# Package 'SpatialRegimes'

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Type Package

Title Spatial Constrained Clusterwise Regression

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**Description** A collection of functions for estimating spatial regimes, aggregations of neighboring spatial units that are homogeneous in functional terms. The term spatial regime, therefore, should not be understood as a synonym for cluster. More precisely, the term cluster does not presuppose any functional relationship between the variables considered, while the term regime is linked to a regressive relationship underlying the spatial process.

Depends spdep, quantreg, GWmodel, plm, spatialreg

License GPL (>= 2)

Suggests R.rsp, utils

VignetteBuilder R.rsp

NeedsCompilation no

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# Contents

Index

Awsreg				•													•					2	2
Sareg .																						3	3
SimData				•																		5	5
SkaterF				•			•		•								•	•				6	5
																						9	)

Awsreg

# Description

This function implements a spatial clusterwise regression based on the procedure suggested by Andreano *et al.* (2017) and Bille' *et al.* (2017).

# Usage

Awsreg(data, coly,colx,kernel,kernel2,coords,bw,tau,niter,conv,eta,numout,sout)

# Arguments

data	A data.frame.
coly	The dependent variable in the c("y_ols") form.
colx	The covariates in the $c("x1", "x2")$ form.
kernel	Kernel function used to calculate distances between units (default is "bisquare", other values: "exponential", "gaussian", "tricube").
kernel2	Kernel function used to calculate distances between units in the second step (default is "gaussian", other values: "exponential").
coords	The coordinates in terms of longitude and latitude.
bw	The bandwidth parameter of the initial weights.
tau	The confidence test parameter of the difference between regression parameters.
niter	The maximum number of iterations.
conv	The smallest accepted difference between the weights in two successive itera- tions.
eta	The parameter that regulates which is the weight of the weights of the previous iteration in the moving average that updates the new weights.
numout	The minimum number of areal units accepted for each cluster.
sout	Minimum value of weights such as to be considered equal to zero. Parameter used essentially to control clusters consisting of too few areal units.

# Details

Author really thanks Bille' A.G. for her contribution to revising the original code.

# Value

A object of Awsreg class with:

groups Estimated clusters.

Sareg

#### Author(s)

R. Benedetti

#### References

Andreano, M.S., Benedetti, R., and Postiglione, P. (2017). "Spatial regimes in regional European growth: an iterated spatially weighted regression approach", Quality & Quantity. 51, 6, 2665-2684.

Bille', A.G., Benedetti, R., and Postiglione, P. (2017). "A two-step approach to account for unobserved spatial heterogeneity", Spatial Economic Analysis, 12, 4, 452-471.

# Examples

```
data(SimData)
SimData = SimData[1:50,]
coords = cbind(SimData$long, SimData$lat)
```

#### 

```
SimData$regimes = aws$groups
plot(lat~long,SimData,col=regimes,pch=16)
```

Sareg

Spatial clusterwise regression by a constrained version of the Simulated Annealing

#### Description

This function implements a spatial clusterwise regression based on the procedure suggested by Postiglione *et al.* (2013).

#### Usage

```
Sareg(data, coly,colx, cont, intemp, rho, niter, subit, ncl, bcont)
```

#### Arguments

data	A data.frame
coly	The dependent variable in the c("y_ols") form.
colx	The covariates in the c("x1", "x2") form.
cont	The contiguity matrix.
intemp	The initial temperature.
rho	The temperature decay rate parameter.
niter	The maximum number of iterations.
subit	The number of sub-iterations for each iteration.
ncl	The number of clusters.
bcont	A parameter that regulates the penalty of simulated annealing in non-contiguous configurations of the clusters.

#### Value

A object of Sareg class with:

#### Author(s)

R. Benedetti

#### References

Postiglione, P., Benedetti, R., and Andreano, M.S. (2013). "Using Constrained Optimization for the Identification of Convergence Clubs", Computational Economics, 42, 151-174.

# Examples

```
data(SimData)
SimData = SimData[1:50,]
coords = cbind(SimData$long, SimData$lat)
```

```
dmat <-gw.dist(coords,focus=0,p=2,theta=0,longlat=FALSE)
W <- matrix(0,nrow(dmat),ncol(dmat))
W[dmat < 0.2] <- 1</pre>
```

# SimData

SimData

Simulated data for estimating spatial regimes.

#### Description

Simulated production function like data for estimating spatial regimes; data has been generated for the paper "F. Vidoli, G. Pignataro, R. Benedetti, F. Pammolli, "*Spatially constrained cluster-wise regression: optimal territorial areas in Italian health care*", forthcoming.

#### Usage

data(SimData)

#### Format

SimData is a simulated dataset with 500 observations and 7 variables.

long Longitude

lat Latitude

A Land input

L Labour input

K Capital input

clu Real regime

**y\_ols** Production output

500 units (100 units for each of the 5 regimes) are generated and, for each unit, the longitude and latitude coordinates are randomly drawn by using two Uniform distributions from 0 to 50 and from -70 to 20, i.e. U(0,50) and U(-70,20), respectively. Consequently, we set the matrix of covariates which include the constant, A, L and K variables by drawing from U(1.5,4). For each regime, finally,

a different (in the coefficients) spatial function is set assuming a linear functional form. More in particular, we set 5 different vectors of parameters (including the intercept): beta1 = (13,0.5,0.3,0.2), beta2 = (11,0.8,0.1,0.1), beta3 = (9,0.3,0.2,0.5), beta4 = (7,0.4,0.3,0.3) and beta5 = (5,0.2,0.6,0.2) and a normally distributed error term in N(0,1).

#### Author(s)

Vidoli F.

#### References

F. Vidoli, G. Pignataro and R. Benedetti "*Identification of spatial regimes of the production function of Italian hospitals through spatially constrained cluster-wise regression*", Socio-Economic Planning Sciences (in press) https://doi.org/10.1016/j.seps.2022.101223

# Examples

data(SimData)

SkaterF	Spatial constrained clusterwise regression by Spatial 'K'luster Analy-
	sis by Tree Edge Removal

#### Description

This function implements a spatial constrained clusterwise regression based on the Skater procedure by Assuncao et al. (2002).

#### Usage

SkaterF(edges,data,coly,colx,ncuts,crit,method=1,ind\_col,lat,long,tau.ch)

#### Arguments

edges	A matrix with 2 colums with each row is an edge.
data	A data.frame with the informations over nodes.
coly	The dependent variable in the c("y_ols") form.
colx	The covariates in the c("x1", "x2") form.
ncuts	The number of cuts.
crit	A scalar or two dimensional vector with with criteria for groups. Examples: limits of group size or limits of population size. If scalar, is the minimum criteria for groups.
method	1 (default) for OLS, 2 for Quantile regression, 3 for logit
ind_col	Parameter still not used in this version.
lat	Parameter still not used in this version.
long	Parameter still not used in this version.
tau.ch	Chosen quantile (for method $= 2$ ).

6

#### SkaterF

#### Details

Author really thanks Renato M. Assuncao and Elias T. Krainski for their original code (skater, library spdep).

#### Value

A object of skaterF class with:

groups	A vector with length equal the number of nodes. Each position identifies the group of node.
edges.groups	A list of length equal the number of groups with each element is a set of edges
not.prune	A vector identifying the groups with are not candidates to partition.
candidates	A vector identifying the groups with are candidates to partition.
ssto	The total dissimilarity in each step of edge removal.

#### Author(s)

F. Vidoli

# References

For method = 1: F. Vidoli, G. Pignataro, and R. Benedetti. (2022) "Identification of spatial regimes of the production function of italian hospitals through spatially constrained cluster-wise regression. In: Socio-Economic Planning Sciences, page 101223, doi: https://doi.org/10.1016/j.seps.2022.101223

For method = 2: Vidoli, F., Sacchi A. & Sanchez Carrera E. (2025) "Spatial regimes in heterogeneous territories: The efficiency of local public spending" In: Economic modelling https://doi.org/10.1016/j.econmod.2025.10

# Examples

```
neighbours = tri2nb(coords, row.names = NULL)
bh.nb <- neighbours
lcosts <- nbcosts(bh.nb, SimData)
nb <- nb2listw(bh.nb, lcosts, style="B")
mst.bh <- mstree(nb,5)
edges1 = mst.bh[,1:2]
```

ncuts1 = 4 crit1 = 10 coly1 = c("y\_ols") colx1 = c("A","L","K")

# OLS

```
sk = SkaterF(edges = edges1,
              data= SimData,
              coly = coly1,
              colx= colx1,
              ncuts=ncuts1,
              crit=crit1,
              method=1)
SimData$regimes = sk$groups
# plot(lat~long,SimData,col=regimes,pch=16)
## quantile 0.8
# sk2 = SkaterF(edges = edges1,
#
              data= SimData,
#
              coly = coly1,
#
              colx= colx1,
#
              ncuts=ncuts1,
#
              crit=crit1,
#
              method=2,tau.ch=0.8)
#
# SimData$regimes_q = sk2$groups
# plot(lat~long,SimData,col=regimes_q,pch=16)
```

# Index

Awsreg, 2

Sareg, 3 SimData, 5 SkaterF, 6