Package: mglmn (via r-universe)

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Type Package

Title Model Averaging for Multivariate GLMM with Null Models

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Description Tools for univariate and multivariate generalized linear
mixed models with model averaging and null model technique.

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URL https://github.com/mattocci27/mglmn

BugReports https://github.com/mattocci27/mglmn/issues

Depends R (>= 3.5)

Imports mvabund, gllvm, snowfall

RoxygenNote 7.1.1

Suggests testthat

Repository https://mattocci27.r-universe.dev

RemoteUrl https://github.com/mattocci27/mglmn

RemoteRef HEAD

RemoteSha 2eaf43cd917fae184655144544b510035eb35bed

Contents

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best.vars	*Best variables*

Description

Returns variables for the best model based on AIC

Usage

```
best.vars(x)
```

Arguments

x   A list of results of 'maglm' and 'mamglmg'

Value

A vector of terms of the best model.

See Also

[maglm](#), [mamglm](#)

Examples

```
#load species composition and environmental data
data(capcay)
adj.sr <- capcay$adj.sr
env_sp <- capcay$env_sp

#to fit a poisson regression model:
res <- maglm(data = env_sp, y = "adj.sr", family = "gaussian")

best.vars(res)
```

---

capcay	*Capcay data*

Description

Species composition and environmental data from Capricornia Cays

Usage

```
data(capcay)
```

Format

A list containing the elements

abund A data frame with 14 observations of abundance of 13 ant species

adj.sr A vector of adjusted species richness of ants based on sample-based rarefaction curves to standardise sampling intensity across sites (see Nakamura et al. 2015 for more details).

env_sp A data frame of 10 environmental variables, which best explained the variation in the matrix of similarity values.

env_assem A data frame of 10 environmental variables, which best explained the variation in the matrix of similarity values.

The data frame abund has the following variables:

Camponotus.mackayensis (numeric) relative abundance of *Camponotus mackayensis*

Cardiocondyla..nuda (numeric) relative abundance of *Cardiocondyla nuda*

Hypoponera.sp..A (numeric) relative abundance of *Hypoponera* spA

Hypoponera.sp..B (numeric) relative abundance of *Hypoponera* spB

Iridomyrmex.sp..A (numeric) relative abundance of *Iridomyrmex* spA

Monomorium.leave (numeric) relative abundance of *Monomorium leave*

Ochetellus.sp..A (numeric) relative abundance of *Ochetellus* spA

Paratrechina.longicornis (numeric) relative abundance of *Paratrechina longicornis*

Paratrechina.sp..A (numeric) relative abundance of *Paratrechina* spA

Tapinoma.sp..A (numeric) relative abundance of *Tapinoma* spA

Tetramorium.bicarinatum (numeric) relative abundance of *Tetramorium bicarinatum*

The data frame env_sp has the following variables:

NativePlSp (numeric) native plant species richness

P.megaAbund (numeric) log-transformed relative abundance of *Pheidole megacephala*

P.megaPA (numeric) presence/absence of *Pheidole megacephala*

HumanVisit (numeric) presence/absence of frequent human visitiation

MaxTemp (numeric) mean daily maximum temp(degree celsius)

Rain4wk (numeric) total rainfall in the past 4 weeks (mm)

DistContinent (numeric) distance to the nearest continent (km)

DistNrIs (numeric) log-transformed distance to the nearest island (km)

Y (numeric) Y coordinate

XY (numeric) X coordinate * Y coordinate

The data frame env_assem has the following variables:

IslandSize (numeric) log-transformed island size (ha)

ExoticPlSp (numeric) log-transformed exotic plant species richness

NativePlSp (numeric) native plant species richness

P.megaPA   (numeric) presence/absence of *Pheidole megacephala*

HumanVisit   (numeric) presence/absence of frequent human visitiation

Rainsamp   (numeric) log-transformed total rainfall during sampling (mm)

DistContinent   (numeric) distance to the nearest continent (km)

DistNrIs   (numeric) log-transformed distance to the nearest island (km)

Y   (numeric) Y coordinate

XY   (numeric) X coordinate * Y coordinate

References

Nakamura A., Burwell C.J., Lambkin C.L., Katabuchi M., McDougall A., Raven R.J. and Neldner V.J. (2015), The role of human disturbance in island biogeography of arthropods and plants: an information theoretic approach, Journal of Biogeography, DOI: 10.1111/jbi.12520

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maglm	*Model averaging for generalized linear models*

Description

Model averaging for GLM based on information theory.

Usage

```
maglm(data, y, family, scale = TRUE, AIC.restricted = FALSE)
```

Arguments

data	Data frame, typically of environmental variables. Rows for sites and colmuns for environmental variables.
y	Vector of independent variables.
family	the 'family' object used.
scale	Whether to scale independent variables (default = TRUE)
AIC.restricted	Whether to use AICc (TRUE) or AIC (FALSE) (default = TRUE).

Value

A list of results

res.table	data frame with "AIC", AIC of the model, "log.L", log-likelihood of the model, "delta.aic", AIC difference to the best model, "wAIC", weighted AIC to the model, "n.vars", number of variables in the model, and each term.
importance	vector of relative importance value of each term, caluclated as as um of the weighted AIC over all of the model in whith the term aperars.
family	the 'family' object used.
scale	Whether to scale independent variables (default = TRUE
AIC.restricted	Whether to use AICc (TRUE) or AIC (FALSE) (default = TRUE).

References

Dobson, A. J. (1990) An Introduction to Generalized Linear Models. London: Chapman and Hall.

Burnham, K.P. & Anderson, D.R. (2002) Model selection and multi-model inference: a practical information-theoretic approach. Springer Verlag, New York.

Nakamura, A., C. J. Burwell, C. L. Lambkin, M. Katabuchi, A. McDougall, R. J. Raven, and V. J. Neldner. (2015) The role of human disturbance in island biogeography of arthropods and plants: an information theoretic approach. Journal of Biogeography 42:1406-1417.

See Also

mamglm, ses.maglm, ses.mamglm

Examples

```
#load species composition and environmental data
data(capcay)
adj.sr <- capcay$adj.sr
env_sp <- capcay$env_sp

#to fit a regression model:
maglm(data = env_sp, y = "adj.sr", family = "gaussian", AIC.restricted = TRUE)
```

---

make.formula	*Utility function*

Description

Utility function for data manipulation, which is implemented in maglm and mamglm.

Usage

```
make.formula(lhs, vars.vec, rand.vec = NULL)
```

Arguments

lhs	Numeric vector of dependent variables.
vars.vec	Character vector of independet variables.
rand.vec	Character vector of random variables (default = NULL).

Value

an object of class '"formula"'

See Also

maglm, mamglm

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mamgllvm	*Model averaging for multivariate generalized linear latent variable models*

Description

Model averaging for multivariate GLLVM based on information theory.

Usage

```
mamgllvm(data, y, family, scale = TRUE, AIC.restricted = FALSE)
```

Arguments

data	Data frame, typically of environmental variables. Rows for sites and colmuns for environmental variables.
y	Name of 'mvabund' object (character)
family	the 'family' object used.
scale	Whether to scale independent variables (default = TRUE)
AIC.restricted	Whether to use AICc (TRUE) or AIC (FALSE) (default = TRUE).

Value

A list of results

res.table	data frame with "AIC", AIC of the model, "log.L", log-likelihood of the model, "delta.aic", AIC difference to the best model, "wAIC", weighted AIC to the model, "n.vars", number of variables in the model, and each term.
importance	vector of relative importance value of each term, caluclated as as um of the weighted AIC over all of the model in whith the term aperars.
family	the 'family' object used.

References

Burnham, K.P. & Anderson, D.R. (2002) Model selection and multi-model inference: a practical information-theoretic approach. Springer Verlag, New York.

Niku, J., Warton, D. I., Hui, F. K. C., and Taskinen, S. (2017). Generalized linear latent variable models for multivariate count and biomass data in ecology. Journal of Agricultural, Biological, and Environmental Statistics, 22:498-522.

Niku, J., Brooks, W., Herliansyah, R., Hui, F. K. C., Taskinen, S., and Warton, D. I. (2018). Efficient estimation of generalized linear latent variable models. PLoS One, 14(5):1-20.

Warton, D. I., Guillaume Blanchet, F., O'Hara, R. B., Ovaskainen, O., Taskinen, S., Walker, S. C. and Hui, F. K. C. (2015). So many variables: Joint modeling in community ecology. Trends in Ecology & Evolution, 30:766-779.

See Also

maglm, ses.maglm, ses.mamglm

Examples

```
#load species composition and environmental data
library(mvabund)
data(capcay)
#use a subset of data in this example to reduce run time
env_assem <- capcay$env_assem[, 1:2]
freq.abs <- mvabund(log(capcay$abund + 1))

#to fit a gaussian regression model to frequency data:
mamgllvm(data = env_assem, y = "freq.abs", family = "gaussian")

#to fit a binomial regression model to presence/absence data"
pre.abs0 <- capcay$abund
pre.abs0[pre.abs0 > 0] = 1
pre.abs <- mvabund(pre.abs0)

mamgllvm(data = env_assem, y = "pre.abs", family = "binomial")
```

---

mamglm   *Model averaging for multivariate generalized linear models*

Description

Model averaging for multivariate GLM based on information theory.

Usage

```
mamglm(data, y, family, scale = TRUE, AIC.restricted = FALSE)
```

Arguments

data	Data frame, typically of environmental variables. Rows for sites and colmuns for environmental variables.
y	Name of 'mvabund' object (character)
family	the 'family' object used.
scale	Whether to scale independent variables (default = TRUE)
AIC.restricted	Whether to use AICc (TRUE) or AIC (FALSE) (default = TRUE).

Value

A list of results

res.table   data frame with "AIC", AIC of the model, "log.L", log-likelihood of the model,
               "delta.aic", AIC difference to the best model, "wAIC", weighted AIC to the
               model, "n.vars", number of variables in the model, and each term.

importance   vector of relative importance value of each term, caluclated as as um of the
               weighted AIC over all of the model in whith the term aperars.

family   the 'family' object used.

References

Burnham, K.P. & Anderson, D.R. (2002) Model selection and multi-model inference: a practical
information-theoretic approach. Springer Verlag, New York.

Wang, Y., Naumann, U., Wright, S.T. & Warton, D.I. (2012) mvabund- an R package for model-
based analysis of multivariate abundance data. Methods in Ecology and Evolution, 3, 471-474.

Warton, D.I., Wright, S.T. & Wang, Y. (2012) Distance-based multivariate analyses confound loca-
tion and dispersion effects. Methods in Ecology and Evolution, 3, 89-101.

Nakamura, A., C. J. Burwell, C. L. Lambkin, M. Katabuchi, A. McDougall, R. J. Raven, and V. J.
Neldner. (2015) The role of human disturbance in island biogeography of arthropods and plants: an
information theoretic approach. Journal of Biogeography 42:1406-1417.

See Also

maglm, ses.maglm, ses.mamglm

Examples

```
#load species composition and environmental data
library(mvabund)
data(capcay)
#use a subset of data in this example to reduce run time
env_assem <- capcay$env_assem[, 1:5]
freq.abs <- mvabund(log(capcay$abund + 1))

#to fit a gaussian regression model to frequency data:
mamglm(data = env_assem, y = "freq.abs", family = "gaussian")

#to fit a binomial regression model to presence/absence data"
pre.abs0 <- capcay$abund
pre.abs0[pre.abs0 > 0] = 1
pre.abs <- mvabund(pre.abs0)

mamglm(data = env_assem, y = "pre.abs", family = "binomial")
```

---

ses.maglm | *Standardized effect size of relative importance values for mamglm*

Description

Standardized effect size of relative importance values for model averaging mutlivariate GLM.

Usage

```
ses.maglm(
  data,
  y,
  family,
  scale = TRUE,
  AIC.restricted = TRUE,
  par = FALSE,
  runs = 999
)
```

Arguments

data	Data frame, typically of environmental variables. Rows for sites and colmuns for environmental variables.
y	Vector of independent variables.
family	the 'family' object used.
scale	Whether to scale independent variables (default = TRUE)
AIC.restricted	Wheter to use AICc (TRUE) or AIC (FALSE) (default = TRUE).
par	Wheter to use parallel computing (default = FALSE)
runs	Number of randomizations.

Details

The currently implemented null model shuffles the set of environmental variables across sites, while maintains species composition. Note that the function would take considerable time to execute.

Value

A data frame of resluts for each term

res.obs	Observed importance of terms
res.rand.mean	Mean importance of terms in null communites
res.rand.sd	Standard deviation of importance of terms in null communites
SES	Standardized effect size of importance of terms (= (res.obs - res.rand.mean) / res.rand.sd)
res.obs.rank	Rank of observed importance of terms vs. null communites
runs	Number of randomizations

References

Dobson, A. J. (1990) An Introduction to Generalized Linear Models. London: Chapman and Hall.

Burnham, K.P. & Anderson, D.R. (2002) Model selection and multi-model inference: a practical information-theoretic approach. Springer Verlag, New York.

Nakamura, A., C. J. Burwell, C. L. Lambkin, M. Katabuchi, A. McDougall, R. J. Raven, and V. J. Neldner. (2015) The role of human disturbance in island biogeography of arthropods and plants: an information theoretic approach. Journal of Biogeography 42:1406-1417.

See Also

maglm, mamglm, ses.mamglm

Examples

```
library(mvabund)
#load species composition and environmental data
data(capcay)
adj.sr <- capcay$adj.sr
#use a subset of data in this example to reduce run time
env_sp <- capcay$env_sp[, 1:5]

#to execute calculations on a single core:
ses.maglm(data = env_sp, y = "adj.sr", par = FALSE,
        family = "gaussian", runs = 4)

## Not run:
#to execute parallel calculations:
sfInit(parallel = TRUE, cpus = 4)
sfExportAll()
ses.maglm(data = env_sp, y = "adj.sr", par = TRUE,
        family = "gaussian", runs = 4)

## End(Not run)
```

| ses.mamglm | *Standardized effect size of relative importance values for mamglm* |

Description

Standardized effect size of relative importance values for model averaging GLM.

Usage

```
ses.mamglm(
  data,
  y,
  family,
```

```
    scale = TRUE,
    AIC.restricted = TRUE,
    par = FALSE,
    runs = 999
)
```

Arguments

`data`	Data frame, typically of environmental variables. Rows for sites and colmuns for environmental variables.
`y`	Name of 'mvabund' object (character)
`family`	the 'family' object used.
`scale`	Whether to scale independent variables (default = TRUE)
`AIC.restricted`	Wheter to use AICc (TRUE) or AIC (FALSE) (default = TRUE).
`par`	Wheter to use parallel computing (default = FALSE)
`runs`	Number of randomizations.

Details

The currently implemented null model shuffles the set of environmental variables across sites, while maintains species composition. Note that the function would take considerable time to execute.

Value

A data frame of resluts for each term

`res.obs`	Observed importance of terms
`res.rand.mean`	Mean importance of terms in null communites
`res.rand.sd`	Standard deviation of importance of terms in null communites
`SES`	Standardized effect size of importance of terms (= (res.obs - res.rand.mean) / res.rand.sd)
`res.obs.rank`	Rank of observed importance of terms vs. null communites
`runs`	Number of randomizations

References

Burnham, K.P. & Anderson, D.R. (2002) Model selection and multi-model inference: a practical information-theoretic approach. Springer Verlag, New York.

Wang, Y., Naumann, U., Wright, S.T. & Warton, D.I. (2012) mvabund- an R package for model-based analysis of multivariate abundance data. Methods in Ecology and Evolution, 3, 471-474.

Warton, D.I., Wright, S.T. & Wang, Y. (2012) Distance-based multivariate analyses confound location and dispersion effects. Methods in Ecology and Evolution, 3, 89-101.

Nakamura, A., C. J. Burwell, C. L. Lambkin, M. Katabuchi, A. McDougall, R. J. Raven, and V. J. Neldner. (2015) The role of human disturbance in island biogeography of arthropods and plants: an information theoretic approach. Journal of Biogeography 42:1406-1417.

Examples

```
library(mvabund)
#load species composition and environmental data
data(capcay)
#use a subset of data in this example to reduce run time
env_assem <- capcay$env_assem[, 1:5]
pre.abs0 <- capcay$abund
pre.abs0[pre.abs0 > 0] = 1
pre.abs <- mvabund(pre.abs0)

#to execute calculations on a single core:
ses.mamglm(data = env_assem, y = "pre.abs",
           par = FALSE, family = "binomial",
           AIC.restricted=FALSE,runs=4)

## Not run:
#to execute parallel calculations:
sfInit(parallel = TRUE, cpus = 4)
sfExportAll()
ses.mamglm(data = env_assem, y = "pre.abs",
           par = TRUE, family = "binomial",
           AIC.restricted = FALSE, runs = 4)

## End(Not run)
```

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