## Zero-inflated models: Poisson and Binomial

# Parametrisation

There is support two types of zero-inflated models, which we name type 0 and type 1. These are defined for both the Binomial, the Poisson and the negative Binomial likelihood. For simplicity we will describe only the Poisson as the other two cases are similar.

### Type 0

The (type 0) likelihood is defined as

$$Prob(y \mid \ldots) = p \times 1_{[y=0]} + (1-p) \times Poisson(y \mid y > 0)$$

where p is a hyperparameter where

$$p = \frac{\exp(\theta)}{1 + \exp(\theta)}$$

and  $\theta$  is the internal representation of p; meaning that the initial value and prior is given for  $\theta$ . This is model is called zeroinflatedpoisson0 (and zeroinflatedbinomial0).

## Type 1

The (type 1) likelihood is defined as

$$Prob(y \mid \ldots) = p \times 1_{[y=0]} + (1-p) \times Poisson(y)$$

where p is a hyperparameter where

$$p = \frac{\exp(\theta)}{1 + \exp(\theta)}$$

and  $\theta$  is the internal representation of p; meaning that the initial value and prior is given for  $\theta$ . This is model is called zeroinflatedpoisson1 (and zeroinflatedbinomial1).

### **Link-function**

As for the Poisson, the Binomial and the negative Binomial.

### Hyperparameters

For Poisson and the Binomial, there is one hyperparameter; where

$$p = \frac{\exp(\theta)}{1 + \exp(\theta)}$$

and the prior and initial value is is given for  $\theta$ .

For the negative Binomial, there are two hyperparameters. The overdispersion parameter n is represented as

$$\theta_1 = \log(n)$$

and the prior is defined on  $\theta_1$ . The zero-inflation parameter p, is represented as

$$p = \frac{\exp(\theta_2)}{1 + \exp(\theta_2)}$$

and the prior and initial value is is given for  $\theta_2$ .

# Specification

```
    family = zeroinflatedbinomial0
    family = zeroinflatedbinomial1
    family = zeroinflatednbinomial0
    family = zeroinflatednbinomial1
    family = zeroinflatedpoisson0
```

• family = zeroinflatedpoisson1

• Required arguments: As for the Binomial, the negative Binomial and Poisson likelihood.

## Hyperparameter spesification and default values

## Zeroinflated Binomial Type 0

**param** -1 0.2

survival FALSE

discrete FALSE

```
hyper
```

```
theta
          name logit probability
          short.name prob
         initial -1
         fixed FALSE
         prior gaussian
         param -1 0.2
          to.theta function(x) log(x/(1-x))
         from.theta function(x) \exp(x)/(1+\exp(x))
survival FALSE
discrete FALSE
Zeroinflated Binomial Type 1
hyper
    theta
         name logit probability
         short.name prob
         initial -1
         fixed FALSE
          prior gaussian
```

to.theta function(x) log(x/(1-x))

from.theta function(x)  $\exp(x)/(1+\exp(x))$ 

# Zeroinflated NegBinomial Type 0 hyper theta1 name log size short.name size initial 2.30258509299405 fixed FALSE prior loggamma param 11 to.theta function(x) log(x) from.theta function(x) exp(x) theta2 name logit probability short.name prob initial -1 fixed FALSE prior gaussian **param** -1 0.2 to.theta function(x) log(x/(1-x))from.theta function(x) $\exp(x)/(1+\exp(x))$ survival FALSE discrete FALSE Zeroinflated NegBinomial Type 1 hyper theta1 name log size short.name size initial 2.30258509299405 fixed FALSE prior loggamma param 11 to.theta function(x) log(x) from.theta function(x) exp(x) theta2

name logit probability
short.name prob

to.theta function(x) log(x/(1-x))

initial -1fixed FALSEprior gaussianparam -1 0.2

```
from.theta function(x) \exp(x)/(1+\exp(x))
survival FALSE
discrete FALSE
Zeroinflated Poisson Type 0
hyper
    theta
         name logit probability
         short.name prob
         initial -1
         fixed FALSE
         prior gaussian
         param -1 0.2
         to.theta function(x) log(x/(1-x))
         from.theta function(x) \exp(x)/(1+\exp(x))
survival FALSE
discrete FALSE
Zeroinflated Poisson Type 1
hyper
    theta
         name logit probability
         short.name prob
         initial -1
         fixed FALSE
         prior gaussian
         param -1 0.2
         to.theta function(x) log(x/(1-x))
         from.theta function(x) \exp(x)/(1+\exp(x))
survival FALSE
discrete FALSE
```

## Example

In the following example we estimate the parameters in a simulated example for both type 0 and type 1.

```
Poisson
## type 0
n=100
a = 1
b = 1
z = rnorm(n)
eta = a + b*z
p = 0.2
E = sample(c(1,5,10,15), size=n, replace=TRUE)
lambda = E*exp(eta)
## first sample y|y>0
y = rpois(n, lambda = lambda)
is.zero = (y == 0)
while(sum(is.zero) > 0)
    y[is.zero] = rpois(sum(is.zero), lambda[is.zero])
    is.zero = (y == 0)
## then set some of these to zero
y[rbinom(n, size=1, prob=p) == 1] = 0
data = list(y=y,z=z)
formula = y ~ 1+z
result0 = inla(formula, family = "zeroinflatedpoisson0", data = data, E=E)
summary(result0)
## type 1
y = rpois(n, lambda = lambda)
y[ rbinom(n, size=1, prob=p) == 1 ] = 0
data = list(y=y,z=z)
formula = y \sim 1+z
result1 = inla(formula, family = "zeroinflatedpoisson1", data = data, E=E)
summary(result1)
Binomial
## type 0
n=100
a = 1
b = 1
z = rnorm(n)
eta = a + b*z
p = 0.2
Ntrials = sample(c(1,5,10,15), size=n, replace=TRUE)
prob = exp(eta)/(1 + exp(eta))
y = rbinom(n, size = Ntrials, prob = prob)
is.zero = (y == 0)
```

while(sum(is.zero) > 0)

{

```
y[is.zero] = rbinom(sum(is.zero), size = Ntrials[is.zero], prob = prob[is.zero])
is.zero = (y == 0)
}
y[ rbinom(n, size=1, prob=p) == 1 ] = 0
data = list(y=y,z=z)
formula = y ~ 1+z
result0 = inla(formula, family = "zeroinflatedbinomial0", data = data, Ntrials = Ntrials)
summary(result0)

## type 1
y = rbinom(n, size = Ntrials, prob = prob)
y[ rbinom(n, size=1, prob=p) == 1 ] = 0
data = list(y=y,z=z)
formula = y ~ 1+z
result1 = inla(formula, family = "zeroinflatedbinomial1", data = data, Ntrials=Ntrials)
summary(result1)
```

### Notes

None.

### Extentions

There are some extentions available which currently is only implemented for the cases where its needed/requested.

**Type 2** Is like Type 1 but where (for the Poisson)

$$p = 1 - \left(\frac{E \exp(x)}{1 + E \exp(x)}\right)^{\alpha}$$

where  $\alpha > 0$  is the hyperparameter instead of p (and  $E \exp(x)$  is the mean). Available as zeroinflatedpoisson2 and for negative binomial as zeroinflatedpoinson2.

The internal representation is  $\theta = \log(\alpha)$  and prior is defined on  $\log(\alpha)$ .

### Zeroinflated Poisson Type 2

# hyper

```
theta
    name logit probability
    short.name prob
    initial -1
    fixed FALSE
    prior gaussian
    param -1 0.2
    to.theta function(x) log(x/(1-x))
    from.theta function(x) exp(x)/(1+exp(x))
survival FALSE
discrete FALSE
```

# Zeroinflated Negative Binomial Type 2

# hyper

```
theta1
        name log size
        short.name size
        initial 2.30258509299405
         fixed FALSE
        prior loggamma
        param 11
        to.theta function(x) log(x)
        from.theta function(x) exp(x)
    theta2
        name log alpha
         short.name a
        initial 0.693147180559945
         fixed FALSE
        prior gaussian
         param 2 1
         to.theta function(x) log(x)
         from.theta function(x) exp(x)
survival FALSE
discrete FALSE
```