# Gaussian

### Parametrisation

The Gaussian distribution is

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

for continuously responses y where

 $\mu$ : is the mean

 $\tau$ : is the precision

s: is a fixed scaling, s > 0.

### **Link-function**

The mean and variance of y are given as

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

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**

- family = gaussian
- Required arguments: y and s (keyword scale)

The scalings have default value 1.

# Hyperparameter spesification and default values

### hyper

### theta

name log precision short.name prec initial 4 fixed FALSE prior loggamma param 1 5e-05 to.theta from.theta

```
survival FALSEdiscrete FALSElink default identitypdf gaussian
```

### 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.