## Gaussian

## Parametrisation

The Gaussian distribution is

$$f(y) = \frac{\sqrt{w\tau}}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}w\tau (y - \mu)^2\right)$$

for continuously responses y where

 $\mu$ : is the mean

 $\tau$ : is the precision

w: is a fixed weight, w > 0.

## Link-function

The mean and variance of y are given as

$$\mu$$
 and  $\sigma^2 = \frac{1}{w\tau}$ 

and the mean is linked to the linear predictor by

$$\mu = \eta$$

## Hyperparameters

The precision is represented as

$$\theta = \log \tau$$

and the prior is defined on  $\theta$ .

## **Specification**

- $\bullet$  family = gaussian
- Required arguments: y and w (keyword weights)

The weights has default value 1.

## Hyperparameter spesification and default values

# hyper

theta

name log precision

 ${f short.name}$  prec

initial 4

fixed FALSE

prior loggamma

**param** 1 5e-05

to.theta function(x) log(x)

from.theta function(x) exp(x)

survival FALSE

discrete FALSE

# Example

In the following example we estimate the parameters in a simulated example with Gaussian responses, giving  $\tau$  a Gamma-prior with parameters (1, 0.01) and initial value (for the optimisations) of exp(2.0).

```
n=100
a = 1
b = 1
z = rnorm(n)
eta = a + b*z
tau = 100
scale = exp(rnorm(n))
prec = scale*tau
y = rnorm(n, mean = eta, sd = 1/sqrt(prec))
data = list(y=y, z=z)
formula = y ~ 1+z
result = inla(formula, family = "gaussian", data = data,
        control.data = list(hyper = list(
                                     prec = list(
                                             prior = "loggamma",
                                             param = c(1.0,0.01),
                                             initial = 2))),
              scale=scale, keep=TRUE)
summary(result)
```

#### Notes

None.