# Package 'spNNGP'

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Description	
rNNGP Function for fitting univariate Bayesian spatial regression models	
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R topics documented:	
NeedsCompilation yes	
Repository CRAN	
<pre>URL http://blue.for.msu.edu/software.html</pre>	
Encoding UTF-8	
License GPL (>= 2)	
<b>Description</b> Fits Gaussian univariate Bayesian spatial regression models using Nearest Neighbor Gaussian Processes (NNGP).	
<b>Depends</b> R (>= 1.8.0), coda, Formula	
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Version 0.1.0	
<b>Title</b> Spatial Regression Models using Nearest Neighbor Gaussian Processes	

Neighbor Gaussian Processes (NNGP).

The function rNNGP fits Gaussian univariate Bayesian spatial regression models using Nearest

rNNGP

### Usage

#### **Arguments**

formula a symbolic description of the regression model to be fit. See example below.

data an optional data frame containing the variables in the model. If not found in data,

the variables are taken from environment (formula), typically the environment

from which rNNGP is called.

coords an  $n \times 2$  matrix of the observation coordinates in  $R^2$  (e.g., easting and northing).

n.neighbors number of neighbors used in the NNGP

starting a list with each tag corresponding to a parameter name. Valid tags are beta,

sigma.sq, tau.sq, phi, and nu. The value portion of each tag is the parameter's

starting value.

tuning a list with each tag corresponding to a parameter name. Valid tags are sigma.sq,

tau.sq, phi, and nu. The value portion of each tag defines the variance of the

Metropolis sampler Normal proposal distribution.

priors a list with each tag corresponding to a parameter name. Valid tags are sigma.sq.ig,

tau.sq.ig, phi.unif. Variance parameters, simga.sq and tau.sq, are assumed to follow an inverse-Gamma distribution, whereas the spatial decay phi and smoothness nu parameters are assumed to follow Uniform distributions. The hyperparameters of the inverse-Gamma are passed as a vector of length two, with the first and second elements corresponding to the *shape* and *scale*, respectively. The hyperparameters of the Uniform are also passed as a vector of length two with the first and second elements corresponding to the lower and

upper support, respectively.

cov.model a quoted keyword that specifies the covariance function used to model the spatial

dependence structure among the observations. Supported covariance model key words are: "exponential", "matern", "spherical", and "gaussian". See

below for details.

n.samples the number of posterior samples to collect.

n.omp.threads a positive integer indicating the number of threads to use for SMP parallel pro-

cessing. The package must be compiled for OpenMP support. For most Intelbased machines, we recommend setting n.omp.threads to up to the number of

hyperthreaded cores.

verbose if TRUE, model specification and progress of the sampler is printed to the screen.

Otherwise, nothing is printed to the screen.

n.report the interval to report Metropolis sampler acceptance and MCMC progress.

... currently no additional arguments.

### **Details**

Model parameters can be fixed at their starting values by setting their tuning values to zero. The *no nugget* model is specified by setting tau.sq to zero in the starting and tuning lists.

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#### Value

```
An object of class rNNGP, which is a list comprising:
```

```
p.beta.samples a coda object of posterior samples for the regression coefficients.

p.theta.samples a coda object of posterior samples for covariance parameters.

run.time execution times for building the nearest neighbor index and MCMC sampler reported using proc.time().
```

The return object will include additional data used for subsequent prediction and/or model fit evaluation.

## Author(s)

```
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```

#### References

Datta, A., S. Banerjee, A.O. Finley, and A.E. Gelfand. (2016) Hierarchical Nearest-Neighbor Gaussian process models for large geostatistical datasets. Journal of the American Statistical Association, 111:800-812.

## **Examples**

```
## Not run:
rmvn <- function(n, mu=0, V = matrix(1)){</pre>
  p <- length(mu)</pre>
  if(any(is.na(match(dim(V),p))))
    stop("Dimension problem!")
  D <- chol(V)
  t(matrix(rnorm(n*p), ncol=p)
}
set.seed(1)
n <- 5000
coords <- cbind(runif(n,0,1), runif(n,0,1))</pre>
X <- as.matrix(cbind(1, rnorm(n)))</pre>
B \leftarrow as.matrix(c(1,5))
p <- length(B)</pre>
sigma.sq <- 2
tau.sq <- 1
phi <- 3/0.5
D <- as.matrix(dist(coords))</pre>
R <- exp(-phi*D)</pre>
```

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