

# The gamlss packages

Mikis Stasinopoulos<sup>1</sup>   Bob Rigby<sup>1</sup>

<sup>1</sup>STORM, London Metropolitan University

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# The R packages

- `gamlss` the original package
- `gamlss.dist` all `gamlss.family` distributions
- `gamlss.data` different sets of data
- `gamlss.demo` demos for distributions and smoothing
- `gamlss.nl` non-linear term fitting
- `gamlss.tr` generating truncated distributions
- `gamlss.cens` for censored (left, right or interval) response variables
- `gamlss.mx` finite mixtures distributions and random effects
- `gamlss.add` for extra additive terms
- `gamlss.util` for extra utilities

The GAMLSS framework packages can be downloaded and installed from CRAN, the R library at <http://www.r-project.org/>. Test versions may be found at the GAMLSS web site at <http://www.gamlss.com/>.

# The gamlss package

## Classification of the available functions

- Fitting or Updating a Model
- Extracting Information from the Fitted Model
- Selecting a Model
- Plotting and Diagnostics
- Centile Estimation

# Fitting or Updating a Model

`gamlss()` for fitting and creating a `gamlss` object

`refit()` to refit a `gamlss` object (i.e. continue iterations)

`update()` to update a given `gamlss` model object

`gamlssML()` fitting a parametric distribution to a single (response) variable

`histDist()` to fit and plot a parametric distribution

`fitDist()` select a parametric distribution from an appropriate list

## Extracting Information from the Fitted Model

- `GAIC()` generalised Akaike information criterion (or AIC)
- `coef()` the linear coefficients
- `deviance()` the global deviance  $-2 \log L$
- `fitted()` the fitted values for a distribution parameter
- `predict()` to predict from new data individual distribution parameter values
- `predictAll()` to predict from new data all the distribution parameter values
- `print()` : to print a gamlss object
- `residuals()` to extract the normalised (randomised) quantile residuals
- `summary()` to summarise the fit in a gamlss object
- `vcov()` to extract the variance-covariance matrix of the <sup>beta</sup> estimates.

## Selecting a Model

`add1()` `drop1()` to add or drop a single term

`find.hyper()` to find the hyper-parameters

`stepGAIC()` to select explanatory terms in one parameter

`stepGAICAll.A()` to select explanatory terms in all the parameters (strategy A)

`stepGAICAll.B()` to select explanatory terms in all the parameters (strategy B)

`stepTGD()` selecting variables using a test set the global deviance for new (test) data set given a fitted gamlss model.

# Diagnostics

`plot()` a plot of four graphs for the normalized (randomized) quantile residuals

`pdf.plot()` for plotting the pdf functions for a given fitted `gamlss` object or a given `gamlss.family` distribution

`Q.stats()` for printing and plotting the Q statistics of Royston and Wright (2000).

`rqres.plot()` for plotting QQ-plots of different realisations of randomised residuals (for discrete distributions)

`wp()` worm plot of the residuals from a fitted `gamlss` object

`dtop()` detrended Own's plot of the residuals



## Centile estimation

`centiles()` to plot centile curves against an x-variable.

`centiles.com()` to compare centiles curves for more than one object.

`centiles.split()` as for `centiles()`, but splits the plot at specified values of x.

`centiles.pred()` to predict and plot centile curves for new x-values.

`centiles.fan()` fan plot of centile curves

`fitted.plot()` to plot fitted values for all the parameters against an x-variable

## Other useful functions

`prof.dev()` the profile global deviance of one of the distribution parameters

`prof.term()` for plotting the profile global deviance of one of the model (beta) parameters

`show.link()` for showing available link functions

`term.plot()` for plotting additive (smoothing) terms

## The gamlss() function

```
gamlss( formula = ~1, sigma.formula}= ~1,  
        nu.formula = ~1, tau.formula = ~1,  
        family = NO(),  
        data = sys.parent(), weights = NULL,  
        contrasts = NULL, method = RS(), start.from = NULL,  
        mu.start = NULL, sigma.start = NULL,  
        nu.start = NULL, tau.start = NULL,  
        mu.fix = FALSE, sigma.fix = FALSE, nu.fix = FALSE,  
        tau.fix = FALSE, control = gamlss.control(...),  
        i.control = glim.control(...), ...)
```

## Arguments of the gamlss() function

```
formula = y ~ x1 + x3
sigma.fo = ~ x1
nu.fo = ~ x2
tau.fo = ~ 1
data = abdom
family = LO
weights = freq
method = mixed(10,50)
control = gamlss.control(trace=FALSE)
```

## Starting values

Generally are not needed:

```
start.mu= 2 or start.mu=fitted(m1, "mu")
```

The same applies for other parameters

```
start.sigma , start.nu, start.tau
```

[Starting from a previous model](#)

```
start.from= model1
```

## Demonstration

```
library(gamlss)
data(abdom)
plot(y~x, col="blue", xlab="age", ylab="circumference",
     data=abdom)

# fitting polynomials
abd0<-gamlss(y~poly(x,3), family=N0, data=abdom)
abd00<-gamlss(y~x+I(x^2)+I(x^3), family=N0, data=abdom)
abd00
summary(abd00)

# fitting P-splines
abd1<-gamlss(y~pb(x), family=N0, data=abdom)
abd1
summary(abd1)
```

## Demonstration (continue)

```
# Fitting a model for sigma
abd2 <- gamlss(y~pb(x),sigma.formula=~pb(x),family=NO,
              data=abdom)
plot(abd2)

# term plot with partial residuals
term.plot(abd2, what="mu", se=TRUE, partial=TRUE)
term.plot(abd2, what="sigma", se=TRUE, partial=TRUE)

# worm plots
wp(abd2)
wp(abd2,ylim.all=1.5)
```

## Demonstration (continue)

```
# fitting cs()
abd3 <- gamlss(y ~ cs(x), sigma.formula = ~cs(x),
              data = abdom, family = NO)

# fitting loess
abd4 <- gamlss(y~lo(x,span=.4),sigma.formula=~lo(x,span=.4)
              data=abdom, family=NO)

# fitting different distribution
abd5 <- gamlss(y~pb(x),sigma.formula=~pb(x), data=abdom,
              family=TF)
GAIC(abd1,abd2,abd3,abd4, abd5)
GAIC(abd1,abd2,abd3,abd4, abd5, k=3 )
```



## Demonstration (continue)

```
# plotting a continuous distribution
PPP <- par(mfrow=c(2,2))
plot(function(y) dGA(y, mu=10 ,sigma=0.3), 0.1, 25)
plot(function(y) pGA(y, mu=10 ,sigma=0.3), 0.1, 25)
plot(function(y) qGA(y, mu=10 ,sigma=0.3), 0, 1)
hist(rGA(100,mu=10,sigma=.3))
par(PPP)
```

## Demonstration (continue)

```
PPP <- par(mfrow=c(2,2))
plot(function(y) dNBI(y, mu = 10, sigma =0.5 ),
      from=0, to=40, n=40+1, type="h",
      main="pdf", ylab="pdf(x)")
cdf <- stepfun(0:39, c(0, pNBI(0:39, mu=10, sigma=0.5 ))
              , f = 0)
plot(cdf,main="cdf", ylab="cdf(x)", do.points=FALSE )
invcdf <-stepfun(seq(0.01,.99,length=39), qNBI(seq(0.01,.99)
        length=40), mu=10, sigma=0.5 ), f = 0)
plot(invcdf, main="inverse cdf",ylab="inv-cdf(x)",
     do.points=FALSE )
tN <- table(Ni <- rNBI(1000,mu=5, sigma=0.5))
r <- barplot(tN, col='lightblue')
```

# END

for more information see

[www.gamlss.org](http://www.gamlss.org)