# Skew-Normal likelihood

### Parametrisation

The standardised Skew-Normal distribution is

$$f(z) = \frac{2}{\omega_{\alpha}} \phi \left( \frac{z - \xi_{\alpha}}{\omega_{\alpha}} \right) \Phi \left( \alpha \frac{z - \xi_{\alpha}}{\omega_{\alpha}} \right)$$

where  $\omega_{\alpha}$  and  $\xi_{\alpha}$  are so that the mean is zero and variance is one, and they depends both on the skewness parameter  $\alpha$ .

The skew-normal likelihood is defined as the density wrt y, where

$$z = (y - \eta)\sqrt{w\tau} \sim f(z)$$

and

 $\eta$ : is the the linear predictor

 $\tau$ : is the precision

w: is a fixed scale or weight, w > 0,

### Link-function

The mean equals the linear predictor

$$\mu = \eta$$

# Hyperparameters

The precision is represented as

$$\theta_1 = \log \tau$$

and the prior is defined on  $\theta_1$ .

The (standardised) skewness  $\gamma$ , is represented as

$$\gamma = 0.988(2\frac{\exp(\theta_2)}{1 + \exp(\theta_2)} - 1)$$

and the prior is defined on  $\theta_2$ . The standardised skewness depends on  $\alpha$  as

$$\gamma = \frac{4 - \pi}{2} \frac{\left(\delta\sqrt{2/\pi}\right)^3}{\left(1 - 2\delta^2/\pi\right)^{3/2}}, \qquad \delta = \frac{\alpha}{\sqrt{1 + \alpha^2}}$$

# **Specification**

- family = sn
- Required arguments: y and w (keyword scale, and w = 1 by default).

#### Hyperparameter specification and default values

doc The Skew-Normal likelihoood hyper theta1 hyperid 74001 name log precision short.name prec initial 4 fixed FALSE prior loggamma **param** 1 5e-05 to.theta function(x) log(x) from.theta function(x) exp(x) theta2 hyperid 74002 name logit skew short.name skew initial 0.00123456789 fixed FALSE **prior** pc.sn param 10 to.theta function(x, skew.max = 0.988) log((1 + x / skew.max) / (1 - x / skew.max))from.theta function(x, skew.max = 0.988) skew.max \* (2 \* exp(x) / (1 + exp(x)) - 1)status experimental survival FALSE discrete FALSE link default identity pdf snExample library(sn) set.seed(246) n = 300x = rnorm(n, sd = 1)eta = 1+xskewness = 0.25y = numeric(n)prec <- 100 for(i in 1:n) { ## map moments to sn-parameters c(xi, omega, alpha)

param = INLA:::inla.sn.reparam(moments = c(eta[i], 1/prec, skewness)) y[i] = rsn(1, xi=param\$xi, omega = param\$omega, alpha = param\$alpha)

## Notes

- This implementation replaces older ones ("sn" and "sn2") from 16<sup>th</sup> September 2020.
- A N(a,0) prior is interpreted as a constant prior with density equal to one.