## Autoregressive model of order 1 (AR1)

#### **Parametrization**

The autoregressive model of order 1 (AR1) for the Gaussian vector  $\mathbf{x} = (x_1, \dots, x_n)$  is defined as:

$$x_1 \sim \mathcal{N}(0, (\tau(1-\phi^2))^{-1})$$
  
 $x_i = \phi \ x_{i-1} + \epsilon_i; \quad \epsilon_i \sim \mathcal{N}(0, \tau^{-1}) \quad i = 2, \dots, n$ 

where

$$|\phi| < 1$$

### Hyperparameters

The precision parameter  $\kappa$  is represented as

$$\theta_1 = \log(\kappa)$$

where  $\kappa$  is the marginal precision,

$$\kappa = \tau (1 - \phi^2).$$

The parameter  $\phi$  is represented as

$$\theta_2 = \log\left(\frac{1+\phi}{1-\phi}\right)$$

and the prior is defined on  $\theta = (\theta_1, \theta_2)$ .

### Specification

The AR1 model is specified inside the f() function as

The (optional) argument values is a numeric or factor vector giving the values assumed by the covariate for which we want the effect to be estimated. See the example for RW1 for an application.

#### Example

In this exaple we implement a ar1 model where  $\theta_1$  has a log-Gamma prior with parameters 1 and 0.001 and  $\theta_2$  has a Gaussian prior with parameters 0 and 0.001

```
#simulate data
n = 100
phi = 0.8
prec = 10
## note that the marginal precision would be
marg.prec = prec * (1-phi^2)

E=sample(c(5,4,10,12),size=n,replace=T)
eta = as.vector(arima.sim(list(order = c(1,0,0), ar = phi), n = n,sd=sqrt(1/prec)))
y=rpois(n,E*exp(eta))
data = list(y=y,z=1:n)

## fit the model
formula = y~f(z,model="ar1",prior=c("loggamma","gaussian"),param=c(1,0.001,0,0.001))
result = inla(formula,family="poisson", data = data)
```

# Notes

None