## The Gamma-distribution

### Parametrisation

The Gamma-distribution has the following density

$$\pi(y) = \frac{b^a}{\Gamma(a)} y^{a-1} \exp(-by), \qquad a > 0, \quad b > 0, \quad y > 0,$$

where  $E(y) = \mu = a/b$  and  $Var(y) = 1/\tau = a/b^2$ , where  $\tau$  is the precision and  $\mu$  is the mean. We will use the following parameterisation for the precision

$$\tau = (s\phi)/\mu^2$$

where  $\phi$  is the precision parameter (or  $1/\phi$  is the dispersion parameter) and s > 0 is a fixed scaling, which gives this density

$$\pi(y) = \frac{1}{\Gamma(s\phi)} \left(\frac{(s\phi)}{\mu}\right)^{(s\phi)} y^{(s\phi)-1} \exp\left(-(s\phi)\frac{y}{\mu}\right)$$

#### **Link-function**

The linear predictor  $\eta$  is linked to the mean  $\mu$  using a default log-link

$$\mu = \exp(\eta)$$

# Hyperparameter

The hyperparameter is the precision parameter  $\phi$ , which is represented as

$$\phi = \exp(\theta)$$

and the prior is defined on  $\theta$ .

## **Specification**

- family = gamma for regression models and family = gamma.surv for survival models.
- Required arguments: for gamma.surv, y (to be given in a format by using inla.surv()), and for gamma, y and s.

The scalings have default value 1.

## Hyperparameter spesification and default values

doc The Gamma likelihood

hyper

theta

hyperid 58001 name precision parameter short.name prec initial 4.60517018598809 fixed FALSE

```
prior loggamma
         param 1 0.01
         to.theta function(x) log(x)
         from.theta function(x) exp(x)
survival FALSE
discrete FALSE
link default log quantile
pdf gamma
doc The Gamma likelihood (survival)
hyper
     theta
         hyperid 58101
         name precision parameter
         short.name prec
         initial 0
         fixed FALSE
         prior loggamma
         param 1 0.01
         to.theta function(x) log(x)
         from.theta function(x) exp(x)
survival TRUE
discrete FALSE
link default log quantile
pdf gammasurv
Example
In the following example we estimate the parameters in a simulated example.
n = 1000
x = rnorm(n)
eta = 1 + x
mu = exp(eta)
prec.scale = runif(n, min = 0.5, max = 2)
prec.par = 1.2
a = prec.par * prec.scale
b = mu / (prec.par * prec.scale)
y = rgamma(n, shape = a, scale = b)
r = inla(y ~1 + x, data = data.frame(y, x),
        scale = prec.scale, family = "gamma")
```

## Notes

None.