

# Beyond mean modelling: GAMLSS models

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Material available at:

[https://github.com/mfasiolo/workshop\\_WARSAW19](https://github.com/mfasiolo/workshop_WARSAW19)

These slides cover:

- 1 What are GAMLSS models?
- 2 When are they useful?
- 3 GAMLSS modelling in mgcv and mgcViz

# Intro to GAMLSS models

Recall GAM model structure:

$$y|\mathbf{x} \sim \text{Distr}\{y|\theta_1 = \mu(\mathbf{x}), \theta_2, \dots, \theta_p\},$$

where

$$\mu(\mathbf{x}) = g^{-1}\left\{\sum_{j=1}^m f_j(\mathbf{x})\right\},$$

and  $g$  is the link function.

Example, Scaled Student-t distribution:

- location  $\mu(\mathbf{x}) = \mathbb{E}(y|\mathbf{x})$
- scale  $\theta_2 = \sigma$
- shape  $\theta_3 = \nu$

# Intro to GAMLSS models

In Generalized Additive Models for Location Scale and Shape (GAMLSS) (Rigby and Stasinopoulos, 2005) we let scale and shape change with the covariates  $\mathbf{x}$ .

GAMLSS model structure:

$$y|\mathbf{x} \sim \text{Distr}\{y|\theta_1 = \mu_1(\mathbf{x}), \theta_2 = \mu_2(\mathbf{x}), \dots, \theta_p = \mu_p(\mathbf{x})\},$$

where

$$\begin{aligned}\mu_1(\mathbf{x}) &= g_1^{-1}\left\{\sum_{j=1}^m f_j^1(\mathbf{x})\right\}, \\ &\dots \\ \mu_p(\mathbf{x}) &= g_p^{-1}\left\{\sum_{j=1}^m f_j^p(\mathbf{x})\right\},\end{aligned}$$

and  $g_1, \dots, g_p$  are link function.

Example: **Gaussian model for location and scale**

Model is

$$y|\mathbf{x} \sim N\{y|\mu(\mathbf{x}), \sigma(\mathbf{x})\}$$

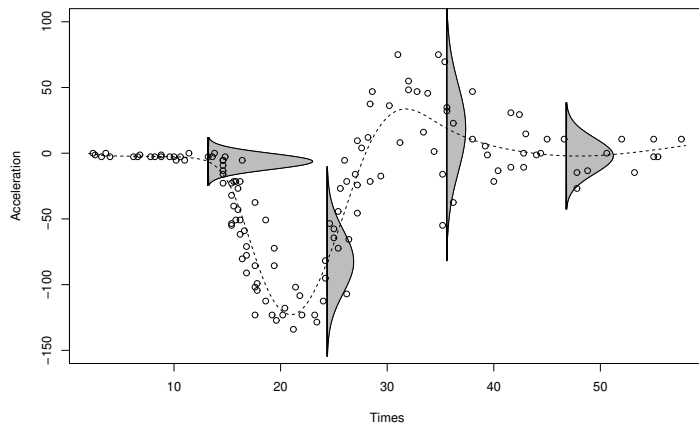
where

$$\mathbb{E}(y|\mathbf{x}) = \mu(\mathbf{x}) = \sum_{j=1}^m f_j^1(\mathbf{x})$$

$$\text{var}(y|\mathbf{x})^{1/2} = \sigma(\mathbf{x}) = \exp \left\{ \sum_{j=1}^m f_j^2(\mathbf{x}) \right\}$$

that is  $g_2 = \log$  to guarantee  $\sigma > 0$ .

# Intro to GAMLSS models



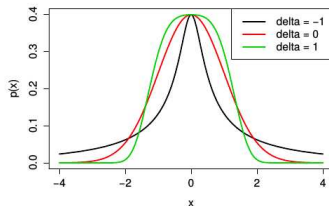
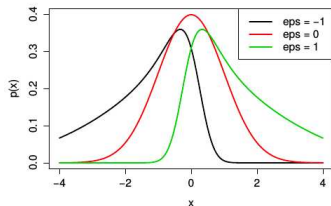
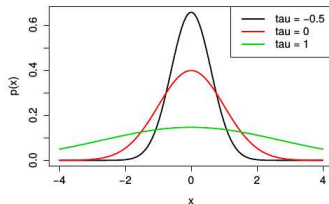
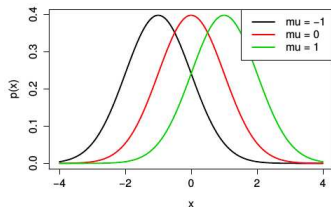
**Figure:** Gaussian model with variable mean and variance.

In `mgcv`: `gam(list(y~s(x), ~s(x)), family=gaulss)`.

# Intro to GAMLSS models

## Example: **Sinh-arcsinh (shash)** distribution

Four parameter distribution where location, scale, skewness (asymmetry) and kurtosis (tail behaviour) can depend on  $x$  (Jones and Pewsey, 2009).



# Intro to GAMLSS models

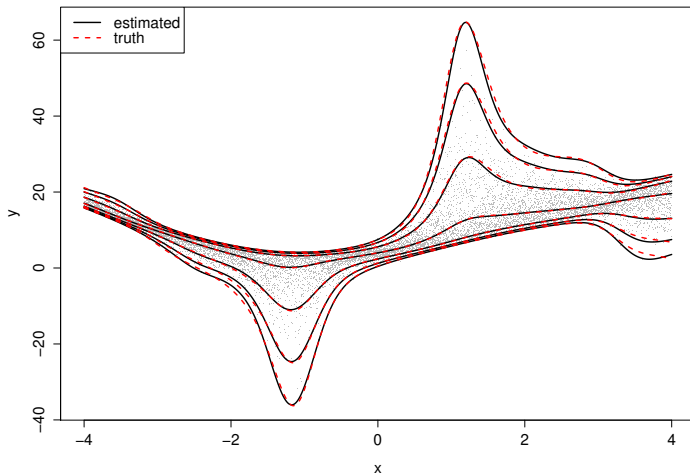


Figure: `gam(list(y~s(x), ~s(x), ~s(x), ~s(x)), family=shash).`



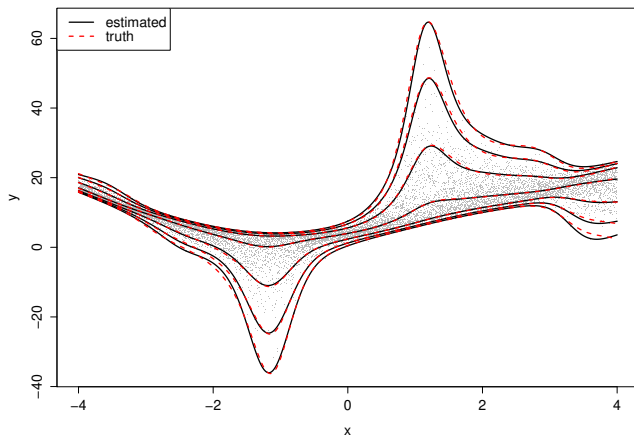
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# Intro to GAMLSS models

## When are they useful?

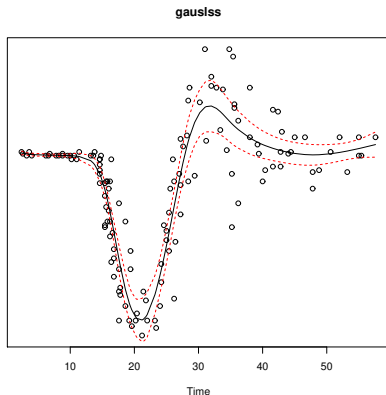
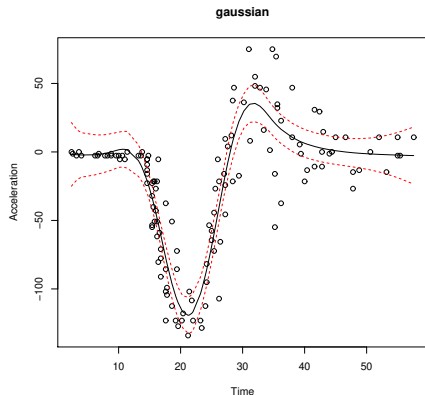
R1: you might be interested in whole distribution  $y|\mathbf{x}$  not just  $\mathbb{E}(y|\mathbf{x})$ .



# Intro to GAMLSS models

## When are they useful?

R2: standard GAM inference (e.g. p-value & confidence interval) is valid if the model for  $p(y|\mathbf{x})$  is correct

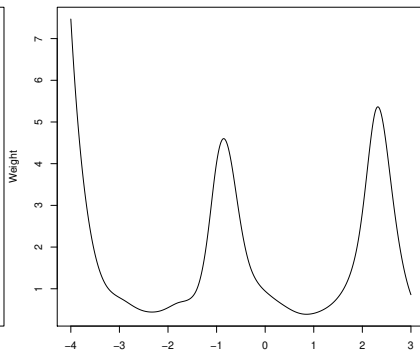
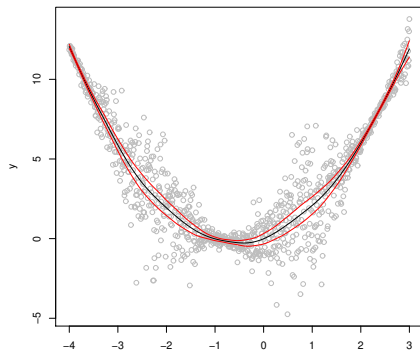


# Intro to GAMLSS models

## When are they useful?

R3: the accuracy of the fit is improved if the weight of each observation is inversely proportional to  $\text{Var}(y|\mathbf{x})$ , e.g. consider

$$\hat{\beta} = \underset{\beta}{\operatorname{argmin}} \sum_{i=1}^n \frac{(y_i - \mathbf{x}_i^T \beta)^2}{\sigma(\mathbf{x}_i)^2}.$$



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# References I

- Jones, M. and A. Pewsey (2009). Sinh-arcsinh distributions. *Biometrika* 96(4), 761–780.
- Rigby, R. A. and D. M. Stasinopoulos (2005). Generalized additive models for location, scale and shape. *Journal of the Royal Statistical Society: Series C (Applied Statistics)* 54(3), 507–554.