

# Package ‘gamlss.inf’

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**Type** Package

**Title** Fitting Mixed (Inflated and Adjusted) Distributions

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**Imports** methods

**Author** Marco Enea [aut, cre, cph],  
Mikis Stasinopoulos [aut],  
Bob Rigby [aut],  
Abu Hossain [aut]

**Maintainer** Marco Enea <marco.enea@unipa.it>

**Description** This is an add-on package to 'gamlss'. The purpose of this package is to allow users to fit GAMLSS (Generalised Additive Models for Location Scale and Shape) models when the response variable is defined either in the intervals  $[0,1)$ ,  $(0,1]$  and  $[0,1]$  (inflated at zero and/or one distributions), or in the positive real line including zero (zero-adjusted distributions). The mass points at zero and/or one are treated as extra parameters with the possibility to include a linear predictor for both. The package also allows transformed or truncated distributions from the GAMLSS family to be used for the continuous part of the distribution. Standard methods and GAMLSS diagnostics can be used with the resulting fitted object.

**License** GPL-2 | GPL-3

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gamlss.inf-package      *Models for Mixed (Inflated and Adjusted) Response Variables.*

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## Description

This package allows mixed distribution fitting in GAMLSS. A mixed distribution is one containing both continuous and discrete parts, see Chapter 5 of Stasinopoulos et al. (2017). There are some mixed distribution in the GAMLSS implementation in R like the BEINF, BEINF0, BEINF1 for data defined on  $[0,1]$  or ZAGA, ZAIG for data defined on a positive real line but the choice is very limited. This package enhance the availability of mixed distribution within the GAMLSS framework.

For historical reasons the authors use the terminology "Inflated" for models on  $[0,1]$ , "Adjusted" for models on  $[0,Inf]$ . We will follow the same terminology here. So this package allows the fit of an inflated GAMLSS model when the response variable is defined in the intervals  $(0,1)$ ,  $(0,1]$  and  $[0,1]$  and the fit of zero adjusted models when the response variable is defined in the positive real line, (but where there are zeros in the data).

For models with inflated proportion response variables the package provides up to two extra parameters, a mass point at zero and a mass point at one. Adding an extra inflation point at zero (or at one), is equivalent to fit two separate GAMLSS models, a GAMLSS model with a continuous distribution defined at the interval  $(0,1)$ , and a logit model for zero (or ones). When both zero and one are present, i.e.  $[0,1]$ , a multinomial model is needed to fit the non- $(0,1)$  part.

For the zero adjusted models with a response defined on the positive real line (but where zeros exist in the data), the actual fitting can be achieved by fitting two separate GAMLSS models one with a distribution on the real positive line and one binomial model for the zeros and non-zeros.

This package uses the two models fitting procedures but the resulting fitted object behaves like a typical GAMLSS object so a lot of standard GAMLSS diagnostics can be used with it. It also allows transformed or truncated `gamlss.family` distributions to be used for the continuous part of the distribution therefore giving a great flexibility of the type of mixed distribution to be used.

The functions `gamlssInf0to1()` and `gamlssZadj()` can fit the Inflated and Adjusted models, respectively, and they are described in the two vignettes available with the package.

## Details

Package: gamlss.inf  
Type: Package  
Version: 1.0-1  
Date: 2019-03-12  
License: GPL-2 | GPL-3

## Author(s)

Marco Enea <marco.enea@unipa.it>, Mikis Stasinopoulos <mikis.stasinopoulos@gamlss.org>, Bob Rigby and Abu Hossain

Maintainer: Marco Enea <marco.enea@unipa.it>

## References

Hossain, A., Stasinopoulos, M., Rigby, R. and Enea, M. (2015). Centile estimation for a proportion response variable. *Statistics in Medicine*, doi: 10.1002/sim.6748.

Ospina R. and Ferrari S. L. P. (2010) Inflated beta distributions, *Statistical Papers*, **23**, 111-126.

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.

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, <http://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. <https://www.crcpress.com/Flexible-Regression-and-Smoothing-Using-GAMLSS-in-R/Stasinopoulos-Rigby-Heller-Voudouris-Bastiani/book/9781138197909>.

## See Also

[gamlss](#)

## Examples

```
# An artificial example using simulated data in [0,1)

# Firstly, we use function gen.Family() to create
# the logit skew student t (logitSST) distribution which is defined in the (0,1)
# interval. Then we use function gen.Inf0to1() to create the 0-inflated logitSST
# distribution defined in [0,1).
```

```

gen.Family("SST", "logit")
gen.Inf0to1("logitSST", "Zero")

#now we can generate the data and run the model

set.seed(10)
Y <- rlogitSSTInf0(300,mu=0.5,sigma=0.7,nu=0.5,tau=5,xi0=0.5,log=FALSE)
m1 <- gamlssInf0to1(y=Y,mu.formula=~1,
                  sigma.formula=~1,
                  nu.formula=~1,
                  tau.formula=~1,
                  xi0.formula=~1,
                  family=logitSST,
                  trace = TRUE)

summary(m1)

```

---

centiles.Inf0to1

*Plotting centile curves for a gamlssInf0to1 and gamlssZadj object*


---

## Description

Function `centiles.Inf0to1()` plots centile curves for distributions belonging to the GAMLSS family of distributions defined in the intervals (0,1],[0,1) and [0,1]. The function also tabulates the sample percentages below each centile curve (for comparison with the model percentages given by the argument `cent`). A restriction of the function is that it applies to models with one explanatory variable only.

## Usage

```

centiles.Inf0to1(obj, xvar = NULL, cent = c(0.4, 2, 10, 25, 50, 75,
      90, 98, 99.6), legend = TRUE, ylab = "y", xlab = "x", main = NULL,
      main.gsub = "@", xleg = min(xvar), yleg = max(obj$y), xlim = range(xvar),
      ylim = range(obj$y), save = FALSE, plot = TRUE, points = TRUE,
      pch = 15, cex = 0.5, col = gray(0.7), col.centiles = 1:length(cent) +
      2, lty.centiles = 1, lwd.centiles = 1, ...)
centiles.Zadj(obj, xvar = NULL, cent = c(0.4, 2, 10, 25, 50, 75,
      90, 98, 99.6), legend = TRUE, ylab = "y", xlab = "x", main = NULL,
      main.gsub = "@", xleg = min(xvar), yleg = max(obj$y), xlim = range(xvar),
      ylim = range(obj$y), save = FALSE, plot = TRUE, points = TRUE,
      pch = 15, cex = 0.5, col = gray(0.7), col.centiles = 1:length(cent) +
      2, lty.centiles = 1, lwd.centiles = 1, ...)

```

## Arguments

<code>obj</code>	a fitted <code>gamlss</code> object from fitting a <code>gamlss</code> distribution
<code>xvar</code>	the unique explanatory variable

cent	a vector with elements the % centile values for which the centile curves have to be evaluated
legend	whether a legend is required in the plot or not, the default is legend=TRUE
ylab	the y-variable label
xlab	the x-variable label
main	the main title here as character. If NULL the default title "centile curves using NO" (or the relevant distributions name) is shown
main.gsub	if the main.gsub (with default "@") appears in the main title then it is substituted with the default title.
xleg	position of the legend in the x-axis
yleg	position of the legend in the y-axis
xlim	the limits of the x-axis
ylim	the limits of the y-axis
save	whether to save the sample percentages or not with default equal to FALSE. In this case the sample percentages are printed but are not saved
plot	whether to plot the centiles
points	whether the data points should be plotted, default is TRUE
pch	the character to be used as the default in plotting points see par
cex	size of character see par
col	plotting colour see par
col.centiles	Plotting colours for the centile curves
lty.centiles	line type for the centile curves
lwd.centiles	The line width for the centile curves
...	for extra arguments

### Details

Centiles are calculated using the fitted values in `obj` and `xvar` must correspond exactly to the predictor in `obj` to plot correctly.

`col.centiles`, `lty.centiles` and `lwd.centiles` may be vector arguments and are recycled to the length `cent` if necessary.

### Value

A centile plot is produced and the sample centiles below each centile curve are printed (or saved)

### Warning

This function is appropriate only when one continuous explanatory variable is fitted in the model

### Author(s)

Mikis Stasinopoulos <mikis.stasinopoulos@gamlss.org>, Bob Rigby <r.rigby@londonmet.ac.uk> with contribution from Steve Ellison

## 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.
- Stasinopoulos D. M., Rigby R.A. and Akantziliotou C. (2006) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).
- 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, <http://www.jstatsoft.org/v23/i07>.
- Houssain, A., Stasinopoulos, M., Rigby, R. and Enea, M. (2015). Centile estimation for a proportion response variable. Accepted for publication on *Statistics in Medicine*.

## See Also

[gamlssInf0to1](#), [gamlss](#), [centiles.split](#), [centiles.com](#)

## Examples

```
## Not run:
gen.Family("SST", "logit")
gen.Inf0to1("logitSST", "One")
set.seed(30)
x <- seq(0,1,l=2000)
dat <- data.frame(x)
dat$Y <- rlogitSSTInf1(2000,mu=-3+10*x-0.7*x^2,sigma=0.9,nu=0.5,
                      tau=5,xi1=plogis(-0.5*ifelse(x>0.7,-1,13)),log=FALSE)

m1 <- gamlssInf0to1(y=Y,mu.formula=~pb(x), sigma.formula=~pb(x),
                   nu.formula=~pb(x), tau.formula=~pb(x),
                   xi1.formula=~pb(x),
                   data=dat, family=logitSST)

centiles.Inf0to1(m1,xvar= dat$x, cent=c(2,10,25,50,75,90,98),
col.centiles=c(1,7:2),
ylab="proportion", xlab="x",legend=FALSE,main="(c) Inf. logitSST")

## End(Not run)
```

---

`gamlssInf0to1`

*GAMLSS model for a proportion response variable with point(s) mass at 0 and or 1.*

---

**Description**

Function `gamlssInf0to1()` allows to fit inflated gamlss models when the response variable distribution is defined in the intervals  $[0,1)$ ,  $(0,1]$  and  $[0,1]$ . The `gamlssInf0to1` model for inflated proportion variables is a gamlss model provided of up to two extra parameters for the mass point(s). In the case of inflation point at zero (one), this is equivalent to fit two separate models, a gamlss model for the  $(0,1)$  part, and a logit model for zero (one) vs non-zero (non-one) part. When both zero and one are present, a multinomial model is involved to fit the non- $(0,1)$  part.

**Usage**

```
gamlssInf0to1(y = NULL, mu.formula = ~1, sigma.formula = ~1,
              nu.formula = ~1, tau.formula = ~1,
              xi0.formula = ~1, xi1.formula = ~1, data = NULL,
              family = BE, weights = rep(1, length(Y_)),
              trace = FALSE, ...)
```

**Arguments**

<code>y</code>	the proportion response variable with inflation at zero and/or one
<code>mu.formula</code>	a model formula for $\mu$
<code>sigma.formula</code>	a model formula for $\sigma$
<code>nu.formula</code>	a model formula for $\nu$
<code>tau.formula</code>	a model formula for $\tau$
<code>xi0.formula</code>	a model formula for the probability at zero
<code>xi1.formula</code>	a model formula for the probability at one
<code>data</code>	a data frame containing the variables occurring in the formula.
<code>family</code>	any gamlss distribution family defined in $(0,1)$
<code>weights</code>	a vector of weights as in <a href="#">gamlss</a>
<code>trace</code>	logical, if TRUE information on model estimation will be printed during the fitting
<code>...</code>	for extra parameters

**Details**

The default family is a Beta distribution (BE), but other  $(0,1)$  distributions can be used, e.g. those generated from existing continuous gamlss family distributions by using [gen.Family](#) with link "logit".

**Value**

returns a `gamlssInf0to1` object which has its own methods

**Author(s)**

Mikis Stasinopoulos, Robert Rigby, Abu Hossain and Marco Enea

## References

- Hossain, A., Stasinopoulos, M., Rigby, R. and Enea, M. (2015). Centile estimation for a proportion response variable. *Statistics in Medicine*, doi: 10.1002/sim.6748.
- 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.
- Stasinopoulos D. M., Rigby R.A. and Akantziliotou C. (2006) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).
- 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, <http://www.jstatsoft.org/v23/i07>.

## See Also

[gamlss.family](#), [BEINF](#), [BE](#), [BEo](#), [BEZI](#), [BE0I](#)

## Examples

```
# 1. An artificial example using simulated data

# Firstly, we use function gen.Family() to create the logit skew
# student t (logitSST) distribution defined in the (0,1) interval,
# and function gen.Inf0to1() to create the 0-inflated logitSST
# distribution defined in [0,1).

gen.Family("SST", "logit")
gen.Inf0to1("logitSST", "Zero")

#now we can generate the data and run the model
set.seed(10)
Y <- rlogitSSTInf0(500,mu=0.5,sigma=0.7,nu=0.5,tau=5,xi0=0.5,log=FALSE)
dat <- data.frame(Y)
dat$x <- rnorm(500)
m1 <- gamlssInf0to1(y=Y,mu.formula=~x, sigma.formula=~x,
                   nu.formula=~x, tau.formula=~x,
                   xi0.formula=~x,data=dat, family=logitSST)
summary(m1)

# 2. Example of equivalent gamlss models for an inflated-at-1 Beta distribution

Y <- rBEINF1(500,mu=0.5,sigma=0.7,nu=0.5)
m2 <- gamlss(Y~1,sigma.formula=~1,nu.formula=~1,family=BEINF1)
m3.1 <- gamlss(Y[Y<1]~1,sigma.formula=~1,family=BE)
m3.2 <- gamlss(I(Y==1)~1,family=BI)
m4 <- gamlssInf0to1(Y,mu.formula=~1,sigma.formula=~1,xi1=~1,family=BE)
stopifnot(all.equal(deviance(m2),(deviance(m3.1)+deviance(m3.2))),
          all.equal(deviance(m2),deviance(m4)))
```



gamlssZadj

*Fitting positive real line response variable with zeros.***Description**

Function `gamlssZadj()` allows to fit zero adjusted `gamlss` models when the response variable distribution is defined on the positive real line. The `gamlssZadj` model for adjusted positive variables is a `gamlss` model provides one extra parameters for the mass point at zero. This is equivalent to fit two separate models, a `gamlss` model for the  $(0, \text{Inf})$  part, and a logit model for zero part versus the non-zero part. The function works similarly but provides one fitted object.

**Usage**

```
gamlssZadj(y = NULL, mu.formula = ~1, sigma.formula = ~1,
           nu.formula = ~1, tau.formula = ~1,
           xi0.formula = ~1, data = NULL,
           family = GA,
           weights = rep(1, length(Y_)), trace = FALSE, ...)
```

**Arguments**

<code>y</code>	the response variable
<code>mu.formula</code>	a model formula for $\mu$
<code>sigma.formula</code>	a model formula for $\sigma$
<code>nu.formula</code>	a model formula for $\nu$
<code>tau.formula</code>	a model formula for $\tau$
<code>xi0.formula</code>	a model formula for $\xi_0$
<code>data</code>	a data frame containing the variables occurring in the formula.
<code>family</code>	any <code>gamlss</code> distribution family defined on the real line
<code>weights</code>	a vector of weights as in <a href="#">gamlss</a>
<code>trace</code>	logical, if TRUE information on model estimation will be printed during the fitting
<code>...</code>	for extra arguments to pass to <code>gamlss</code>

**Details**

The default family is a gamma distribution (GA), but other distributions on the positive real line can be used, e.g. those generated from existing continuous `gamlss` family distributions using say `gen.Family()` with "log" or `gen.trun()` from package **gamlss.tr**

**Value**

. Returns a `gamlssZadj` object which has its own methods

**Author(s)**

Mikis Stasinopoulos, Robert Rigby and Marco Enea

**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.

Stasinopoulos D. M., Rigby R.A. and Akantziliotou C. (2006) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).

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, <http://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. <https://www.crcpress.com/Flexible-Regression-and-Smoothing-Using-GAMLSS-in-R/Stasinopoulos-Rigby-Heller-Voudouris-Bastiani/book/9781138197909>.

**See Also**

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

**Examples**

```
y0 <- rZAGA(1000, mu=.3, sigma=.4, nu=.15)# p0=0.13
g0 <- gamlss(y0~1, family=ZAGA)
t0 <- gamlssZadj(y=y0, mu.formula=~1, family=GA, trace=TRUE)
AIC(g0,t0, k=0)
```

---

gen.Inf0to1

*Functions to generate inflated 0-to-1 distributions from existing continuous gamlss.family distributions defined in (0,1).*

---

**Description**

There are six functions here. Only the function `gen.Inf0to1()` should be used. The remaining five functions will be automatically created once `gen.Inf0to1()` has been run.

**Usage**

```
gen.Inf0to1(family = "BE", type.of.Inflation = c( "Zero&One", "Zero", "One"),
  ...)
Inf0to1.d(family = "BE", type.of.Inflation = c( "Zero&One", "Zero", "One"),
  ...)
Inf0to1.p(family = "BE", type.of.Inflation = c( "Zero&One", "Zero", "One"),
  ...)
```

```

Inf0to1.q(family = "BE", type.of.Inflation = c( "Zero&One", "Zero", "One"),
  ...)
Inf0to1.r(family = "BE", type.of.Inflation = c( "Zero&One", "Zero", "One"),
  ...)
plotInf0to1(family = "BE", type.of.Inflation = c( "Zero&One", "Zero", "One"),
  ...)

```

## Arguments

family	a continuous (0,1) distribution (extremes not included) gamlss.family distribution
type.of.Inflation	the type of inflation
...	for passing extra arguments

## Details

Functions `Inf0to1.d`, `Inf0to1.p`, `Inf0to1.q` and `Inf0to1.r` allow to create the density function, distribution function, quantile function and random generation, respectively. Function `plotInf0to1` can be used to create the plot the distributions.

Alternatively, the function `gen.Inf0to1` creates the all the standard d,p,q,r functions plus the plotting function.

For example, let us take the case of the logit SST distribution with inflation at 1. First generate the "logitSST" distribution by using `gen.Family("SST", "logit")`, and then, by use `gen.Inf0to1("logitSST", "One")`. The functions `dlogitSSTInf1`, `plogitSSTInf1`, `qlogitSSTInf1`, `rlogitSSTInf1` and `plotlogitSSTInf1` will be automatically generated. Note that `gen.Inf0to1` never creates a fitting function of the type "logitSSTInf1", but the existing `logitSST` must be specified instead as an argument family of function `gamlssInf0to1()`.

## Value

The function `gen.Inf0to1` returns the d, p, q and r functions plus the plotting function.

## Author(s)

Mikis Stasinopoulos <mikis.stasinopoulos@gamlss.org>, Bob Rigby and Marco Enea

## 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.
- Stasinopoulos D. M., Rigby R.A. and Akantziliotou C. (2006) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).
- 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, <http://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. <https://www.crcpress.com/Flexible-Regression-and-Smoothing-Using-GAMLSS-in-R/Stasinopoulos-Rigby-Heller-Voudouris-Bastiani/book/9781138197909>.

## Examples

```
# 1.
gen.Inf0to1("BE", "Zero&One")
rBETAInf0to1 <- Inf0to1.r("BE", "Zero&One")
all.equal(rBETAInf0to1, rBEInf0to1)
plotBEInf0to1()
plotBEInf0to1(mu=0.3, sigma=0.35, xi0=0.5, xi1=0.3)

# 2.
gen.Family("SST", "logit")
gen.Inf0to1("logitSST", "One")
set.seed(30)

args(rlogitSSTInf1)
y <- rlogitSSTInf1(1000, mu=0.2, sigma=0.5, nu=1, tau=5, xi1=0.2)
quantile(y, c(0.1, 0.25, 0.5, 0.75, 0.9))

args(qlogitSSTInf1)
qlogitSSTInf1(p=c(0.1, 0.25, 0.5, 0.75, 0.9), mu=0.2, sigma=0.5, nu=1, tau=5, xi1=0.2)
plotlogitSSTInf1(mu=0.2, sigma=0.5, nu=1, tau=5, xi1=0.2)
```

---

gen.Zadj

*Functions to generate zero adjusted distributions from existing continuous gamlss.family distributions defined on positive real line.*

---

## Description

There are six functions here. Only the function `gen.Zadj()` should be used. The remaining four functions will be automatically created once `gen.Zadj()` has been run.

## Usage

```
gen.Zadj(family = "GA", ...)
Zadj.d(family = "GA", ...)
Zadj.p(family = "GA", ...)
Zadj.q(family = "GA", ...)
Zadj.r(family = "GA", ...)
plotZadj(family = "GA", ...)
```

## Arguments

```
family      a continuous positive real line distribution
...         for additional arguments
```

## Details

Functions `Zadj.d`, `Zadj.p`, `Zadj.q` and `Zadj.r` allow to create the density function, distribution function, quantile function and random generation, respectively. Function `plotZadj` can be used to create a plot for the distribution.

Alternatively, the function `gen.Zadj` creates the all the standard `d`, `p`, `q`, `r` functions plus the plotting function.

## Value

The function `gen.Zadj` returns the `d`, `p`, `q` and `r` functions plus the plotting function.

## Author(s)

Mikis Stasinopoulos <[mikis.stasinopoulos@gamlss.org](mailto:mikis.stasinopoulos@gamlss.org)>, Bob Rigby and Marco Enea

## 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.

Stasinopoulos D. M., Rigby R.A. and Akantziliotou C. (2006) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).

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## See Also

[gamlssZadj](#), `~~~`

## Examples

```
# 1.
gen.Zadj("BCT")

plotBCTZadj()
plotBCTZadj(mu=3, sigma=0.35, xi0=0.5)

# 2.
gen.Family("SST", "log")
gen.Zadj("logSST")
plotlogSSTZadj()
```

---

predict.gamlssinf0to1 *Extract Predictor Values and Standard Errors For New Data In a  
gamlssinf0to1 Model*

---

### Description

predict.gamlssinf0to1 is the gamlssinf0to1 specific method which produce predictors for a new data set for a specified parameter from a gamlssinf0to1 objects. The predict.gamlssinf0to1 can be used to extract the linear predictors, fitted values and specific terms in the model at new data values in the same way that the predict.lm() and predict.glm() functions can be used for lm or glm objects. Note that linear predictors, fitted values and specific terms in the model at the current data values can also be extracted using the function lpred() (which is called from predict if new data is NULL).

### Usage

```
## S3 method for class 'gamlssinf0to1'
predict(object, parameter = c("mu", "sigma", "nu", "tau", "xi0", "xi1"),
        newdata = NULL, type = c("link", "response", "terms"),
        terms = NULL, se.fit = FALSE, data = NULL, ...)
```

### Arguments

object	a gamlssinf0to1 fitted model
parameter	which distribution (or inflation) parameter is required, default parameter="mu"
newdata	a data frame containing new values for the explanatory variables used in the model
type	the default, gets the linear predictor for the specified distribution (or inflation) parameter. type="response" gets the fitted values for the parameter while type="terms" gets the fitted terms contribution
terms	if type="terms", which terms to be selected (default is all terms)
se.fit	if TRUE the approximate standard errors of the appropriate type are extracted if exist
data	the data frame used in the original fit if is not defined in the call
...	for extra arguments

### Details

The predict function assumes that the object given in newdata is a data frame containing the right x-variables used in the model. This could possible cause problems if transformed variables are used in the fitting of the original model. For example, let us assume that a transformation of age is needed in the model i.e. nage<-age^.5. This could be fitted as mod<-gamlss(y~cs(age^.5), data=mydata) or as nage<-age^.5; mod<-gamlss(y~cs(nage), data=mydata). The later could more efficient if

the data are in thousands rather in hundreds. In the first case, the code `predict(mod, newdata=data.frame(age=c(34, 56)))` would produce the right results. In the second case a new data frame has to be created containing the old data plus any new transform data. This data frame has to be declared in the `data` option. The option `newdata` should contain a `data.frame` with the new names and the transformed values in which prediction is required, (see the last example).

### Value

A vector or a matrix depending on the options.

### Author(s)

Abu Hossain, Mikis Stasinopoulos <[mikis.stasinopoulos@gamlss.org](mailto:mikis.stasinopoulos@gamlss.org)>, Bob Rigby and Marco Enea

### References

Hossain, A., Stasinopoulos, M., Rigby, R. and Enea, M. (2015). Centile estimation for a proportion response variable. *Statistics in Medicine*, doi: 10.1002/sim.6748.

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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, <http://www.jstatsoft.org/v23/i07>.

### See Also

[gamlssInf0to1](#)

### Examples

```
gen.Family("SST", "logit")
gen.Inf0to1("logitSST", "Zero")
set.seed(10)
Y <- rlogitSSTInf0(500, mu=0.5, sigma=0.7, nu=0.5, tau=5, xi0=0.5, log=FALSE)
dat <- data.frame(Y)
dat$x <- rnorm(500)
m1 <- gamlssInf0to1(y=Y, mu.formula=~x, sigma.formula=~x,
                  nu.formula=~x, tau.formula=~x,
                  xi0.formula=~x, data=dat, family=logitSST)
predict(m1)
```

---

predict.gamlssZadj	<i>Extract Predictor Values and Standard Errors For New Data in a gamlssZadj Model</i>
--------------------	--

---

### Description

predict.gamlssZadj is the specific method which produce predictors for a new data set for a specified parameter from a gamlssZadj objects. The predict.gamlssZadj can be used to extract the linear predictors, fitted values and specific terms in the model at new data values in the same way that the predict.lm() and predict.glm() functions can be used for lm or glm objects. Note that linear predictors, fitted values and specific terms in the model at the current data values can also be extracted using the function lpred() (which is called from predict if new data is NULL).

### Usage

```
## S3 method for class 'gamlssZadj'
predict(object, parameter = c("mu", "sigma", "nu", "tau", "xi0"),
        newdata = NULL, type = c("link", "response", "terms"),
        terms = NULL, se.fit = FALSE, data = NULL, ...)
```

### Arguments

object	a gamlssZadj fitted model
parameter	which distribution (or inflation) parameter is required, default parameter="mu"
newdata	a data frame containing new values for the explanatory variables used in the model
type	the default, gets the linear predictor for the specified distribution (or inflation) parameter. type="response" gets the fitted values for the parameter while type="terms" gets the fitted terms contribution
terms	if type="terms", which terms to be selected (default is all terms)
se.fit	if TRUE the approximate standard errors of the appropriate type are extracted if exist
data	the data frame used in the original fit if is not defined in the call
...	for extra arguments

### Details

The predict function assumes that the object given in newdata is a data frame containing the right x-variables used in the model. This could possible cause problems if transformed variables are used in the fitting of the original model. For example, let us assume that a transformation of age is needed in the model i.e. `nage<-age^.5`. This could be fitted as `mod<-gamlss(y~cs(age^.5), data=mydata)` or as `nage<-age^.5; mod<-gamlss(y~cs(nage), data=mydata)`. The later could more efficient if the data are in thousands rather in hundreds. In the first case, the code `predict(mod, newdata=data.frame(age=c(34, 56)))` would produce the right results. In the second case a new data frame has to be created containing the old data plus any new transform data. This data frame has to be declared in the data option. The option newdata should contain a data.frame with the new names and the transformed values in which prediction is required, (see the last example).



**Value**

A vector or a matrix depending on the options.

**Author(s)**

Abu Hossain, Mikis Stasinopoulos <mikis.stasinopoulos@gamlss.org>, Bob Rigby and Marco Enea

**References**

Hossain, A., Stasinopoulos, M., Rigby, R. and Enea, M. (2015). Centile estimation for a proportion response variable. *Statistics in Medicine*, doi: 10.1002/sim.6748.

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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, <http://www.jstatsoft.org/v23/i07>.

**See Also**

[gamlssZadj](#)

**Examples**

```
set.seed(3210)
x <- (runif(800)*4)-2
data(sda)
fmu <- splinefun(sda$x, sda$mu)
curve(fmu, -2,2)
fsigma <- splinefun(sda$x, sda$sigma)
curve(fsigma, -2,2)
fnu <- function(x)
  {f <- splinefun(sda$x, sda$nu)
  f(x)/6
}
curve(fnu, -2,2)
set.seed(321)
y0 <- rZAGA(800, mu=fmu(x), sigma=fsigma(x), nu=fnu(x))
da <- data.frame(y0,x)
g0p <- gamlss(y0~pb(x), sigma.fo=~pb(x), nu.fo=~pb(x), data=da, family=ZAGA)
t0p <- gamlssZadj(y=y0, mu.fo=~pb(x), sigma.fo=~pb(x), data=da,
  trace=TRUE, xi0.fo=~pb(x), family="GA")

plot(predict(g0p,"nu",type="link"),
  predict(t0p,"xi0",type="link"))
```

---

sda

*Data for using for simulation*

---

### Description

Those data values are used to create simulated data

### Usage

```
data("sda")
```

### Format

A data frame with 120 observations on the following 5 variables.

x the explanatory variable

mu the fitted mu

sigma the fitted sigma

nu the fitted nu

tau the fitted tau

### Source

The data are fitted values of model

### 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.

Stasinopoulos D. M., Rigby R.A. and Akantziliotou C. (2006) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).

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, <http://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. <https://www.crcpress.com/Flexible-Regression-and-Smoothing-Using-GAMLSS-in-R/Stasinopoulos-Rigby-Heller-Voudouris-Bastiani/book/9781138197909>.

**Examples**

```

data(sda)
fmu <- splinefun(sda$x, sda$mu)
curve(fmu, -2,2)
fsigma <- splinefun(sda$x, sda$sigma)
curve(fsigma, -2,2)
fnu <- splinefun(sda$x, sda$nu)
curve(fnu, -2,2)
ftau <- splinefun(sda$x, sda$tau)
curve(ftau, -2,2)

```

---

summary.gamlssinf0to1 *Summarizes an inflated GAMLSS fitted model*

---

**Description**

These are specific methods for the generic function `summary` which summarize objects returned by `gamlssinf0to1` or `gamlssZadj`.

**Usage**

```

## S3 method for class 'gamlssinf0to1'
summary(object, type = c("vcov", "qr"),
        robust=FALSE, save = FALSE, hessian.fun = c("R", "PB"),
        digits = max(3, getOption("digits") - 3),...)

## S3 method for class 'gamlssZadj'
summary(object, type = c("vcov", "qr"),
        robust=FALSE, save = FALSE, hessian.fun = c("R", "PB"),
        digits = max(3, getOption("digits") - 3),...)

```

**Arguments**

<code>object</code>	a <code>gamlssinf0to1</code> or <code>gamlssZadj</code> fitted model
<code>type</code>	the default value <code>vcov</code> uses the <code>vcov()</code> method for <code>gamlss</code> to get the variance-covariance matrix of the estimated beta coefficients, see details below. The alternative <code>qr</code> is the original method used in <code>gamlss</code> to estimate the standard errors but it is not reliable since it does not take into account the inter-correlation between the distributional parameters <code>mu</code> , <code>sigma</code> , <code>nu</code> and <code>tau</code> , while the inflation parameters <code>xi0</code> and <code>xi1</code> are uncorrelated anyway.
<code>robust</code>	whether robust (sandwich) standard errors are required
<code>save</code>	whether to save the environment of the function so to have access to its values
<code>hessian.fun</code>	whether when calculate the Hessian should use the "R" function <code>optimHess()</code> or a function based on Pinheiro and Bates <code>nlme</code> package, "PB".
<code>digits</code>	the number of digits in the output
<code>...</code>	for extra arguments

**Details**

Using the default value `type="vcov"`, the `vcov()` method is used to get the variance covariance matrix (and consequently the standard errors) of the beta parameters. The variance covariance matrix is calculated using the inverse of the numerical second derivatives of the observed information matrix. This is a more reliable method since it take into the account the inter-correlation between the all the parameters. The `type="qr"` assumes that the parameters are fixed at the estimated values. Note that both methods are not appropriate and should be used with caution if smoothing terms are used in the fitting.

**Value**

Print summary of a `gamlssInf0to1` or a `gamlssZadj` object

**Author(s)**

Abu Hossain, Mikis Stasinopoulos <[mikis.stasinopoulos@gamlss.org](mailto:mikis.stasinopoulos@gamlss.org)>, Bob Rigby and Marco Enea

**References**

Houssain, A., Stasinopoulos, M., Rigby, R. and Enea, M. (2015). Centile estimation for a proportion response variable. Accepted for publication on *Statistics in Medicine*.

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Stasinopoulos D. M., Rigby R.A. and Akantziliotou C. (2003) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).

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, <http://www.jstatsoft.org/v23/i07>.

**See Also**

[gamlssInf0to1](#), [gamlssZadj](#)

**Examples**

```
# The gamlssZadj example
set.seed(3210)
x <- (runif(1000)*4)-2
data(sda)
fmu <- splinefun(sda$x, sda$mu)
fsigma <- splinefun(sda$x, sda$sigma)
fnu <- function(x)
  {f <- splinefun(sda$x, sda$nu)
  f(x)/6
}
```

```

y0 <- rZAGA(1000, mu=fmu(x), sigma=fsigma(x), nu=fnu(x))
da <- data.frame(y0,x)
t0p <- gamlssZadj(y=y0, mu.fo=~pb(x), sigma.fo=~pb(x),data=da,
                 trace=TRUE, xi0.fo=~pb(x), family="GA")
summary(t0p)

```

---

term.plotInf0to1	<i>Plot regression terms for a specified parameter of a fitted gamlss-Inf0to1 object</i>
------------------	--

---

### Description

This is a wrapper to function [term.plot](#). `term.plotInf0to1` produces term plots for a specified parameter from a `gamlssinf0to1` object.

### Usage

```

term.plotInf0to1(object, parameter = c("mu", "sigma", "nu", "tau",
                                       "xi0", "xi1"),...)

```

### Arguments

object	a <code>gamlssinf0to1</code> fitted model
parameter	which distribution (or inflation) parameter is required, default parameter="mu"
...	extra arguments, the same of <a href="#">term.plot</a> (except 'what')

### Details

see function [term.plot](#)

### Value

A plot of fitted terms.

### Author(s)

Marco Enea, Mikis Stasinopoulos, Bob Rigby and Abu Hossain

### References

Hossain, A., Stasinopoulos, M., Rigby, R. and Enea, M. (2015). Centile estimation for a proportion response variable. *Statistics in Medicine*, doi: 10.1002/sim.6748.

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.

Stasinopoulos D. M., Rigby R.A. and Akantziotou C. (2003) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).

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, <http://www.jstatsoft.org/v23/i07>.

### See Also

[gamlssInf0to1](#)

### Examples

```
gen.Family("SST", "logit")
gen.Inf0to1("logitSST", "Zero")
set.seed(10)
Y <- rlogitSSTInf0(500, mu=0.5, sigma=0.7, nu=0.5, tau=5, xi0=0.5, log=FALSE)
dat <- data.frame(Y)
dat$x <- rnorm(500)
m1 <- gamlssInf0to1(y=Y, mu.formula=~x, sigma.formula=~x,
                  nu.formula=~x, tau.formula=~x,
                  xi0.formula=~x, data=dat, family=logitSST)

# term.plot for the mu parameter
term.plot(m1$dist, parameter="mu")
term.plotInf0to1(m1, parameter="mu", col.shaded = 2)

# term.plot for the binomial parameter
term.plot(m1$multinom, parameter="mu")
term.plotInf0to1(m1, parameter="xi0", col.shaded = 2)
```

---

term.plotZadj	<i>Plot regression terms for a specified parameter of a fitted gamlssZadj object</i>
---------------	--

---

### Description

\ This is a wrapper to function [term.plot](#). `codeterm.plotZadj` produces term plots for a specified parameter from a `gamlssZadj` object.

### Usage

```
term.plotZadj(object, parameter = c("mu", "sigma", "nu", "tau", "xi0"), ...)
```

**Arguments**

object            a gamlssZadj fitted model  
parameter        which distribution (or inflation) parameter is required, default parameter="mu"  
...                extra arguments, the same of [term.plot](#) (except 'what')

**Details**

see function [term.plot](#)

**Value**

A plot of fitted terms.

**Author(s)**

Marco Enea, Mikis Stasinopoulos, Bob Rigby and Abu Hossain

**References**

Hossain, A., Stasinopoulos, M., Rigby, R. and Enea, M. (2015). Centile estimation for a proportion response variable. *Statistics in Medicine*, doi: 10.1002/sim.6748.

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.

Stasinopoulos D. M., Rigby R.A. and Akantziliotou C. (2003) Instructions on how to use the GAMLSS package in R. Accompanying documentation in the current GAMLSS help files, (see also <http://www.gamlss.org/>).

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, <http://www.jstatsoft.org/v23/i07>.

**See Also**

[gamlssZadj](#)

**Examples**

```
set.seed(3210)
x <- (runif(1000)*4)-2
data(sda)
fmu <- splinefun(sda$x, sda$mu)
curve(fmu, -2,2)
fsigma <- splinefun(sda$x, sda$sigma)
curve(fsigma, -2,2)
fnu <- function(x)
  {f <- splinefun(sda$x, sda$nu)
  f(x)/6
}
```

```
curve(fnu, -2,2)
set.seed(321)
y0 <- rZAGA(1000, mu=fmu(x), sigma=fsigma(x), nu=fnu(x))
da <- data.frame(y0,x)
g0p <- gamlss(y0~pb(x), sigma.fo=~pb(x), nu.fo=~pb(x), data=da, family=ZAGA)
t0p <- gamlssZadj(y=y0, mu.fo=~pb(x), sigma.fo=~pb(x),data=da,
                 trace=TRUE, xi0.fo=~pb(x), family="GA")

# term.plot for the mu parameter
term.plot(g0p);title("gamlss")
term.plot(t0p$dist,"mu");title("gamlssZadj")
term.plotZadj(t0p,"mu",col.shaded = 3);title("gamlssZadj")

# term.plot for the sigma parameter
term.plot(g0p, "sigma");title("gamlss")
term.plot(t0p$dist,"sigma");title("gamlssZadj")
term.plotZadj(t0p,"sigma",col.shaded = 3);title("gamlssZadj")

# term.plot for the binomial parameter
term.plot(g0p, "nu");title("gamlss")
term.plot(t0p$binom,"mu");title("gamlssZadj")
term.plotZadj(t0p,"xi0",col.shaded = 3);title("gamlssZadj")
```



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