Discrete generalized Pareto distribution

Parametrisation

The discrete generalized Pareto (GP) distribution with positive shape parameter has cumulative distribution function

$$F(y; \sigma, \xi) = 1 - \left(1 + \xi \frac{y+1}{\sigma}\right)^{-1/\xi}, \qquad y = 0, 1, 2, \ldots$$

for a discrete response y where

  ξ: is the tail parameter, $\xi > 0$

  σ: is the scale parameter, $\sigma > 0$

Link function

The linear predictor η controls the α quantile of the corresponding continous GP

$$P(y \leq q_\alpha) = \alpha$$

and $q_\alpha = \exp(\eta)$. The scaling σ, is then a function of (q_α, ξ), as

$$\sigma = \frac{\xi \exp(\eta)}{(1-\alpha)^{-\xi} - 1}$$

Hyperparameters

The GP model has one hyperparameter. The tail $\xi > 0$ is represented as

$$\xi = \xi_{\text{low}} + (\xi_{\text{high}} - \xi_{\text{low}}) \frac{\exp(\theta)}{1 + \exp(\theta)}$$

and the prior is defined on θ, with constant low and high values. The prior is FIXED to `pc.gevtail`, see `inla.doc("pc.gevtail")` for more info.

Specification

- `family="dgp"`

- Required arguments: y and the quantile α.

The quantile is given as `control.family=list(control.link=list(quantile=`α`))`.

Hyperparameter spesification and default values

doc `Discrete generalized Pareto likelihood`

hyper

    theta

        hyperid `101301`
        name `tail`
        short.name `xi`
        output.name `Tail parameter for the dgp observations`
        output.name.intern `Intern tail parameter for the dgp observations`
        initial `2`
        fixed `FALSE`
        prior `pc.gevtail`
        param `7 0 0.5`
        to.theta `function(x, interval = c(REPLACE.ME.low, REPLACE.ME.high)) log(-(interval`
        from.theta `function(x, interval = c(REPLACE.ME.low, REPLACE.ME.high)) interval[1]`

survival `FALSE`

discrete `TRUE`

link `default quantile`

pdf `dgp`

Example

```
F = function(y, sigma, xi) 1.0 - (1.0 + xi * (y+1)/sigma)^(-1/xi)
f = function(y, sigma, xi) F(y, sigma, xi) - F(y-1, sigma, xi)

rdgp = function(n, sigma, eta, alpha, xi = 0.001)
{
    if (missing(sigma)) {
        stopifnot(!missing(eta) && !missing(alpha))
        stopifnot(length(eta) == 1)
        sigma = exp(eta) * xi / ((1.0 - alpha)^(-xi) -1.0)
    }
    stopifnot(length(sigma) == 1)
    eps = 1e-7
    y.max = ceiling((eps^(-xi) -1) * sigma/xi)
    return (sample(0:y.max, prob = f(0:y.max, sigma, xi),
                   size=n, replace=TRUE))
}
```

```
n = 300
x = runif(n)-0.5
eta = 5+x
alpha = 0.95
xi = 0.3
y = numeric(n)
for(i in 1:n) {
    y[i] = rdgp(1, eta = eta[i], alpha = alpha, xi=xi)
}

r = inla(y ~ 1+x,
         data = data.frame(y, x),
         family = "dgp",
         control.family = list(
             control.link = list(quantile = alpha),
             hyper = list(tail = list(
                               prior = "pc.gevtail",
                               param = c(7, 0.0, 0.5)))),
         control.predictor = list(compute=TRUE),
         verbose=TRUE)

summary(r)
plot(r, plot.prior=TRUE)

dev.new()
plot(cbind(r$summary.fitted.values$mean, exp(eta)))
abline(a=0, b=1)
```

Notes