

Generalized Poisson

The generalized Poisson distribution is given by

$$f(y|\lambda, w) = \frac{\lambda(\lambda + wy)^{y-1}}{y!} \exp(-(\lambda + wy))$$

for $y = 0, 1, 2, \dots$ and where $\lambda > 0$ and $\max(-1, -\lambda/4) \leq w \leq 1$. The mean and variance of y are

$$\mu = \lambda(1 - w)^{-1} \quad \text{and} \quad \sigma^2 = \lambda(1 - w)^{-3} = \mu(1 - w)^{-2}.$$

Since the dispersion parameter w influence the mean as well as the variance, we will use the following parameterisation (ADD REFERENCES)

$$w = \frac{\varphi\mu^{p-1}}{1 + \varphi\mu^{p-1}},$$

for a fixed p , which gives the following density

$$f(y|\mu, \varphi, p) = \frac{\mu(\mu + \varphi\mu^{p-1}y)^{y-1}}{(1 + \varphi\mu^{p-1})^y y!} \exp\left(-\frac{\mu + \varphi\mu^{p-1}y}{1 + \varphi\mu^{p-1}}\right)$$

for $y = 0, 1, 2, \dots$. We assume $\varphi \geq 0$.

Link-function

The mean and variance of y are given as

$$E(y|.) = \mu \quad \text{and} \quad \text{Var}(y|.) = \mu(1 + \varphi\mu^{p-1})^2$$

and the mean is linked to the linear predictor by

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

Hyperparameters

The overdispersion parameter $\varphi \geq 0$ is represented as

$$\varphi = \exp(\theta)$$

The prior is defined on θ .

Specification

- `family="gpoisson"`
- `control.family = list(gpoisson.p = <p>)` defines the fixed parameter p (default 1).

Hyperparameter spesification and default values

hyper

theta

name overdispersion

short.name phi

initial -3

```

    fixed FALSE
    prior loggamma
    param 1 1
    to.theta function(x) log(x)
    from.theta function(x) exp(x)

survival FALSE

discrete TRUE

link default log

pdf gpoisson

status experimental

```

Example

In the following example we estimate the parameters in a simulated example with generalized Poisson responses.

```

dgpoisson = function(y, mu, phi, p)
{
  a = mu + phi * mu^(p-1.0) * y;
  b = 1. + phi * mu^(p-1.0);
  d = exp(log(mu) + (y-1.0)*log(a) -
          y*log(b) - lfactorial(y) -a/b)
  return (d)
}

rgpoisson = function(n, mu, phi, p)
{
  stopifnot(length(mu) == 1)
  s = sqrt(mu*(1+phi*mu^(p-1)))
  f = 10
  low = as.integer(max(0, mu - f*s))
  high = as.integer(mu + f*s)

  prob = dgpoisson(low:high, mu, phi, p)
  y = sample(low:high, n, replace=TRUE,
             prob = prob)

  return (y)
}

n = 1000
phi = 0
mu = 5
p = 1
y = rgpoisson(n, mu, phi, p)

r = inla(y ~ 1, data = data.frame(y),
        family = "gpoisson",
        control.family = list(gpoisson.p = p))

```