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 θ is the internal representation of p; meaning that the initial value and prior is given for θ . 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 θ is the internal representation of p; meaning that the initial value and prior is given for θ . 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 θ .

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 θ_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 θ_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

hyper

```
theta
```

```
name logit probability
short.name prob
initial -1
fixed FALSE
prior gaussian
param -1 0.2
to.theta
from.theta
```

```
survival FALSE
```

discrete FALSE

link default logit probit cloglog

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Zeroinflated Binomial Type 1

hyper

theta

```
name logit probability
short.name prob
initial -1
fixed FALSE
prior gaussian
param -1 0.2
to.theta
from.theta
```

survival FALSE

```
discrete FALSE
link default logit probit cloglog
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Zeroinflated NegBinomial Type 0
hyper
    theta1
         name log size
          short.name size
         initial 2.30258509299405
         fixed FALSE
          prior loggamma
         param 11
          to.theta
         from.theta
    theta2
         name logit probability
         short.name prob
         initial -1
         fixed FALSE
         prior gaussian
         param -1 0.2
          to.theta
         from.theta
survival FALSE
discrete FALSE
link default log
pdf zeroinflated
Zeroinflated NegBinomial Type 1
hyper
    theta1
          name log size
          short.name size
         initial 2.30258509299405
         fixed FALSE
          prior loggamma
         param 11
          to.theta
```

from.theta

```
theta2
         name logit probability
         short.name prob
         initial -1
         fixed FALSE
         prior gaussian
         param -1 0.2
          to.theta
          from.theta
survival FALSE
discrete FALSE
link default log
pdf zeroinflated
Zeroinflated Poisson Type 0
hyper
    theta
         name logit probability
         short.name prob
         initial -1
         fixed FALSE
         prior gaussian
         param -1 0.2
          to.theta
         from.theta
survival FALSE
discrete FALSE
link default log
pdf zeroinflated
Zeroinflated Poisson Type 1
hyper
    theta
         name logit probability
         short.name prob
         initial -1
         fixed FALSE
         prior gaussian
```

param -1 0.2

to.theta from.theta

```
survival FALSE
discrete FALSE
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```

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 ~ 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 for Poisson as zeroinflatedpoisson2, for binomial as zeroinflatedbinomial2 and for the negative binomial as zeroinflatednbinomial2.

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

Zeroinflated Poisson Type 2 hyper theta name log alpha short.name a initial 0.693147180559945 fixed FALSE prior gaussian param 0.693147180559945 1 to.theta from.theta survival FALSE discrete FALSE link default log pdf zeroinflated Zeroinflated Binomial Type 2 hyper theta name alpha short.name alpha initial -1 fixed FALSE **prior** gaussian **param** -1 0.2 to.theta from.theta survival FALSE discrete FALSE link default logit probit cloglog pdf zeroinflated Zeroinflated Negative Binomial Type 2 hyper theta1 name log size short.name size initial 2.30258509299405

fixed FALSE

```
prior loggamma
         \mathbf{param} \quad 1 \ 1
         to.theta
         from.theta
    theta2
         name log alpha
         short.name a
         initial 0.693147180559945
         fixed FALSE
         prior gaussian
         param 21
         to.theta
         from.theta
survival FALSE
```

discrete FALSE

link default log

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