

Package ‘gamlss.mx’

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Type Package

Title Fitting Mixture Distributions with GAMLSS

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LazyLoad yes

Depends R (>= 2.2.1), gamlss.dist, gamlss, nnet, stats, graphics

Suggests MASS

Description The main purpose of this package is to allow fitting of
mixture distributions with generalised additive models for location
scale and shape models see Chapter 7 of Stasinopoulos et al. (2017) <doi:10.1201/b21973-4>.

License GPL-2 | GPL-3

URL <https://www.gamlss.com/>

NeedsCompilation no

Repository CRAN

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gamlss.mx-package

The GAMLSS add on package for mixture distributions

Description

The main purpose of this package is to allow the user of the GAMLSS models to fit mixture distributions.

Details

Package: gamlss.mx
Type: Package
Version: 0.0
Date: 2005-08-3
License: GPL (version 2 or later)

This package has two main function the `gamlssMX()` which is loosely based on the package `flexmix` of R and the function `gamlssNP()` which is based on the `npmlreg` package of Jochen Einbeck, Ross Darnell and John Hinde (2006) which in turns is based on several GLIM4 macros originally written by Murray Aitkin and Brian Francis. It also contains the function `gqz()` which is written by Nick Sofroniou and the function `gauss.quad()` written by Gordon Smyth.

Author(s)

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References

- Jochen Einbeck, Ross Darnell and John Hinde (2006) `npmlreg`: Nonparametric maximum likelihood estimation for random effect models, R package version 0.34
- Rigby, R. A. and Stasinopoulos D. M. (2005). Generalized additive models for location, scale and shape,(with discussion), *Appl. Statist.*, **54**, part 3, pp 507-554.
- Rigby, R. A., Stasinopoulos, D. M., Heller, G. Z., and De Bastiani, F. (2019) *Distributions for modeling location, scale, and shape: Using GAMLSS in R*, Chapman and Hall/CRC. An older version can be found in <https://www.gamlss.com/>.
- Stasinopoulos D. M. Rigby R.A. (2007) Generalized additive models for location scale and shape (GAMLSS) in R. *Journal of Statistical Software*, Vol. **23**, Issue 7, Dec 2007, <https://www.jstatsoft.org/v23/i07/>.
- Stasinopoulos D. M., Rigby R.A., Heller G., Voudouris V., and De Bastiani F., (2017) *Flexible Regression and Smoothing: Using GAMLSS in R*, Chapman and Hall/CRC.
- Stasinopoulos M.D., Kneib T, Klein N, Mayr A, Heller GZ. (2024) Generalized Additive Models for Location, Scale and Shape: A Distributional Regression Approach, with Applications. Cambridge University Press.
- (see also <https://www.gamlss.com/>).

See Also

[gamlss,gamlss.family](#)

Examples

```
data(enzyme)
mmNO <- gamlssMX(act~1, family=NO, K=2, data=enzyme)
mmNO

# also to make sure that it reaches the maximum
mmNOs <- gamlssMXfits(n=10, act~1, family=NO, K=2, data=enzyme)
fyNO<-dMX(y=seq(0,3,.01), mu=list(1.253, 0.1876), sigma=list(exp(-0.6665), exp(-2.573)),
          pi=list(0.4079609, 0.5920391), family=list("NO","NO"))
hist(enzyme$act,freq=FALSE,ylim=c(0,3.5),xlim=c(0,3),br=21)
lines(seq(0,3,.01),fyNO, col="red")
# equivalent model using gamlssNP
mmNP <- gamlssNP(act~1, data=enzyme, random=~1,sigma.fo=~MASS,family=NO, K=2)
```

dMX

Evaluate the d (pdf) and p (cdf) functions from GAMLSS mixtures

Description

The functions `dMX()` and `pMX()` can be used to evaluate the pdf (p function) and the cdf (p function) respectively from a `gamlss.family` mixture. The functions `getpdfMX()` and `getpdfNP()` can be used to evaluate the fitted d function at a specified observation and therefore for plotting the fitted distribution of a fitted model at this observation.

Usage

```
dMX(y, mu = list(mu1 = 1, mu2 = 5), sigma = list(sigma1 = 1, sigma2 = 1),
    nu = list(nu1 = 1, nu2 = 1), tau = list(tau1 = 1, tau2 = 1),
    pi = list(pi1 = 0.2, pi2 = 0.8), family = list(fam1 = "NO", fam2 = "NO"),
    log = FALSE, ...)
pMX(q, mu = list(mu1 = 1, mu2 = 5), sigma = list(sigma1 = 1, sigma2 = 1),
    nu = list(nu1 = 1, nu2 = 1), tau = list(tau1 = 1, tau2 = 1),
    pi = list(pi1 = 0.2, pi2 = 0.8), family = list(fam1 = "NO", fam2 = "NO"),
    log = FALSE, ...)
getpdfMX(object = NULL, observation = 1)
getpdfNP(object = NULL, observation = 1)
```

Arguments

<code>y, q</code>	vector of quantiles
<code>mu</code>	a vector of mu's
<code>sigma</code>	a vector of sigma's

nu	a vector of nu's
tau	a vector of tau's
pi	a vector of pi's
family	a vector of GAMLSS family's
log	whether the log of the function or not
object	a fitted gamlssMX object
observation	the observation number in which we want to plot the fitted mixture
...	for extra arguments

Value

Returns values or pdf or cdf.

Author(s)

Mikis Stasinopoulos

References

Rigby, R. A. and Stasinopoulos D. M. (2005). Generalized additive models for location, scale and shape,(with discussion), *Appl. Statist.*, **54**, part 3, pp 507-554.

Rigby, R. A., Stasinopoulos, D. M., Heller, G. Z., and De Bastiani, F. (2019) *Distributions for modeling location, scale, and shape: Using GAMLSS in R*, Chapman and Hall/CRC. An older version can be found in <https://www.gamlss.com/>.

Stasinopoulos D. M. Rigby R.A. (2007) Generalized additive models for location scale and shape (GAMLSS) in R. *Journal of Statistical Software*, Vol. **23**, Issue 7, Dec 2007, <https://www.jstatsoft.org/v23/i07/>.

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Stasinopoulos M.D., Kneib T, Klein N, Mayr A, Heller GZ. (2024) Generalized Additive Models for Location, Scale and Shape: A Distributional Regression Approach, with Applications. Cambridge University Press.

(see also <https://www.gamlss.com/>).

Examples

```
fyNO<-dMX(y=seq(0,3,.01), mu=list(1.253, 0.1876), sigma=list(exp(-0.6665), exp(-2.573)),
          pi=list(0.4079609, 0.5920391), family=list("NO","NO"))
plot(fyNO~seq(0,3,.01), type="l")
FyNO<-pMX(q=seq(0,3,.01), mu=list(1.253, 0.1876), sigma=list(exp(-0.6665), exp(-2.573)),
          pi=list(0.4079609, 0.5920391), family=list("NO","NO"))
plot(FyNO~seq(0,3,.01), type="l")
```

enzyme

Data used in gamlss.mx

Description

enzyme : The data comprise independent measurement of enzyme activity in the blood of 245 individuals. The data were analysed by Bechker *at al.* (1993).

brains : the brain size, brain, and body weight, body, for 28 differnt animals.

Usage

```
data(enzyme)
data(brains)
```

Format

enzyme : data frame with 245 observations on the following variable act.

brains : data frame with 28 observations on the following variables. body, brain

act a numeric vector showing enzyme activity in the blood of 245 individuals.

body a numeric vector showing the body weight of 28 differnt animals

brain a numeric vector showing the brain size of 28 differnt animals

References

Rigby, R. A. and Stasinopoulos D. M. (2005). Generalized additive models for location, scale and shape,(with discussion), *Appl. Statist.*, **54**, part 3, pp 507-554.

Rigby, R. A., Stasinopoulos, D. M., Heller, G. Z., and De Bastiani, F. (2019) *Distributions for modeling location, scale, and shape: Using GAMLSS in R*, Chapman and Hall/CRC. An older version can be found in <https://www.gamlss.com/>.

Stasinopoulos D. M. Rigby R.A. (2007) Generalized additive models for location scale and shape (GAMLSS) in R. *Journal of Statistical Software*, Vol. **23**, Issue 7, Dec 2007, <https://www.jstatsoft.org/v23/i07/>.

Stasinopoulos D. M., Rigby R.A., Heller G., Voudouris V., and De Bastiani F., (2017) *Flexible Regression and Smoothing: Using GAMLSS in R*, Chapman and Hall/CRC.

Stasinopoulos M.D., Kneib T, Klein N, Mayr A, Heller GZ. (2024) Generalized Additive Models for Location, Scale and Shape: A Distributional Regression Approach, with Applications. Cambridge University Press.

(see also <https://www.gamlss.com/>).

Examples

```

data(enzyme)
hist(enzyme$act)
data(brains)
brains$lbrain<-log(brains$brain)
brains$lbody<-log(brains$body)
with(brains, plot(lbrain~lbody))

```

gamlssMX

Function to fit finite mixture of gamlss family distributions

Description

The function gamlssMX is design for fitting a K fold non parametric mixture of gamlss family distributions.

Usage

```

gamlssMX(formula = formula(data), pi.formula = ~1,
          family = "NO", weights, K = 2, prob = NULL,
          data, control = MX.control(...),
          g.control = gamlss.control(trace = FALSE, ...),
          zero.component = FALSE, ...)
gamlssMXfits(n = 5, formula = formula(data), pi.formula = ~1,
             family = "NO", weights, K = 2, prob = NULL,
             data, control = MX.control(),
             g.control = gamlss.control(trace = FALSE),
             zero.component = FALSE, ... )

```

Arguments

formula	This argument it should be a formula (or a list of formulae of length K) for modelling the mu parameter of the model. Note that modelling the rest of the distributional parameters it can be done by using the usual ... which passes the arguments to gamlss()
pi.formula	This should be a formula for modelling the prior probabilities as a function of explanatory variables. Note that no smoothing of other additive terms are allowed here only the usual linear terms. The modelling here is done using the multinom() function from package nnet
family	This should be a gamlss.family distribution (or a list of distributions). Note that if different distributions are used here their parameters should be comparable for ease of interpretation.
weights	prior weights if needed
K	the number of finite mixtures with default K=2
prob	prior probabilities if required for starting values

data	the data frame needed for the fit. Note that this is compulsory if <code>pi</code> formula is used.
control	This argument sets the control parameters for the EM iterations algorithm. The default settings are given in the <code>MX.control</code> function
g.control	This argument can be used to pass to <code>gamlss()</code> control parameters, as in <code>gamlss.control</code>
n	the number of fits required in <code>gamlssMXfits()</code>
zero.component	whether zero component models exist, default is FALSE
...	for extra arguments

Author(s)

Mikis Stasinopoulos and Bob Rigby

References

Rigby, R. A. and Stasinopoulos D. M. (2005). Generalized additive models for location, scale and shape, (with discussion), *Appl. Statist.*, **54**, part 3, pp 507-554.

Rigby, R. A., Stasinopoulos, D. M., Heller, G. Z., and De Bastiani, F. (2019) *Distributions for modeling location, scale, and shape: Using GAMLSS in R*, Chapman and Hall/CRC. An older version can be found in <https://www.gamlss.com/>.

Stasinopoulos D. M. Rigby R.A. (2007) Generalized additive models for location scale and shape (GAMLSS) in R. *Journal of Statistical Software*, Vol. **23**, Issue 7, Dec 2007, <https://www.jstatsoft.org/v23/i07/>.

Stasinopoulos D. M., Rigby R.A., Heller G., Voudouris V., and De Bastiani F., (2017) *Flexible Regression and Smoothing: Using GAMLSS in R*, Chapman and Hall/CRC.

Stasinopoulos M.D., Kneib T, Klein N, Mayr A, Heller GZ. (2024) Generalized Additive Models for Location, Scale and Shape: A Distributional Regression Approach, with Applications. Cambridge University Press.

(see also <https://www.gamlss.com/>).

See Also

[gamlss](#), [gamlss.family](#)

Examples

```
library(MASS)
data(geyser)
# fitting 2 finite normal mixtures
m1<-gamlssMX(waiting~1,data=geyser,family=NO, K=2)

#fitting 2 finite gamma mixtures
m2<-gamlssMX(waiting~1,data=geyser,family=GA, K=2)
# fitting a model for pi
# first create a data frame
geyser1<-matrix(0,ncol=2, nrow=298)
geyser1[,1] <-geyser$waiting[-1]
```

```

geyser1[,2] <-geyser$duration[-299]
colnames(geyser1)<- c("waiting", "duration")
geyser1 <-data.frame(geyser1)
# get the best of 5 fits
m3<-gamlssMXfits(n=5, waiting~1, pi.formula=~duration, data=geyser1,family=NO, K=2)
m3

```

gamlssNP

A function to fit finite mixtures using the gamlss family of distributions

Description

This function will fit a finite (or normal) mixture distribution where the kernel distribution can belong to any gamlss family of distributions using the EM algorithm. The function is based on functions `alldist()` and `allvc` of the `npmlreg` package of Jochen Einbeck, John Hinde and Ross Darnell.

Usage

```

gamlssNP(formula, random = ~1, family = NO(), data = NULL, K = 4,
          mixture = c("np", "gq"),
          tol = 0.5, weights, pluginz, control = NP.control(...),
          g.control = gamlss.control(trace = FALSE, ...), ...)

```

Arguments

<code>formula</code>	a formula defining the response and the fixed effects for the μ parameters
<code>random</code>	a formula defining the random part of the model
<code>family</code>	a gamlss family object
<code>data</code>	the data frame which for this function is mandatory even if it the data are attached
<code>K</code>	the number of mass points/integretion points (supported values are 1:10,20)
<code>mixture</code>	the mixing distribution, "np" for non-parametric or "gq" for Gaussian Quadrature
<code>tol</code>	the toletance scalar usually between zero and one
<code>weights</code>	prior weights
<code>pluginz</code>	optional
<code>control</code>	this sets the control parameters for the EM iterations algorithm. The default setting is the <code>NP.control</code> function
<code>g.control</code>	the gamlss control function, <code>gamlss.control</code> , passed to the gamlss fit
<code>...</code>	for extra arguments

Details

The function `gamlssNP()` is a modification of the R functions `alldist()` and `allvc` created by Jochen Einbeck and John Hinde. Both functions were originally created by Ross Darnell (2002). Here the two functions are merged to one `gamlssNP` and allows finite mixture from `gamlss` family of distributions.

The following are comments from the original Einbeck and Hinde documentation.

"The nonparametric maximum likelihood (NPML) approach was introduced in Aitkin (1996) as a tool to fit overdispersed generalized linear models. Aitkin (1999) extended this method to generalized linear models with shared random effects arising through variance component or repeated measures structure. Applications are two-stage sample designs, when firstly the primary sampling units (the upper-level units, e.g. classes) and then the secondary sampling units (lower-level units, e.g. students) are selected, or longitudinal data. Models of this type have also been referred to as multi-level models (Goldstein, 2003). This R function is restricted to 2-level models. The idea of NPML is to approximate the unknown and unspecified distribution of the random effect by a discrete mixture of k exponential family densities, leading to a simple expression of the marginal likelihood, which can then be maximized using a standard EM algorithm. When option 'gq' is set, then Gauss-Hermite masses and mass points are used and considered as fixed, otherwise they serve as starting points for the EM algorithm. The position of the starting points can be concentrated or extended by setting `tol` smaller or larger than one, respectively. Variance component models with random coefficients (Aitkin, Hinde & Francis, 2005, p. 491) are also possible, in this case the option `random.distribution` is restricted to the setting 'np'. The weights have to be understood as frequency weights, i.e. setting all weights equal to 2 will duplicate each data point and hence double the disparity and deviance. Warning: There might be some options and circumstances which had not been tested and where the weights do not work." Note that in keeping with the `gamlss` notation disparity is called global deviance.

Value

The function `gamlssNP` produces an object of class "gamlssNP". This object contain several components.

<code>family</code>	the name of the <code>gamlss</code> family
<code>type</code>	the type of distribution which in this case is "Mixture"
<code>parameters</code>	the parameters for the kernel <code>gamlss</code> family distribution
<code>call</code>	the call of the <code>gamlssNP</code> function
<code>y</code>	the response variable
<code>bd</code>	the binomial demominator, only for BI and BB models
<code>control</code>	the NP.control settings
<code>weights</code>	the vector of weights of te expanded fit
<code>G.deviance</code>	the global deviance
<code>N</code>	the number of observations in the fit
<code>rqres</code>	a function to calculate the normalized (randomized) quantile residuals of the object (here is the <code>gamlss</code> object rather than <code>gamlssNP</code> and it should change??)
<code>iter</code>	the number of external iterations in the last <code>gamlss</code> fitting (?? do we need this?)

type	the type of the distribution or the response variable here set to "Mixture"
method	which algorithm is used for the gamlss fit, RS(), CG() or mixed()
contrasts	the type of contrasts use in the fit
converged	whether the gamlss fit has converged
residuals	the normalized (randomized) quantile residuals of the model
mu.fv	the fitted values of the extended mu model, also sigma.fv, nu.fv, tau.fv for the other parameters if present
mu.lp	the linear predictor of the extended mu model, also sigma.lp, nu.lp, tau.lp for the other parameters if present
mu.wv	the working variable of the extended mu model, also sigma.wv, nu.wv, tau.wv for the other parameters if present
mu.wt	the working weights of the mu model, also sigma.wt, nu.wt, tau.wt for the other parameters if present
mu.link	the link function for the mu model, also sigma.link, nu.link, tau.link for the other parameters if present
mu.terms	the terms for the mu model, also sigma.terms, nu.terms, tau.terms for the other parameters if present
mu.x	the design matrix for the mu, also sigma.x, nu.x, tau.x for the other parameters if present
mu.qr	the QR decomposition of the mu model, also sigma.qr, nu.qr, tau.qr for the other parameters if present
mu.coefficients	the linear coefficients of the mu model, also sigma.coefficients, nu.coefficients, tau.coefficients for the other parameters if present
mu.formula	the formula for the mu model, also sigma.formula, nu.formula, tau.formula for the other parameters if present
mu.df	the mu degrees of freedom also sigma.df, nu.df, tau.df for the other parameters if present
mu.nl.df	the non linear degrees of freedom, also sigma.nl.df, nu.nl.df, tau.nl.df for the other parameters if present
df.fit	the total degrees of freedom use by the model
df.residual	the residual degrees of freedom left after the model is fitted
data	the original data set
EMiter	the number of EM iterations
EMconverged	whether the EM has converged
allresiduals	the residuas for the long fit
mass.points	the estimates mass point (if "np" mixture is used)
K	the number of mass points used
post.prob	contains a matrix of posteriori probabilities,
prob	the estimated mixture probalilities

aic	the Akaike information criterion
sbc	the Bayesian information criterion
formula	the formula used in the expanded fit
random	the random effect formula
pweights	prior weights
ebp	the Empirical Bayes Predictions (Aitkin, 1996b) on the scale of the linear predictor

Note that in case of Gaussian quadrature, the coefficient given at 'z' in coefficients corresponds to the standard deviation of the mixing distribution.

As a by-product, gamlssNP produces a plot showing the global deviance against the iteration number. Further, a plot with the EM trajectories is given. The x-axis corresponds to the iteration number, and the y-axis to the value of the mass points at a particular iteration. This plot is not produced when mixture is set to "gq"

Author(s)

Mikis Stasinopoulos based on function created by Jochen Einbeck John Hinde and Ross Darnell

References

- Aitkin, M. and Francis, B. (1995). Fitting overdispersed generalized linear models by nonparametric maximum likelihood. *GLIM Newsletter* 25 , 37-45.
- Aitkin, M. (1996a). A general maximum likelihood analysis of overdispersion in generalized linear models. *Statistics and Computing* 6 , 251-262.
- Aitkin, M. (1996b). Empirical Bayes shrinkage using posterior random effect means from non-parametric maximum likelihood estimation in general random effect models. *Statistical Modelling: Proceedings of the 11th IWSM 1996* , 87-94.
- Aitkin, M., Francis, B. and Hinde, J. (2005) *Statistical Modelling in GLIM 4*. Second Edition, Oxford Statistical Science Series, Oxford, UK.
- Einbeck, J. & Hinde, J. (2005). A note on NPML estimation for exponential family regression models with unspecified dispersion parameter. Technical Report IRL-GLWY-2005-04, National University of Ireland, Galway.
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- Rigby, R. A. and Stasinopoulos D. M. (2005). Generalized additive models for location, scale and shape,(with discussion), *Appl. Statist.*, **54**, part 3, pp 507-554.
- Rigby, R. A., Stasinopoulos, D. M., Heller, G. Z., and De Bastiani, F. (2019) *Distributions for modeling location, scale, and shape: Using GAMLSS in R*, Chapman and Hall/CRC. An older version can be found in <https://www.gamlss.com/>.
- Stasinopoulos D. M. Rigby R.A. (2007) Generalized additive models for location scale and shape (GAMLSS) in R. *Journal of Statistical Software*, Vol. **23**, Issue 7, Dec 2007, <https://www.jstatsoft.org/v23/i07/>.

Stasinopoulos D. M., Rigby R.A., Heller G., Voudouris V., and De Bastiani F., (2017) *Flexible Regression and Smoothing: Using GAMLSS in R*, Chapman and Hall/CRC.

(see also <https://www.gamlss.com/>).

See Also

[gamlss](#), [gamlss.family](#)

Examples

```
data(enzyme)
# equivalent model using gamlssNP
mmNP1 <- gamlssNP(act~1, data=enzyme, random=~1, family=NO, K=2)
mmNP2 <- gamlssNP(act~1, data=enzyme, random=~1, sigma.fo=~MASS, family=NO, K=2)
AIC(mmNP1, mmNP2)
```

MX.control

The control function for gamlssMX

Description

The function sets controls for the `gamlssMX` function.

Usage

```
MX.control(cc = 1e-04, n.cyc = 200, trace = FALSE,
           seed = NULL, plot = TRUE, sample = NULL, ...)
```

Arguments

<code>cc</code>	convergent criterion for the EM
<code>n.cyc</code>	number of cycles for EM
<code>trace</code>	whether to print the EM iterations
<code>seed</code>	a number for setting the seeds for starting values
<code>plot</code>	whether to plot the sequence of global deviance up to convergence
<code>sample</code>	how large the sample to be in the starting values
<code>...</code>	for extra arguments

Value

Returns a list

Author(s)

Mikis Stasinopoulos and Bob Rigby

References

Rigby, R. A. and Stasinopoulos D. M. (2005). Generalized additive models for location, scale and shape,(with discussion), *Appl. Statist.*, **54**, part 3, pp 507-554.

Rigby, R. A., Stasinopoulos, D. M., Heller, G. Z., and De Bastiani, F. (2019) *Distributions for modeling location, scale, and shape: Using GAMLSS in R*, Chapman and Hall/CRC. An older version can be found in <https://www.gamlss.com/>.

Stasinopoulos D. M. Rigby R.A. (2007) Generalized additive models for location scale and shape (GAMLSS) in R. *Journal of Statistical Software*, Vol. **23**, Issue 7, Dec 2007, <https://www.jstatsoft.org/v23/i07/>.

Stasinopoulos D. M., Rigby R.A., Heller G., Voudouris V., and De Bastiani F., (2017) *Flexible Regression and Smoothing: Using GAMLSS in R*, Chapman and Hall/CRC.

Stasinopoulos M.D., Kneib T, Klein N, Mayr A, Heller GZ. (2024) Generalized Additive Models for Location, Scale and Shape: A Distributional Regression Approach, with Applications. Cambridge University Press.

(see also <https://www.gamlss.com/>).

See Also

[gamlss](#), [gamlssMX](#), [gamlssMXfits](#)

NP.control

Control function for gamlssNP

Description

This is a control function for gamlssNP function.

Usage

```
NP.control(EMcc = 0.001, EMn.cyc = 200, damp = TRUE,
           trace = TRUE, plot.opt = 3, ...)
```

Arguments

EMcc	convergence criterion for the EM
EMn.cyc	number of cycles for the EM
damp	Not in used
trace	whether to print the EM iterations
plot.opt	plotting the
...	for extra arguments

Value

Returns a list.

Author(s)

Mikis Stasinopoulos

References

Einbeck, J. Darnell R. and Hinde J. (2006) npmlreg: Nonparametric maximum likelihood estimation for random effect models, R package version 0.34

Rigby, R. A. and Stasinopoulos D. M. (2005). Generalized additive models for location, scale and shape,(with discussion), *Appl. Statist.*, **54**, part 3, pp 507-554.

Rigby, R. A., Stasinopoulos, D. M., Heller, G. Z., and De Bastiani, F. (2019) *Distributions for modeling location, scale, and shape: Using GAMLSS in R*, Chapman and Hall/CRC. An older version can be found in <https://www.gamlss.com/>.

Stasinopoulos D. M. Rigby R.A. (2007) Generalized additive models for location scale and shape (GAMLSS) in R. *Journal of Statistical Software*, Vol. **23**, Issue 7, Dec 2007, <https://www.jstatsoft.org/v23/i07/>.

Stasinopoulos D. M., Rigby R.A., Heller G., Voudouris V., and De Bastiani F., (2017) *Flexible Regression and Smoothing: Using GAMLSS in R*, Chapman and Hall/CRC.

Stasinopoulos M.D., Kneib T, Klein N, Mayr A, Heller GZ. (2024) Generalized Additive Models for Location, Scale and Shape: A Distributional Regression Approach, with Applications. Cambridge University Press.

(see also <https://www.gamlss.com/>).

See Also

[gamlss](#), [gamlssNP](#)

plotMP

plotting mass points

Description

A utility function for plotting two dimension non-parametric distribution. The function uses the persp() function.

Usage

```
plotMP(x, y, prob, theta = 20, phi = 20, expand = 0.5, col = "lightblue",  
       xlab = "intercept", ylab = "slope", ...)
```

Arguments

x	a vector containg points in the x axis
y	a vector containg points in the y axis
prob	vector containing probabilities which should add up to one
theta, phi, expand, col	arguments to pass to the persp() function
xlab	the x label
ylab	the y label
...	additinal argument to be passed to persp()

Details

The function call

Value

A graph is produced.

Author(s)

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References

Rigby, R. A. and Stasinopoulos D. M. (2005). Generalized additive models for location, scale and shape,(with discussion), *Appl. Statist.*, **54**, part 3, pp 507-554.

Rigby, R. A., Stasinopoulos, D. M., Heller, G. Z., and De Bastiani, F. (2019) *Distributions for modeling location, scale, and shape: Using GAMLSS in R*, Chapman and Hall/CRC. An older version can be found in <https://www.gamlss.com/>.

Stasinopoulos D. M. Rigby R.A. (2007) Generalized additive models for location scale and shape (GAMLSS) in R. *Journal of Statistical Software*, Vol. **23**, Issue 7, Dec 2007, <https://www.jstatsoft.org/v23/i07/>.

Stasinopoulos D. M., Rigby R.A., Heller G., Voudouris V., and De Bastiani F., (2017) *Flexible Regression and Smoothing: Using GAMLSS in R*, Chapman and Hall/CRC.

Stasinopoulos M.D., Kneib T, Klein N, Mayr A, Heller GZ. (2024) Generalized Additive Models for Location, Scale and Shape: A Distributional Regression Approach, with Applications. Cambridge University Press.

(see also <https://www.gamlss.com/>).

See Also

[gamlssNP](#), [persp](#)

Examples

```
gamma_0 <- c( -4.4, -3,-2.2, -.5, 0.1, 1, 1.5, 2.2, 3.5, 4.1 )
gamma_1 <- c( 2.2, 1.2, 0.1, -1, -2.3, -4.6 , 5.1, -3.2, 0.1, -1.2)
prob <- c(0.1, .05, .12, 0.25, 0.08, 0.12, 0.10, 0.05, 0.10, 0.03 )
plotMP(gamma_0, gamma_1,prob)
```

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