carota*, *Pritzelago\_alpina\_sstr*, *Bromus\_erectus\_sstr* and *Saxifraga\_oppositifolia*.

## Author(s)

Antoine Guisan <antoine.guisan@unil.ch>, Anne Dubuis <anne.dubuis@gmail.com> and Valeria Di Cola <valeria.dicola@unil.ch>

## References

- Guisan, A. 1997. Distribution de taxons végétaux dans un environnement alpin: Application de modélisations statistiques dans un système d'information géographique. PhD Thesis, University of Geneva, Switzerland.
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- Guisan, A. & J.P. Theurillat. 2000. Assessing alpine plant vulnerability to climate change: A modeling perspective. *Integrated Assessment*, **1**, 307-320.
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## Examples

```
data(ecospat.testData)
dim(ecospat.testData)
names(ecospat.testData)
```

---

ecospat.testNiche	<i>Test data for the niche overlap analysis</i>
-------------------	-------------------------------------------------

---

## Description

Data frame that contains occurrence sites for each species, long, lat and the name of the species at each site.

## Usage

```
data(ecospat.testNiche)
```

## Format

ecospat.testNiche is a data frame with the following columns:

species sp1, sp2, sp3 and sp4.

long Longitude, in Swiss plane coordinate system of the vegetation plot.

lat Latitude, in Swiss plane coordinate system of the vegetation plot.

Spp Scientific name of the species used in the exmaple: Bromus\_erectus\_sstr, Saxifraga\_oppositifolia, Daucus\_carota and Pritzelago\_alpina\_sstr.

## Details

List of occurence sites for the species.

## Author(s)

Antoine Guisan <antoine.guisan@unil.ch>, Anne Dubuis <anne.dubuis@gmail.com> and Valeria Di Cola <valeria.dicola@unil.ch>

## See Also

[ecospat.testData](#)

## Examples

```
data(ecospat.testNiche)
dim(ecospat.testNiche)
names(ecospat.testNiche)
```

---

ecospat.testTree	<i>Test tree for the ecospat library</i>
------------------	------------------------------------------

---

## Description

The tree object is a phylogenetic tree of class 'phylo' (see read.tree) that contains data of 50 angiosperm species from the Western Swiss Alps.

## Format

ecospat.testTree is a tree contains the following species:

[1] "Rumex\_acetosa" [2] "Polygonum\_bistorta" [3] "Polygonum\_viviparum" [4] "Rumex\_crispus"  
 [5] "Cerastium\_latifolium" [6] "Silene\_aucaulis" [7] "Gypsophila\_repens" [8] "Vaccinium\_gaultherioides"  
 [9] "Soldanella\_alpina" [10] "Cruciata\_laevipes" [11] "Galium\_album" [12] "Galium\_anisophyllum"  
 [13] "Galium\_megalospermum" [14] "Gentiana\_verna" [15] "Gentiana\_bavarica" [16] "Gentiana\_purpurea"  
 [17] "Gentiana\_lutea" [18] "Bartsia\_alpina" [19] "Rhinanthus\_alectorolophus" [20] "Prunella\_grandiflora"  
 [21] "Acinos\_alpinus" [22] "Plantago\_alpina" [23] "Plantago\_lanceolata" [24] "Veronica\_officinalis"  
 [25] "Veronica\_aphylla" [26] "Veronica\_alpina" [27] "Veronica\_chamaedrys" [28] "Veronica\_persica"  
 [29] "Globularia\_cordifolia" [30] "Globularia\_nudicaulis" [31] "Myosotis\_alpestris" [32] "Myosotis\_arvensis" [33] "Aposeris\_foetida" [34] "Centaurea\_montana" [35] "Hieracium\_lactucella" [36]  
 "Leontodon\_helveticus" [37] "Leontodon\_autumnalis" [38] "Hypochaeris\_radicata" [39] "Achillea\_atrata" [40] "Achillea\_millefolium" [41] "Homogyne\_alpina" [42] "Senecio\_doronicum" [43]  
 "Adenostyles\_glabra" [44] "Arnica\_montana" [45] "Aster\_bellidiflorus" [46] "Bellis\_perennis"  
 [47] "Doronicum\_grandiflorum" [48] "Phyteuma\_orbiculare" [49] "Phyteuma\_spicatum" [50] "Campanula\_rotundifolia"

## Author(s)

Charlotte Ndiribe <charlotte.ndiribe@unil.ch>, Nicolas Salamin <nicolas.salamin@unil.ch> and Antoine Guisan <antoine.guisan@unil.ch>

## References

Ndiribe, C., L. Pellissier, S. Antonelli, A. Dubuis, J. Pottier, P. Vittoz, A. Guisan and N. Salamin. 2013. Phylogenetic plant community structure along elevation is lineage specific. *Ecology and Evolution*, **3**, 4925-4939.

## Examples

```
fpath <- system.file("extdata", "ecospat.testTree.tre", package="ecospat")
tree<-read.tree(fpath)
plot(tree)
```

---

ecospat.varpart	<i>Variation Partitioning for GLM or GAM</i>
-----------------	----------------------------------------------

---

**Description**

Perform variance partitioning for binomial GLM or GAM based on the deviance of two groups or predicting variables.

**Usage**

```
ecospat.varpart (model.1, model.2, model.12)
```

**Arguments**

model.1	GLM / GAM calibrated on the first group of variables.
model.2	GLM / GAM calibrated on the second group of variables.
model.12	GLM / GAM calibrated on all variables from the two groups.

**Details**

The deviance is calculated with the adjusted geometric mean squared improvement rescaled for a maximum of 1.

**Value**

Return the four fractions of deviance as in Randin et al. 2009: partial deviance of model 1 and 2, joined deviance and unexplained deviance.

**Author(s)**

Christophe Randin <christophe.randin@unibas.ch>, Helene Jaccard and Nigel Gilles Yoccoz

**References**

Randin, C.F., H. Jaccard, P. Vittoz, N.G. Yoccoz and A. Guisan. 2009. Land use improves spatial predictions of mountain plant abundance but not presence-absence. *Journal of Vegetation Science*, **20**, 996-1008.

**Examples**

```
## Not run:
ecospat.cv.example()
ecospat.varpart (model.1= get ("glm.Achillea_atrata", envir=ecospat.env),
model.2= get ("glm.Achillea_millefolium", envir=ecospat.env),
model.12= get ("glm.Achillea_millefolium", envir=ecospat.env))

## End(Not run)
```

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