

Package ‘jtdm’

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Type Package

Title Joint Modelling of Functional Traits

Version 0.1-0

Description A collection of tools to fit and analyse a Joint Trait Distribution Model (JTDM). In particular, the package computes joint probabilities and multivariate confidence intervals and their partial response curves as a function of a selected environmental covariate. The tools implemented by the package are described in Poggiato, Gaüzere, Martinez-Almoyna, Deschamps, Renaud, Violle, Münkemüller and Thuiller (In preparation).

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Encoding UTF-8

LazyData true

Imports ggforce, mniw, mvtnorm, parallel, stats, utils, ggplot2

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Depends R (>= 2.10)

URL <https://github.com/giopogg/jtdm>

VignetteBuilder knitr

Suggests knitr,
rmarkdown,
devtools

R topics documented:

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ellipse_plot	<i>Partial response curve of the pairwise most suitable community-level strategy and of the pairwise envelop of possible community-level strategy</i>
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Description

Partial response curve of the pairwise most suitable community-level strategy and of the pairwise envelop of possible community-level strategy. In order to build the response curve, the function builds a dataframe where the focal variable varies along a gradient and the other (non-focal) variables are fixed to their mean (but see FixX parameter for fixing non-focal variables to user-defined values). The chosen traits are specified in indexTrait. Then uses the jtdm_predict function to compute the most suitable community-level strategy and the residual covariance matrix to build the envelop of possible CLS.

Usage

```
ellipse_plot(
  m,
  indexGradient,
  indexTrait,
  FullPost = F,
  grid.length = 20,
  FixX = NULL,
  confL = 0.95
)
```

Arguments

m	a model fitted with jtdm_fit
indexGradient	The name (as specified in the column names of X) of the focal variable.
indexTrait	A vector of the two names (as specified in the column names of Y) containing the two (or more!) traits we want to compute the community level strategy of.
FullPost	If FullPost = TRUE, the function returns samples from the predictive distribution of joint probabilities. If FullPost= FALSE, joint probabilities are computed only using the posterior mean of the parameters.
grid.length	The number of points along the gradient of the focal variable. Default to 20 (which ensures a fair visualisation).
FixX	Optional. A parameter to specify the value to which non-focal variables are fixed. This can be useful for example if we have some categorical variables (e.g. forest vs meadows) and we want to obtain the partial response curve for a given value of the variable. It has to be a list of the length and names of the columns of X. For example, if the columns of X are "MAT", "MAP", "Habitat" and we want to fix "Habitat" to 1, then FixX=list(MAT=NULL,MAP=NULL,Habitat=1.). Default to NULL.
confL	The confidence level of the confidence ellipse (i.e. of the envelop of possible community-level strategies). Default is 0.95.

Value

Plot of the partial response curve of the pairwise most suitable community-level strategy and of the pairwise envelop of possible community-level strategy

Examples

```
data(Y)
data(X)
# Short MCMC to obtain a fast example: results are unreliable !
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 1000)

# plot the pairwise SLA-LNC partial response curve along the GDD gradient
ellipse_plot(m,indexTrait = c("SLA","LNC"),indexGradient="GDD")
# plot the pairwise SLA-LNC partial response curve along the GDD gradient
# in forest (i.e. when forest=1)
ellipse_plot(m,indexTrait = c("SLA","LNC"),indexGradient="GDD",
             FixX=list(GDD=NULL,FDD=NULL,forest=1))
```

getB

*Get the inferred regression coefficients***Description**

Get the samples from the posterior distribution of the regression coefficient matrix B, together with the posterior mean and quantiles. The regression coefficient matrix B is a matrix where the number of rows is defined by the number of traits that are modelled, and the number of columns is the number of columns of the matrix m\$X (the number of explanatory variables after transformation via formula)

Usage

```
getB(m)
```

Arguments

m a model fitted with jtdm_fit

Value

A list containing:

Bsamples	Sample from the posterior distribution of the regression coefficient matrix. It is an array where the first dimension is the number of traits, the second the number of columns in m\$X (the number of variables after transformation via formula) and the third the number of MCMC samples.
Bmean	Posterior mean of the regression coefficient matrix.
Bq975, Bq025	97.5% and 0.25% posterior quantiles of the regression coefficient matrix.

Examples

```
data(Y)
data(X)
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 1000)
# get the inferred regression coefficients
B=getB(m)
```

get_sigma

Get the inferred residual covariance matrix

Description

Get the samples from the posterior distribution of the residual covariance matrix, together with the posterior mean and quantiles.

Usage

```
get_sigma(m)
```

Arguments

m a model fitted with jtdm_fit

Value

A list containing:

Ssamples	Sample from the posterior distribution of the residual covariance matrix. It is an array where the first two dimensions are the rows and columns of the matrix, and the third dimensions are the samples from the posterior distribution
Smean	Posterior mean of the residual covariance matrix.
Sq975, Sq025	97.5% and 0.25% posterior quantiles of the residual covariance matrix.

Examples

```
data(Y)
data(X)
# Short MCMC to obtain a fast example: results are unreliable !
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 1000)
# get the inferred residual covariance
Sigma =get_sigma(m)
```

global

Global

Description

Declare global variables

joint_trait_prob	<i>Computes joint probabilities.</i>
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Description

Computes the joint probability of CWM traits in regions in the community-trait space specified by bounds and in sites specified in Xnew.

Usage

```
joint_trait_prob(
  m,
  indexTrait,
  bounds,
  Xnew = NULL,
  FullPost = T,
  mcmc.samples = NULL,
  parallel = FALSE
)
```

Arguments

m	a model fitted with <code>jtdm_fit</code>
indexTrait	A vector of the names (as specified in the column names of Y) of the two (or more!) traits we want to compute the joint probabilities of.
bounds	The parameter to specify a region in the community-trait space where the function computes the joint probabilities of traits. It is a list of the length of "index-Trait", each element of the list is a vector of length two. The vector represents the inferior and superior bounds of the region for the specified trait. For example, if we consider two traits, <code>bounds=list(c(10,Inf),c(10,Inf))</code> corresponds to the region in the community-trait space where both traits both take values greater than 10.
Xnew	Optionally, a data frame in which to look for variables with which to predict. If omitted, the fitted linear predictors are used.
FullPost	If <code>FullPost = TRUE</code> , the function returns samples from the predictive distribution of joint probabilities. If <code>FullPost= FALSE</code> , joint probabilities are computed only using the posterior mean of the parameters. <code>FullPost</code> cannot be equal to "mean" here.
mcmc.samples	Optional, default to <code>NULL</code> , only works when <code>FullPost=FALSE</code> . Defines the number of MCMC samples to compute the posterior distribution of joint probabilities. Needs to be between 1 and <code>m\$model\$sample x length(m\$model\$mcmc)</code>
parallel	Optional, only works when <code>FullPost = TRUE</code> . When <code>TRUE</code> , the function uses <code>mclapply</code> to parallelise the calculation of the posterior distribution joint probabilities.

Details

This function is time consuming when `FullPost=T`. Consider setting `parallel=T` and/or to set `mcmc.samples` to a value smaller than the length of the MCMC chains.

Value

A list containing:

PROBsamples	Samples from the posterior distribution of the joint probability.NULL if Full-Post=FALSE.
PROBmean	Posterior mean of the joint probability.
PROBq975,PROBq025	97.5% and 0.25% posterior quantiles of the joint probability. NULL if Full-Post=FALSE.

Examples

```
data(Y)
data(X)
#We sample only few samples from the posterior in order to reduce
# the computational time of the examples.
#Increase the number of samples to obtain robust results
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 10)
# Compute probability of SLA and LNC to be joint-high at sites in the studies
joint = joint_trait_prob(m,indexTrait=c("SLA","LNC"),
                        bounds=list(c(mean(Y[, "SLA"]),Inf),c(mean(Y[, "SLA"]),Inf)))
```

joint_trait_prob_gradient

Computes partial response curves of joint probabilities

Description

Computes the partial responses curves of joint probability of CWM traits as a function of a focal variable. The regions in which joint probabilities are computed are specified by bounds. In order to build the response curve, the function builds a dataframe where the focal variable varies along a gradient and the other (non-focal) variables are fixed to their mean (but see FixX parameter for fixing non-focal variables to user-defined values). Then, uses joint_trait_prob to compute the joint probability in these dataset.

Usage

```
joint_trait_prob_gradient(
  m,
  indexTrait,
  indexGradient,
  bounds,
  grid.length = 200,
  XFocal = NULL,
  FixX = NULL,
  FullPost = T,
  mcmc.samples = NULL,
  parallel = FALSE
)
```

Arguments

<code>m</code>	A model fitted with <code>jtdm_fit</code>
<code>indexTrait</code>	A vector of the names (as specified in the column names of Y) of the two (or more!) traits we want to compute the joint probabilities of.
<code>indexGradient</code>	The name (as specified in the column names of X) of the focal variable.
<code>bounds</code>	The parameter to specify a region in the community-trait space where the function computes the joint probabilities of traits. It is a list of the length of "indexTrait", each element of the list is a vector of length two. The vector represents the inferior and superior bounds of the region for the specified trait. For example, if we consider two traits, <code>bounds=list(c(10,Inf),c(10,Inf))</code> corresponds to the region in the community-trait space where both traits both take values greater than 10.
<code>grid.length</code>	The number of points along the gradient of the focal variable. Default to 200.
<code>XFocal</code>	Optional. A gradient of the focal variable provided by the user. If provided, the function will use this gradient instead of building a regular one. Default to NULL.
<code>FixX</code>	Optional. A parameter to specify the value to which non-focal variables are fixed. This can be useful for example if we have some categorical variables (e.g. forest vs meadows) and we want to obtain the partial response curve for a given value of the variable. It has to be a list of the length and names of the columns of X. For example, if the columns of X are "MAT", "MAP", "Habitat" and we want to fix "Habitat" to 1, then <code>FixX=list(MAT=NULL,MAP=NULL,Habitat=1.)</code> . Default to NULL.
<code>FullPost</code>	If <code>FullPost = TRUE</code> , the function returns samples from the predictive distribution of joint probabilities. If <code>FullPost= FALSE</code> , joint probabilities are computed only using the posterior mean of the parameters. <code>FullPost</code> cannot be equal to "mean" here.
<code>mcmc.samples</code>	Optional, default to NULL, only works when <code>FullPost=FALSE</code> . Defines the number of MCMC samples to compute the posterior distribution of joint probabilities. Needs to be between 1 and <code>m\$model\$sample x length(m\$model\$mcmc)</code>
<code>parallel</code>	Optional, only works when <code>FullPost = TRUE</code> . When <code>TRUE</code> , the function uses <code>mclapply</code> to parallelise the calculation of the posterior distribution joint probabilities.

Details

This function is time consuming when `FullPost=T`. Consider setting `parallel=T` and/or to set `mcmc.samples` to a value smaller than the length of the MCMC chains.

Value

A list containing:

<code>GradProbssamples</code>	Sample from the posterior distribution of the joint probability along the gradient. It is a vector whose length is the number of MCMC samples. NULL if <code>FullPost=FALSE</code> .
<code>GradProbsmean</code>	Posterior mean of the joint probability along the gradient.

GradProbsq975,GradProbsq025
 97.5% and 0.25% posterior quantiles of the joint probability along the gradient.
 NULL if FullPost=FALSE.

gradient The gradient of the focal variable built by the function.

Examples

```
data(Y)
data(X)
#We sample only few samples from the posterior in order to reduce
#the computational time of the examples.
#Increase the number of samples to obtain robust results
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 10)
# Compute probability of SLA and LNC to be joint-high at sites in the studies

# Compute the joint probability of SLA and LNC
#to be joint-high along the GDD gradient
joint = joint_trait_prob_gradient(m,indexTrait=c("SLA","LNC"),
                                indexGradient="GDD",
                                bounds=list(c(mean(Y[, "SLA"]),Inf),c(mean(Y[, "SLA"]),Inf)))
# Compute the joint probability of SLA and LNC to be joint-high along the
# GDD gradient when forest = 1 (i.e. in forests)
joint = joint_trait_prob_gradient(m,indexTrait=c("SLA","LNC"),
                                indexGradient="GDD",
                                bounds=list(c(mean(Y[, "SLA"]),Inf),c(mean(Y[, "SLA"]),Inf)),
                                FixX=list(GDD=NULL,FDD=NULL,forest=1))
```

jtdm

jtdm.

Description

Package to fit a Join Trait Distribution Model and to analyse its result to understand and predict the community-level strategy. See Poggiato et al. In prep.

jtdmCV

K-fold cross validation predictions and goodness of fit metrics

Description

Run K-fold cross validation predictions of the model m on a specified dataset.

Usage

```
jtdmCV(m, K = 5, sample = 1000, partition = NULL)
```


Arguments

m	a model fitted with jtdm_fit
K	The number of folds of the K-fold cross validation
sample	Number of samples from the posterior distribution. Since we sample from the exact posterior distribution, the number of samples is relative lower than MCMC samplers. As a rule of thumb, 1000 samples should provide correct inference.
partition	A partition of the dataset specified by the user. It is a vector (whose length are the number of sites), where each element specifies the fold index of the site.

Value

A list containing:

Pred	Sample from the posterior predictive distribution in cross validation. It is an array where the first dimension is the number of sites in Xnew, the second is the number of traits modelled and the third the number of MCMC samples. NULL if FullPost=FALSE.
PredMean	Posterior mean of posterior predictive distribution in cross validation.
Predq975,Predq025	97.5% and 0.25% posterior quantiles of the posterior predictive distribution in cross validation. NULL if FullPost=FALSE.
R2	R squared of predictions in cross validation.
RMSE	Root square mean error between squared of predictions in cross validation.

Examples

```
data(Y)
data(X)
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 1000)
# Run 3-fold cross validation on m
pred = jtdmCV(m, K=5, sample = 1000)
```

jtdm_fit

Fitting joint trait distribution models

Description

jtdm_fit is used to fit a Joint trait distribution model. Requires the response variable Y (the sites x traits matrix) and the explanatory variables X. This function samples from the posterior distribution of the parameters, which has been analytically determined. Therefore, there is no need for classical MCMC convergence checks.

Usage

```
jtdm_fit(Y, X, formula, sample = 5000)
```

Arguments

Y	The sites x traits matrix containing community (weighted) means of each trait at each site.
X	The design matrix, i.e. sites x predictor matrix containing the value of each explanatory variable (e.g. the environmental conditions) at each site.
formula	An object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted. The details of model specification are given under 'Details'
sample	Number of samples from the posterior distribution. Since we sample from the exact posterior distribution, the number of samples is relative lower than MCMC samplers. As a rule of thumb, 1000 samples should provide correct inference.

Details

A formula has an implied intercept term. To remove this use either $y \sim x - 1$ or $y \sim 0 + x$. See formula for more details of allowed formulae.

Value

A list containing:

model	An object of class "jtdm_fit", containing the samples from the posterior distribution of the regression coefficients (B) and residual covariance matrix (Sigma), together with the likelihood of the model.
Y	A numeric vector of standard errors on parameters
X_raw	The design matrix specified as input
X	The design matrix transformed as specified in formula
formula	The formula specified as input

Examples

```
data(Y)
data(X)
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 1000)
```

jtdm_predict

Predict method for joint trait distribution model

Description

Obtains predictions from a fitted joint trait distribution model and optionally computes their R squared and root mean square error (RMSE)

Usage

```
jtdm_predict(m = m, Xnew = NULL, Ynew = NULL, validation = F, FullPost = T)
```

Arguments

<code>m</code>	a model fitted with <code>jtdm_fit</code>
<code>Xnew</code>	optionally, a data frame in which to look for variables with which to predict. If omitted, the fitted linear predictors are used
<code>Ynew</code>	Optional. The observed response variables at sites specified in <code>Xnew</code> . It is used to compute goodness of fit metrics when <code>validation=T</code> .
<code>validation</code>	boolean parameter to decide whether we want to compute goodness of fit measures. If true, then <code>Ynew</code> is needed.
<code>FullPost</code>	The type of predictions to be obtain. If <code>FullPost = TRUE</code> , the function returns samples from the predictive distribution. If <code>FullPost="mean"</code> , the function computes the posterior distribution of the regression term BX_{new} . If <code>FullPost=F</code> , the function only returns the posterior mean of the regression term ($BmeanX_{new}$).

Details

To obtain a full assesment of the posterior distribution, the function should be ran with `FullPost=TRUE`, although this can be time consuming. `FullPost="mean"` is used to compute partial response curves, while `FullPost=FALSE` is used to compute goodness of fit metrics.

Value

A list containing:

<code>Pred</code>	Sample from the posterior distribution of the posterior predictive distribution. It is an array where the first dimension is the number of sites in <code>Xnew</code> , the second is the number of traits modelled and the third the number of MCMC samples. NULL if <code>FullPost=FALSE</code> .
<code>PredMean</code>	Posterior mean of posterior predictive distribution
<code>Predq975,Predq025</code>	97.5% and 0.25% posterior quantiles of the posterior predictive distribution. NULL if <code>FullPost=FALSE</code> .
<code>R2</code>	R squared of predictions (squared Pearson correlation between <code>Ynew</code> and the predictions). NULL if <code>validation=FALSE</code> .
<code>RMSE</code>	Root square mean error between squared of predictions. NULL if <code>validation=FALSE</code> .

Examples

```
data(Y)
data(X)
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 1000)
# marginal predictions of traits in the sites of X
pred = jtdm_predict(m)
```

partial_response	<i>Computes and plots the trait-environment relationship of a given CWM trait and a given environmental variable</i>
------------------	--

Description

Computes and plots the trait-environment relationship of a given CWM trait and a focal environmental variable. In order to build the response curve, the function builds a dataframe where the focal environmental variable varies along a gradient and the other (non-focal) variables are fixed to their mean (but see FixX parameter for fixing non-focal variables to user-defined values).

Usage

```
partial_response(
  m,
  indexGradient,
  indexTrait,
  XFocal = NULL,
  grid.length = 200,
  FixX = NULL,
  FullPost = "mean"
)
```

Arguments

m	a model fitted with <code>jtdm_fit</code>
indexGradient	The name (as specified in the column names of X) of the focal variable.
indexTrait	The name (as specified in the column names of Y) of the focal trait.
XFocal	Optional. A gradient of the focal variable provided by the user. If provided, the function will use this gradient instead of building a regular one. Default to NULL.
grid.length	The number of points along the gradient of the focal variable. Default to 200.
FixX	Optional. A parameter to specify the value to which non-focal variables are fixed. This can be useful for example if we have some categorical variables (e.g. forest vs meadows) and we want to obtain the partial response curve for a given value of the variable. It has to be a list of the length and names of the columns of X. For example, if the columns of X are "MAT", "MAP", "Habitat" and we want to fix "Habitat" to 1, then <code>FixX=list(MAT=NULL,MAP=NULL,Habitat=1.)</code> . Default to NULL.
FullPost	The type of predictions to be obtained. If <code>FullPost = TRUE</code> , the function returns samples from the predictive distribution. If <code>FullPost="mean"</code> , the function computes the posterior distribution of the regression term $B\%*\%X$). Default to "mean", here <code>FullPost</code> cannot be FALSE.

Value

A list containing:

p	A plot of the trait-environment relationship.
---	---

predictions A data frame containing the predicted trait-environmental relationships including the gradient of the focal environmental variable, mean trait predictions and quantiles (can be useful to code customized plot).

Examples

```
data(Y)
data(X)
# Short MCMC to obtain a fast example: results are unreliable !
m = jtdm_fit(Y=Y, X=X, formula=as.formula("~GDD+FDD+forest"), sample = 1000)
# SLA-GDD relationship
plot = partial_response(m, indexGradient="GDD", indexTrait="SLA")
plot$p
# SLA-GDD relationship in forest (i.e. when forest=1)
plot = partial_response(m, indexGradient="GDD", indexTrait="SLA",
                        FixX=list(GDD=NULL, FDD=NULL, forest=1))
plot$p
```

X	<i>Site x environmental covariates dataset</i>
---	--

Description

Includes the Growing Degree Days (GDD) during the growing season and Freezing Degree Days (FDD) during the growing season averaged over the period 1989-2019

Usage

```
data(X)
```

```
data(X)
```

Format

A matrix

Author(s)

Orchamp consortium

Examples

```
data(X)
```

Y	<i>Site x CWM traits dataset</i>
---	----------------------------------

Description

A site x CWM traits dataset computed using pinpoint abundances of plants and species mean

Usage

data(Y)

Format

A matrix

Author(s)

Orchamp Consortium

Examples

data(Y)

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