Spatial lag model for spatial effects

Parametrization

The slm model is defined as

$$\mathbf{x} = (I_n - \rho W)^{-1} (X\beta + \varepsilon)$$

where I_n is the identity matrix of dimension $n \times n$, W is a spatial weights matrix, X is a matrix of covariates, ρ is a spatial autocorrelation parameter, β are coefficients of the covariates and ε is zero mean Gaussian noise with precision τ .

Hyperparameters

This model has two hyperparameters $\theta = (\theta_1, \theta_2)$. The precision parameter τ is represented as

$$\theta_1 = \log \tau$$

and the prior is defined on θ_1 . The spatial autocorrelation parameter ρ is represented as

$$\rho^* = \frac{\rho - \rho_{\min}}{\rho_{\max} - \rho_{\min}}$$

and then

$$\theta_2 = \log(\rho^*/(1 - \rho^*))$$

and the prior is defined on θ_2 . Here, ρ_{\min} and ρ_{\max} are lower and upper limits of the legal range for ρ .

Specification

The slm model is specified inside the f() function as

args.slm is used to define the slm-spesific parameters in the model.

rho.min and rho.max define the range in which ρ can take values. Note that, ρ^* is in the interval (0,1) and that it is re-scaled to the interval (rho.min, rho.max) when computing $I_n - \rho W$. Initial values on ρ need to be re-scaled to the (0,1) interval.

X defines the matrix of covariates.

W defines the adjacency matrix.

Q.beta defines the precision of the vector of coefficients β in the model.

Hyperparameter spesification and default values

```
doc Spatial lag model
hyper
    theta1
         hyperid 34001
         name log precision
         short.name prec
         initial 4
         fixed FALSE
         prior loggamma
         param 1 5e-05
         to.theta function(x) log(x)
         from.theta function(x) exp(x)
    theta2
         hyperid 34002
         name rho
         short.name rho
         initial 0
         fixed FALSE
         prior normal
         param 0 10
         to.theta function(x) log(x/(1-x))
         from.theta function(x) 1/(1+exp(-x))
constr FALSE
nrow.ncol FALSE
augmented FALSE
aug.factor 1
aug.constr
n.div.by
n.required TRUE
set.default.values TRUE
\mathbf{pdf} slm
status experimental
```

Example

```
## Example using the Boston dataset from package spdep
require(INLA)
require(spdep)
data(boston)
## Index for the latent model
n <- nrow(boston.c)</pre>
boston.c$idx <- 1:n
## Define adjacency using a row-standardised matrix
lw <- nb2listw(boston.soi)</pre>
W <- as(as_dgRMatrix_listw(lw), "CsparseMatrix")</pre>
## Model definition
f1 <- log(CMEDV) ~ CRIM + ZN + INDUS + CHAS + I(NOX^2)+ I(RM^2) + AGE +
  log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT)
mmatrix <- model.matrix(f1, boston.c)</pre>
## Zero-variance for error term
zero.variance = list(prec=list(initial = 25, fixed=TRUE))
## Compute eigenvalues for SLM model, used to obtain rho.min and
## rho.max
e = eigenw(lw)
re.idx = which(abs(Im(e)) < 1e-6)
rho.max = 1/max(Re(e[re.idx]))
rho.min = 1/min(Re(e[re.idx]))
rho = mean(c(rho.min, rho.max))
## Precision matrix for beta coefficents' prior
betaprec <- .0001
Q.beta = Diagonal(n=ncol(mmatrix), betaprec)
## Priors on the hyperparameters
hyper = list(
        prec = list(
                prior = "loggamma",
                param = c(0.01, 0.01)),
        rho = list(
                initial=0,
                prior = "logitbeta",
                param = c(1,1))
## Fit model
slmm1 <- inla( log(CMEDV) ~ -1 +</pre>
              f(idx, model="slm",
                args.slm=list(
                         rho.min = rho.min,
                         rho.max = rho.max,
                         W=W,
                         X=mmatrix,
                         Q.beta=Q.beta),
                hyper=hyper),
               data=boston.c, family="gaussian",
```

Notes

The estimates of β are included at the end of the vector of random effects. See the example for details on how to extract them.