# Package 'spdep'

September 2, 2013

**Version** 0.5-64

Date 2013-09-02

Title Spatial dependence: weighting schemes, statistics and models

Encoding latin1

Author Roger Bivand < Roger . Bivand@nhh . no>

Maintainer Roger Bivand < Roger . Bivand@nhh . no>

**Depends** R (>= 2.14.0), methods, sp (>= 0.9), Matrix (>= 0.999375-9)

Imports LearnBayes, deldir, boot, splines, coda, nlme, MASS

**Suggests** 

parallel, spam(>= 0.13-1), RANN, RColorBrewer, lattice, xtable, maptools (>= 0.5-4), foreign

Description A collection of functions to create spatial weights matrix objects from polygon contiguities, from point patterns by distance and tessellations, for summarizing these objects, and for permitting their use in spatial data analysis, including regional aggregation by minimum spanning tree; a collection of tests for spatial autocorrelation,including global Moran's I, APLE, Geary's C, Hubert/Mantel general cross product statistic, Empirical Bayes estimates and Assunção/Reis Index,Getis/Ord G and multicoloured join count statistics, local Moran's I and Getis/Ord G, saddlepoint approximations and exact tests for global and local Moran's I; and functions for estimating spatial simultaneous autoregressive (SAR) lag and error models, impact measures for lag models, weighted and unweighted SAR and CAR spatial regression models,semi-parametric and Moran eigenvector spatial filtering, GM SAR error models, and generalized spatial two stage least squares models.

License GPL (>= 2)

NeedsCompilation yes

Repository CRAN

**Date/Publication** 2013-09-02 13:24:39

# R topics documented:

afcon			
aggregate.nb			5
airdist			6
anova.sarlm			7
aple			8
aple.mc			9
aple.plot			11
as_dgRMatrix_listw			12
auckland			13
autocov_dist			14
paltimore			16
phicv			18
poston			18
pptest.sarlm			20
eard	•		22
pell2nb	•		23
choynowski	•		24
columbus	•	• •	25
liffnb	•	• •	27
Inearneigh	•	• •	28
lo_ldet	•	• •	29
droplinks	•		36
EBest	•	• •	37
EBImoran.mc	•		39
EBlocal	•		41
edit.nb	•		42
eigenw	•	• •	43
eire	•	• •	46
elect80			
errorsarlm			
geary			
geary.mc			
geary.test			
getisord			
globalG.test			
GMerrorsar			61
Graph Components		• •	
graphneigh	•		66
gtapinieign	•	• •	68
	•	• •	71
•	•		71
	•		73
nuddersfield	•		
mpacts			74
nclude.self	•		78
nvIrM	•		79
s.symmetric.nb			81

• •	~
J	82
<b>3</b>	84
<b>3</b>	86
knearneigh	87
knn2nb	89
lag.listw	90
lagmess	91
e	94
S .	99
lm.LMtests	01
lm.morantest	
lm.morantest.exact	
lm.morantest.sad	
localG	
localmoran	
localmoran.exact	13
localmoran.sad	15
LR.sarlm	18
mat2listw	
MCMCsamp	
ME	
moran	
moran.mc	
moran.plot	
moran.test	
mstree	
nb.set.operations	34
nb2blocknb	35
nb2INLA	36
nb2lines	
nb2listw	
nb2mat	
nb2WB	
nbcosts	
nbdists	
nblag	
nc.sids	
NY_data	47
oldcol	48
p.adjustSP	<b>5</b> 0
plot.mst	51
plot.nb	
plot.skater	
poly2nb	
± •	
predict.sarlm	
probmap	
prunecost	
prunemst	61

4 afcon

read.gal	2
read.gwt2nb	3
residuals.sarlm	5
Rotation	5
sacsarlm	7
set.spChkOption	1
similar.listw	2
skater	4
sp.correlogram	7
sp.mantel.mc	)
SpatialFiltering	1
spautolm	3
spdep	
spweights.constants	
ssw	l
stsls	
subset.listw	
subset.nb	
summary.nb	
summary.sarlm	
tolerance.nb	
tri2nb	
trW	
used.cars	
wheat	
write.nb.gal	5
208	8
	•

afcon

Index

Spatial patterns of conflict in Africa 1966-78

# Description

The afcon data frame has 42 rows and 5 columns, for 42 African countries, exclusing then South West Africa and Spanish Equatorial Africa and Spanish Sahara. The dataset is used in Anselin (1995), and downloaded from before adaptation. The neighbour list object africa.rook.nb is the SpaceStat 'rook.GAL', but is not the list used in Anselin (1995) - paper.nb reconstructs the list used in the paper, with inserted links between Mauritania and Morocco, South Africa and Angola and Zambia, Tanzania and Zaire, and Botswana and Zambia. afxy is the coordinate matrix for the centroids of the countries.

# Usage

data(afcon)

aggregate.nb 5

### **Format**

This data frame contains the following columns:

x an easting in decimal degrees (taken as centroid of shapefile polygon)

y an northing in decimal degrees (taken as centroid of shapefile polygon)

totcon index of total conflict 1966-78

name country name

id country id number as in paper

#### Note

All source data files prepared by Luc Anselin, Spatial Analysis Laboratory, Department of Agricultural and Consumer Economics, University of Illinois, Urbana-Champaign.

### **Source**

Anselin, L. and John O'Loughlin. 1992. Geography of international conflict and cooperation: spatial dependence and regional context in Africa. In The New Geopolitics, ed. M. Ward, pp. 39-75. Philadelphia, PA: Gordon and Breach. also: Anselin, L. 1995. Local indicators of spatial association, Geographical Analysis, 27, Table 1, p. 103.

#### **Examples**

```
data(afcon)
plot(africa.rook.nb, afxy)
plot(diffnb(paper.nb, africa.rook.nb), afxy, col="red", add=TRUE)
text(afxy, labels=attr(africa.rook.nb, "region.id"), pos=4, offset=0.4)
moran.test(afcon$totcon, nb2listw(africa.rook.nb))
moran.test(afcon$totcon, nb2listw(paper.nb))
geary.test(afcon$totcon, nb2listw(paper.nb))
```

aggregate.nb

Aggregate a spatial neighbours object

### **Description**

The method aggregates a spatial neighbours object, creating a new object listing the neighbours of the aggregates.

```
## S3 method for class 'nb'
aggregate(x, IDs, remove.self = TRUE, ...)
```

6 airdist

### Arguments

x an nb neighbour object

IDs a character vector of IDs grouping the members of the neighbour object

remove.self default TRUE: remove self-neighbours resulting from aggregation

unused - arguments passed through

#### Value

an nb neighbour object

#### Note

Method suggested by Roberto Patuelli

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

# **Examples**

```
data(used.cars)
data(state)
cont_st <- match(attr(usa48.nb, "region.id"), state.abb)
cents <- as.matrix(as.data.frame(state.center))[cont_st,]
opar <- par(mfrow=c(2,1))
plot(usa48.nb, cents, xlim=c(-125, -65), ylim=c(25, 50))
IDs <- as.character(state.division[cont_st])
agg_cents <- aggregate(cents, list(IDs), mean)
agg_nb <- aggregate(usa48.nb, IDs)
plot(agg_nb, agg_cents[, 2:3], xlim=c(-125, -65), ylim=c(25, 50))
text(agg_cents[, 2:3], agg_cents[, 1], cex=0.6)
par(opar)</pre>
```

airdist

Measure distance from plot

# **Description**

Measure a distance between two points on a plot using locator; the function checks par("plt") and par("usr") to try to ensure that the aspect ratio y/x is 1, that is that the units of measurement in both x and y are equivalent.

```
airdist(ann=FALSE)
```

anova.sarlm 7

### Arguments

ann annotate the plot with line measured and distance

#### Value

a list with members:

distance measured

coords coordinates between which distance is measured

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no >

### See Also

locator

anova.sarlm

Comparison of simultaneous autoregressive models

### **Description**

One of a number of tools for comparing simultaneous autoregressive models, in particular nested models. The function is based on anova.lme() for comparing linear mixed models, and follows that function in using the "anova" generic name.

### Usage

```
## S3 method for class 'sarlm'
anova(object, ...)
```

# Arguments

object object is of class sarlm

... other objects of class sarlm or class lm

### **Details**

If successive models have different numbers of degrees of freedom, a likelihood ratio test will be performed between them. It is important to recall that tests apply to nested models, and this function at least attempts to make sure that the response variable in the models being compared has the same name. Useless results can still be generated when incomparable models are compared, it being the responsibility of the user to check.

#### Value

The function returns a data frame printed by default functions

8 aple

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

```
LR.sarlm.AIC
```

### **Examples**

```
example(columbus)
lm.mod <- lm(CRIME ~ HOVAL + INC, data=columbus)
lag <- lagsarlm(CRIME ~ HOVAL + INC, data=columbus, nb2listw(col.gal.nb))
mixed <- lagsarlm(CRIME ~ HOVAL + INC, data=columbus, nb2listw(col.gal.nb),
    type="mixed")
error <- errorsarlm(CRIME ~ HOVAL + INC, data=columbus, nb2listw(col.gal.nb))
LR.sarlm(mixed, error)
anova(lag, lm.mod)
anova(lag, error, mixed)
AIC(lag, error, mixed)</pre>
```

aple

Approximate profile-likelihood estimator (APLE)

### **Description**

The Approximate profile-likelihood estimator (APLE) of the simultaneous autoregressive model's spatial dependence parameter was introduced in Li et al. (2007). It employs a correction term using the eigenvalues of the spatial weights matrix, and consequently should not be used for large numbers of observations. It also requires that the variable has a mean of zero, and it is assumed that it has been detrended. The spatial weights object is assumed to be row-standardised, that is using default style="W" in nb2listw.

### Usage

```
aple(x, listw, override_similarity_check=FALSE, useTrace=TRUE)
```

# **Arguments**

x a zero-mean detrended continuous variable

listw a listw object from for example nb2listw

override\\_similarity\\_check

default FALSE, if TRUE - typically for row-standardised weights with asymmetric underlying general weights - similarity is not checked

useTrace default TRUE, use trace of sparse matrix W %\*% W (Li et al. (2010)), if FALSE, use crossproduct of eigenvalues of W as in Li et al. (2007)

aple.mc 9

### **Details**

This implementation has been checked with Hongfei Li's own implementation using her data; her help was very valuable.

#### Value

A scalar APLE value.

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Li, H, Calder, C. A. and Cressie N. A. C. (2007) Beyond Moran's I: testing for spatial dependence based on the spatial autoregressive model. Geographical Analysis 39, 357-375; Li, H, Calder, C. A. and Cressie N. A. C. (2012) One-step estimation of spatial dependence parameters: Properties and extensions of the APLE statistic, Journal of Multivariate Analysis 105, 68-84.

#### See Also

```
nb2listw, aple.mc, aple.plot
```

### **Examples**

```
example(wheat)
nbr1 <- poly2nb(wheat, queen=FALSE)
nbr1 <- nblag(nbr1, 2)
nbr12 <- nblag_cumul(nbr1)
cms0 <- with(as(wheat, "data.frame"), tapply(yield, c, median))
cms1 <- c(model.matrix(~ factor(c) -1, data=wheat) %*% cms0)
wheat$yield_detrend <- wheat$yield - cms1
isTRUE(all.equal(c(with(as(wheat, "data.frame"),
    tapply(yield_detrend, c, median))), rep(0.0, 25),
    check.attributes=FALSE))
moran.test(wheat$yield_detrend, nb2listw(nbr12, style="W"))
aple(as.vector(scale(wheat$yield_detrend, scale=FALSE)), nb2listw(nbr12, style="W"))
errorsarlm(yield_detrend ~ 1, wheat, nb2listw(nbr12, style="W"))</pre>
```

aple.mc

Approximate profile-likelihood estimator (APLE) permutation test

# **Description**

A permutation bootstrap test for the approximate profile-likelihood estimator (APLE).

```
aple.mc(x, listw, nsim, override_similarity_check=FALSE, useTrace=TRUE)
```

10 aple.mc

### Arguments

x a zero-mean detrended continuous variable
listw a listw object from for example nb2listw
nsim number of simulations

override\\_similarity\\_check

default FALSE, if TRUE - typically for row-standardised weights with asym-

metric underlying general weights - similarity is not checked

useTrace default TRUE, use trace of sparse matrix W %\*% W (Li et al. (2010)), if FALSE,

use crossproduct of eigenvalues of W as in Li et al. (2007)

#### Value

A boot object as returned by the boot function.

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Li, H, Calder, C. A. and Cressie N. A. C. (2007) Beyond Moran's I: testing for spatial dependence based on the spatial autoregressive model. Geographical Analysis 39, 357-375; Li, H, Calder, C. A. and Cressie N. A. C. (2012) One-step estimation of spatial dependence parameters: Properties and extensions of the APLE statistic, Journal of Multivariate Analysis 105, 68-84.

#### See Also

```
aple, boot
```

```
## Not run:
example(aple)
boot_out <- aple.mc(as.vector(scale(wheat$yield_detrend, scale=FALSE)),</pre>
nb2listw(nbr12, style="W"), nsim=500)
plot(boot_out)
boot_out
library(parallel)
cl <- makeCluster(detectCores())</pre>
set.ClusterOption(cl)
set.rlecuyerSeedOption(1:6)
boot_out <- aple.mc(as.vector(scale(wheat$yield_detrend, scale=FALSE)),</pre>
nb2listw(nbr12, style="W"), nsim=500)
boot_out
set.ClusterOption(NULL)
stopCluster(cl)
## End(Not run)
```

aple.plot

aple.plot	Approximate profile-likelihood estimator (APLE) scatterplot

# Description

A scatterplot decomposition of the approximate profile-likelihood estimator, and a local APLE based on the list of vectors returned by the scatterplot function.

# Usage

```
aple.plot(x, listw, override_similarity_check=FALSE, useTrace=TRUE, do.plot=TRUE, ...) localAple(x, listw, override_similarity_check=FALSE, useTrace=TRUE)
```

# **Arguments**

X	a zero-mean detrended continuous variable
listw	a listw object from for example nb2listw
override\_simi	larity\_check default FALSE, if TRUE - typically for row-standardised weights with asymmetric underlying general weights - similarity is not checked
useTrace	default TRUE, use trace of sparse matrix W %*% W (Li et al. (2010)), if FALSE, use crossproduct of eigenvalues of W as in Li et al. (2007)
do.plot	default TRUE: should a scatterplot be drawn
• • •	other arguments to be passed to plot

### **Details**

The function solves a secondary eigenproblem of size n internally, so constructing the values for the scatterplot is quite compute and memory intensive, and is not suitable for very large n.

# Value

aple.plot returns list with components:

```
X A vector as described in Li et al. (2007), p. 366.
Y A vector as described in Li et al. (2007), p. 367.
```

localAple returns a vector of local APLE values.

# Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

12 as\_dgRMatrix\_listw

### References

Li, H, Calder, C. A. and Cressie N. A. C. (2007) Beyond Moran's I: testing for spatial dependence based on the spatial autoregressive model. Geographical Analysis 39, pp. 357-375; Li, H, Calder, C. A. and Cressie N. A. C. (2012) One-step estimation of spatial dependence parameters: Properties and extensions of the APLE statistic, Journal of Multivariate Analysis 105, 68-84.

#### See Also

aple

### **Examples**

```
## Not run:
example(aple)
plt_out <- aple.plot(as.vector(scale(wheat$yield_detrend, scale=FALSE)),</pre>
nb2listw(nbr12, style="W"), cex=0.6)
crossprod(plt_out$Y, plt_out$X)/crossprod(plt_out$X)
lm_obj \leftarrow lm(Y \sim X, plt_out)
abline(lm_obj)
abline(v=0, h=0, lty=2)
zz <- summary(influence.measures(lm_obj))</pre>
infl <- as.integer(rownames(zz))</pre>
points(plt_out$X[inf1], plt_out$Y[inf1], pch=3, cex=0.6, col="red")
wheat$localAple <- localAple(as.vector(scale(wheat$yield_detrend, scale=FALSE)),</pre>
nb2listw(nbr12, style="W"))
mean(wheat$localAple)
hist(wheat$localAple)
spl <- list("sp.text", coordinates(wheat)[infl,], rep("*", length(infl)))</pre>
spplot(wheat, "localAple", sp.layout=spl)
## End(Not run)
```

as\_dgRMatrix\_listw

Interface between Matrix class objects and weights lists

### Description

Interface between Matrix class objects and weights lists

```
as_dgRMatrix_listw(listw)
as_dsTMatrix_listw(listw)
as_dsCMatrix_I(n)
as_dsCMatrix_IrW(W, rho)
Jacobian_W(W, rho)
```

auckland 13

### **Arguments**

listw a listw object created for example by nb2listw

W a dsTMatrix object created using as\_dsTMatrix\_listw from a symmetric listw

object

rho spatial regression coefficient

n length of diagonal for identity matrix

### Value

Matrix package class objects

### Author(s)

Roger Bivand

### **Examples**

auckland

Marshall's infant mortality in Auckland dataset

### **Description**

(Use example(auckland) to load the data from shapefile and generate neighbour list on the fly).

The auckland data frame has 167 rows (census area units — CAU) and 4 columns. The dataset also includes the "nb" object auckland.nb of neighbour relations based on contiguity, and the "polylist" object auckpolys of polygon boundaries for the CAU. The auckland data frame includes the following columns:

```
data(auckland)
```

14 autocov\_dist

### **Format**

This data frame contains the following columns:

Easting a numeric vector of x coordinates in an unknown spatial reference system

Northing a numeric vector of y coordinates in an unknown spatial reference system

M77\\_85 a numeric vector of counts of infant (under 5 years of age) deaths in Auckland, 1977-1985

Und5\\_81 a numeric vector of population under 5 years of age at the 1981 Census

#### **Details**

The contiguous neighbours object does not completely replicate results in the sources, and was reconstructed from auckpolys; examination of figures in the sources suggests that there are differences in detail, although probably not in substance.

#### **Source**

Marshall R M (1991) Mapping disease and mortality rates using Empirical Bayes Estimators, Applied Statistics, 40, 283–294; Bailey T, Gatrell A (1995) Interactive Spatial Data Analysis, Harlow: Longman — INFOMAP data set used with permission.

### **Examples**

```
require(maptools)
auckland <- readShapePoly(system.file("etc/shapes/auckland.shp",
   package="spdep")[1])
auckland.nb <- poly2nb(auckland)</pre>
```

autocov\_dist

Distance-weighted autocovariate

# Description

Calculates the autocovariate to be used in autonormal, autopoisson or autologistic regression. Three distance-weighting schemes are available.

```
autocov_dist(z, xy, nbs = 1, type = "inverse", zero.policy = NULL,
    style = "W", longlat=NULL)
```

autocov\_dist 15

### **Arguments**

Z	the response variable
Z	the response variable

xy a matrix of coordinates or a SpatialPoints object

nbs neighbourhood radius; default is 1

type the weighting scheme: "one" gives equal weight to all data points in the neigh-

bourhood; "inverse" (the default) weights by inverse distance; "inverse.squared"

weights by the square of "inverse"

zero.policy default NULL, use global option value; if FALSE stop with error for any empty

neighbour sets, if TRUE permit the weights list to be formed with zero-length

weights vectors

style style' can take values W, B, C, U, and S; W gives mean values for neighbours

longlat TRUE if point coordinates are longitude-latitude decimal, in which case dis-

tances are measured in kilometers; if xy is a SpatialPoints object, the value is

taken from the object itself

#### Value

A numeric vector of autocovariate values

#### Author(s)

Carsten F. Dormann and Roger Bivand

### References

Augustin N.H., Mugglestone M.A. and Buckland S.T. (1996) An autologistic model for the spatial distribution of wildlife. *Journal of Applied Ecology*, 33, 339-347; Gumpertz M.L., Graham J.M. and Ristaino J.B. (1997) Autologistic model of spatial pattern of Phytophthora epidemic in bell pepper: effects of soil variables on disease presence. *Journal of Agricultural, Biological and Environmental Statistics*, 2, 131-156.

# See Also

nb2listw

```
example(columbus)
xy <- cbind(columbus$X, columbus$Y)
ac1a <- autocov_dist(columbus$CRIME, xy, nbs=10, style="W",
    type="one")
acinva <- autocov_dist(columbus$CRIME, xy, nbs=10, style="W",
    type="inverse")
acinv2a <- autocov_dist(columbus$CRIME, xy, nbs=10, style="W",
    type="inverse.squared")

plot(ac1a ~ columbus$CRIME, pch=16, asp=1)
points(acinva ~ columbus$CRIME, pch=16, col="red")</pre>
```

16 baltimore

```
points(acinv2a ~ columbus$CRIME, pch=16, col="blue")
abline(0,1)
nb <- dnearneigh(xy, 0, 10)</pre>
lw <- nb2listw(nb, style="W")</pre>
ac1b <- lag(lw, columbus$CRIME)</pre>
all.equal(ac1b, ac1a)
nbd <- nbdists(nb, xy)</pre>
gl \leftarrow lapply(nbd, function(x) 1/x)
lw <- nb2listw(nb, glist=gl)</pre>
acinvb <- lag(lw, columbus$CRIME)</pre>
all.equal(acinvb, acinva)
gl2 <- lapply(nbd, function(x) 1/(x^2))
lw <- nb2listw(nb, glist=gl2)</pre>
acinv2b <- lag(lw, columbus$CRIME)</pre>
all.equal(acinv2b, acinv2a)
glm(CRIME ~ HOVAL + ac1b, data=columbus, family="gaussian")
spautolm(columbus$CRIME ~ HOVAL, data=columbus,
listw=nb2listw(nb, style="W"))
xy <- SpatialPoints(xy)</pre>
acinva <- autocov_dist(columbus$CRIME, xy, nbs=10, style="W",</pre>
type="inverse")
nb <- dnearneigh(xy, 0, 10)</pre>
nbd <- nbdists(nb, xy)</pre>
gl <- lapply(nbd, function(x) 1/x)
lw <- nb2listw(nb, glist=gl)</pre>
acinvb <- lag(lw, columbus$CRIME)</pre>
all.equal(acinvb, acinva)
```

baltimore

House sales prices, Baltimore, MD 1978

### **Description**

House sales price and characteristics for a spatial hedonic regression, Baltimore, MD 1978. X,Y on Maryland grid, projection type unknown.

# Usage

```
data(baltimore)
```

#### Format

A data frame with 211 observations on the following 17 variables.

baltimore 17

STATION a numeric vector

**PRICE** a numeric vector

NROOM a numeric vector

**DWELL** a numeric vector

**NBATH** a numeric vector

PATIO a numeric vector

FIREPL a numeric vector

AC a numeric vector

**BMENT** a numeric vector

NSTOR a numeric vector

GAR a numeric vector

AGE a numeric vector

CITCOU a numeric vector

LOTSZ a numeric vector

SQFT a numeric vector

X a numeric vector

Y a numeric vector

### **Source**

Prepared by Luc Anselin. Original data made available by Robin Dubin, Weatherhead School of Management, Case Western Research University, Cleveland, OH. http://sal.agecon.uiuc.edu/datasets/baltimore.zip

#### References

Dubin, Robin A. (1992). Spatial autocorrelation and neighborhood quality. Regional Science and Urban Economics 22(3), 433-452.

```
data(baltimore)
## maybe str(baltimore) ; plot(baltimore) ...
```

18 boston

bhicv

Data set with 4 life condition indices of Belo Horizonte region

# **Description**

The data are collected in the Atlas of condition indices published by the Joao Pinheiro Foundation and UNDP.

#### **Format**

A shape polygon object with seven variables:

id The identificator

Name Name of city

**Population** The population of city

**HLCI** Health Life Condition Index

**ELCI** Education Life Condition Index

**CLCI** Children Life Condition Index

**ELCI** Economic Life Condition Index

# **Examples**

### see example in 'skater' function help

boston

Corrected Boston Housing Data

# **Description**

The boston.c data frame has 506 rows and 20 columns. It contains the Harrison and Rubinfeld (1978) data corrected for a few minor errors and augmented with the latitude and longitude of the observations. Gilley and Pace also point out that MEDV is censored, in that median values at or over USD 50,000 are set to USD 50,000. The original data set without the corrections is also included in package mlbench as BostonHousing. In addition, a matrix of tract point coordinates projected to UTM zone 19 is included as boston.utm, and a sphere of influence neighbours list as boston.soi.

# Usage

data(boston)

boston 19

#### **Format**

This data frame contains the following columns:

**TOWN** a factor with levels given by town names

**TOWNNO** a numeric vector corresponding to TOWN

TRACT a numeric vector of tract ID numbers

LON a numeric vector of tract point longitudes in decimal degrees

LAT a numeric vector of tract point latitudes in decimal degrees

MEDV a numeric vector of median values of owner-occupied housing in USD 1000

CMEDV a numeric vector of corrected median values of owner-occupied housing in USD 1000

**CRIM** a numeric vector of per capita crime

**ZN** a numeric vector of proportions of residential land zoned for lots over 25000 sq. ft per town (constant for all Boston tracts)

**INDUS** a numeric vector of proportions of non-retail business acres per town (constant for all Boston tracts)

CHAS a factor with levels 1 if tract borders Charles River; 0 otherwise

**NOX** a numeric vector of nitric oxides concentration (parts per 10 million) per town

RM a numeric vector of average numbers of rooms per dwelling

AGE a numeric vector of proportions of owner-occupied units built prior to 1940

**DIS** a numeric vector of weighted distances to five Boston employment centres

**RAD** a numeric vector of an index of accessibility to radial highways per town (constant for all Boston tracts)

**TAX** a numeric vector full-value property-tax rate per USD 10,000 per town (constant for all Boston tracts)

**PTRATIO** a numeric vector of pupil-teacher ratios per town (constant for all Boston tracts)

**B** a numeric vector of 1000\*(Bk - 0.63)^2 where Bk is the proportion of blacks

LSTAT a numeric vector of percentage values of lower status population

### Source

http://lib.stat.cmu.edu/datasets/boston\_corrected.txt

#### References

Harrison, David, and Daniel L. Rubinfeld, Hedonic Housing Prices and the Demand for Clean Air, *Journal of Environmental Economics and Management*, Volume 5, (1978), 81-102. Original data.

Gilley, O.W., and R. Kelley Pace, On the Harrison and Rubinfeld Data, *Journal of Environmental Economics and Management*, 31 (1996), 403-405. Provided corrections and examined censoring.

Pace, R. Kelley, and O.W. Gilley, Using the Spatial Configuration of the Data to Improve Estimation, *Journal of the Real Estate Finance and Economics*, 14 (1997), 333-340.

20 bptest.sarlm

### **Examples**

```
hr0 \leftarrow lm(log(MEDV) \sim CRIM + ZN + INDUS + CHAS + I(NOX^2) + I(RM^2) +
AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT), data=boston.c)
summary(hr0)
logLik(hr0)
gp0 \leftarrow lm(log(CMEDV) \sim CRIM + ZN + INDUS + CHAS + I(NOX^2) + I(RM^2) +
 AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT), data=boston.c)
summary(gp0)
logLik(gp0)
lm.morantest(hr0, nb2listw(boston.soi))
## Not run: gp1 <- errorsarlm(log(CMEDV) ~ CRIM + ZN + INDUS + CHAS + I(NOX^2)
+ I(RM^2) + AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
 data=boston.c, nb2listw(boston.soi), method="Matrix",
 control=list(tol.opt = .Machine$double.eps^(1/4)))
summary(gp1)
gp2 <- lagsarlm(log(CMEDV) ~ CRIM + ZN + INDUS + CHAS + I(NOX^2) + I(RM^2)</pre>
+ AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
 data=boston.c, nb2listw(boston.soi), method="Matrix")
summary(gp2)
## End(Not run)
```

bptest.sarlm

Breusch-Pagan test for spatial models

# **Description**

Performs the Breusch-Pagan test for heteroskedasticity on the least squares fit of the spatial models taking the spatial coefficients rho or lambda into account. This function is a copy of the bptest function in package "lmtest", modified to use objects returned by spatial simultaneous autoregressive models.

### Usage

```
bptest.sarlm(object, varformula=NULL, studentize = TRUE, data=list())
```

# Arguments

object	An object of class "sarlm" from errorsarlm() or lagsarlm().
varformula	a formula describing only the potential explanatory variables for the variance (no dependent variable needed). By default the same explanatory variables are taken as in the main regression model
studentize	logical. If set to TRUE Koenker's studentized version of the test statistic will be used.
data	an optional data frame containing the variables in the varformula

bptest.sarlm 21

#### **Details**

Asymptotically this corresponds to the test given by Anselin (1988), but is not exactly the same. The studentized version is more conservative and perhaps to be prefered. The residuals, and for spatial error models the RHS variables, are adjusted for the spatial coefficient, as suggested bt Luc Anselin (personal communication).

It is also technically possible to make heteroskedasticity corrections to standard error estimates by using the "lm.target" component of sarlm objects - using functions in the lmtest and sandwich packages.

### Value

A list with class "htest" containing the following components:

statistic the value of the test statistic.
p.value the p-value of the test.

parameter degrees of freedom (wrongly reported if varformula given before 0.5-44.

method a character string indicating what type of test was performed.

#### Author(s)

Torsten Hothorn Torsten.Hothorn@rzmail.uni-erlangen.de> and Achim Zeileis <zeileis@ci.tuwien.ac.at>,
modified by Roger Bivand Roger.Bivand@nhh.no>

#### References

T.S. Breusch & A.R. Pagan (1979), A Simple Test for Heteroscedasticity and Random Coefficient Variation. *Econometrica* **47**, 1287–1294

W. Krämer & H. Sonnberger (1986), The Linear Regression Model under Test. Heidelberg: Physica.

L. Anselin (1988) Spatial econometrics: methods and models. Dordrecht: Kluwer, pp. 121-122.

#### See Also

```
errorsarlm, lagsarlm
```

```
example(columbus)
error.col <- errorsarlm(CRIME ~ HOVAL + INC, data=columbus,
    nb2listw(col.gal.nb))
bptest.sarlm(error.col)
bptest.sarlm(error.col, studentize=FALSE)
## Not run:
lm.target <- lm(error.col$tary ~ error.col$tarX - 1)
if (require(lmtest) && require(sandwich)) {
    coeftest(lm.target, vcov=vcovHC(lm.target, type="HCO"), df=Inf)
}
## End(Not run)</pre>
```

22 card

card

Cardinalities for neighbours lists

# Description

The function tallies the numbers of neighbours of regions in the neighbours list.

### Usage

```
card(nb)
```

# **Arguments**

nb

a neighbours list object of class nb

### **Details**

"nb" objects are stored as lists of integer vectors, where the vectors contain either the indices in the range 1:n for n as length(nb) of the neighbours of region i, or as.integer(0) to signal no neighbours. The function card(nb) is used to extract the numbers of neighbours from the "nb" object.

### Value

An integer vector of the numbers of neighbours of regions in the neighbours list.

### Author(s)

Roger Bivand < Roger . Bivand@nhh.no >

### References

Bivand R, Pebesma EJ, Gomez-Rubio V, (2008) *Applied Spatial Data Analysis with R*, Springer, New York, pp. 239-251; Bivand R, Portnov B, (2004) Exploring spatial data analysis techniques using R: the case of observations with no neighbours. In: Anselin L, Florax R, Rey S, (eds.), *Advances in Spatial Econometrics, Methodology, Tools and Applications*. Berlin: Springer-Verlag, pp. 121-142.

### See Also

```
summary.nb
```

```
example(columbus)
table(card(col.gal.nb))
```

cell2nb 23

cell2nb	Generate neighbours list for grid cells	

# Description

The function generates a list of neighbours for a grid of cells. Helper functions are used to convert to and from the vector indices for row and column grid positions, and rook (shared edge) or queen (shared edge or vertex) neighbour definitions are applied by type. If torus is TRUE, the grid is mapped onto a torus, removing edge effects.

# Usage

```
cell2nb(nrow, ncol, type="rook", torus=FALSE)
mrc2vi(rowcol, nrow, ncol)
rookcell(rowcol, nrow, ncol, torus=FALSE, rmin=1, cmin=1)
queencell(rowcol, nrow, ncol, torus=FALSE, rmin=1, cmin=1)
vi2mrc(i, nrow, ncol)
```

# Arguments

nrow	number of rows in the grid
ncol	number of columns in the grid
type	rook or queen
torus	map grid onto torus
rowcol	matrix with two columns of row, column indices
i	vector of vector indices corresponding to rowcol
rmin	lowest row index
cmin	lowset column index

# Value

The function returns an object of class nb with a list of integer vectors containing neighbour region number ids. See card for details of "nb" objects.

### Author(s)

```
Roger Bivand < Roger . Bivand@nhh.no>
```

# See Also

```
summary.nb, card
```

24 choynowski

### **Examples**

```
nb7rt <- cell2nb(7, 7)
summary(nb7rt)
xyc <- attr(nb7rt, "region.id")
xy <- matrix(as.integer(unlist(strsplit(xyc, ":"))), ncol=2, byrow=TRUE)
plot(nb7rt, xy)
nb7rt <- cell2nb(7, 7, torus=TRUE)
summary(nb7rt)</pre>
```

choynowski

Choynowski probability map values

### **Description**

Calculates Choynowski probability map values.

### Usage

```
choynowski(n, x, row.names=NULL, tol = .Machine$double.eps^0.5, legacy=FALSE)
```

### **Arguments**

n a numeric vector of counts of cases
x a numeric vector of populations at risk

row.names row names passed through to output data frame

tol accumulate values for observed counts >= expected until value less than tol legacy default FALSE using vectorised alternating side ppois version, if true use orig-

inal version written from sources and iterating down to tol

## Value

A data frame with columns:

pmap Poisson probability map values: probablility of getting a more "extreme" count

than actually observed, one-tailed with less than expected and more than ex-

pected folded together

type logical: TRUE if observed count less than expected

#### Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

### References

Choynowski, M (1959) Maps based on probabilities, Journal of the American Statistical Association, 54, 385–388; Cressie, N, Read, TRC (1985), Do sudden infant deaths come in clusters? Statistics and Decisions, Supplement Issue 2, 333–349; Bailey T, Gatrell A (1995) Interactive Spatial Data Analysis, Harlow: Longman, pp. 300–303.

columbus 25

### See Also

probmap

#### **Examples**

```
example(auckland)
res <- choynowski(auckland$M77_85, 9*auckland$Und5_81)
resl <- choynowski(auckland$M77_85, 9*auckland$Und5_81, legacy=TRUE)
all.equal(res, resl)
rt <- sum(auckland$M77_85)/sum(9*auckland$Und5_81)</pre>
ch_ppois_pmap <- numeric(length(auckland$Und5_81))</pre>
side <- c("greater", "less")</pre>
for (i in seq(along=ch_ppois_pmap)) {
 ch_ppois_pmap[i] <- poisson.test(auckland$M77_85[i], r=rt,</pre>
    T=(9*auckland$Und5_81[i]), alternative=side[(res$type[i]+1)])$p.value
}
all.equal(ch_ppois_pmap, res$pmap)
res1 <- probmap(auckland$M77_85, 9*auckland$Und5_81)
table(abs(res$pmap - res1$pmap) < 0.00001, res$type)
lt005 <- (res$pmap < 0.05) & (res$type)</pre>
ge005 <- (res$pmap < 0.05) & (!res$type)
cols <- rep("white", length(lt005))</pre>
cols[1t005] <- grey(2/7)
cols[ge005] <- grey(5/7)</pre>
plot(auckland, col=cols)
legend("bottomleft", fill=grey(c(2,5)/7), legend=c("low", "high"), bty="n")
```

columbus

Columbus OH spatial analysis data set

### **Description**

The columbus data frame has 49 rows and 22 columns. Unit of analysis: 49 neighbourhoods in Columbus, OH, 1980 data. In addition the data set includes a polylist object polys with the boundaries of the neighbourhoods, a matrix of polygon centroids coords, and col.gal.nb, the neighbours list from an original GAL-format file. The matrix bbs is DEPRECATED, but retained for other packages using this data set.

### Usage

```
data(columbus)
```

### **Format**

This data frame contains the following columns:

**AREA** computed by ArcView

**PERIMETER** computed by ArcView

26 columbus

```
COLUMBUS\_ internal polygon ID (ignore)
```

**COLUMBUS\\_I** another internal polygon ID (ignore)

POLYID yet another polygon ID

**NEIG** neighborhood id value (1-49); conforms to id value used in Spatial Econometrics book.

**HOVAL** housing value (in \\$1,000)

**INC** household income (in \\$1,000)

**CRIME** residential burglaries and vehicle thefts per thousand households in the neighborhood

**OPEN** open space in neighborhood

PLUMB percentage housing units without plumbing

**DISCBD** distance to CBD

**X** x coordinate (in arbitrary digitizing units, not polygon coordinates)

Y y coordinate (in arbitrary digitizing units, not polygon coordinates)

**NSA** north-south dummy (North=1)

**NSB** north-south dummy (North=1)

EW east-west dummy (East=1)

**CP** core-periphery dummy (Core=1)

**THOUS** constant=1,000

NEIGNO NEIG+1,000, alternative neighborhood id value

#### **Details**

The row names of columbus and the region.id attribute of polys are set to columbus\$NEIGNO.

# Note

All source data files prepared by Luc Anselin, Spatial Analysis Laboratory, Department of Agricultural and Consumer Economics, University of Illinois, Urbana-Champaign, http://sal.agecon.uiuc.edu/datasets/columbus.zip.

### Source

Anselin, Luc. 1988. Spatial econometrics: methods and models. Dordrecht: Kluwer Academic, Table 12.1 p. 189.

```
require(maptools)
columbus <- readShapePoly(system.file("etc/shapes/columbus.shp",
  package="spdep")[1])
col.gal.nb <- read.gal(system.file("etc/weights/columbus.gal",
  package="spdep")[1])</pre>
```

diffnb 27

diffnb

Differences between neighbours lists

# **Description**

The function finds differences between lists of neighbours, returning a nb neighbour list of those found

### Usage

```
diffnb(x, y, verbose=NULL)
```

### **Arguments**

x an object of class nby an object of class nb

verbose default NULL, use global option value; report regions ids taken from object

attribute "region.id" with differences

# Value

A neighbours list with class nb

### Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

```
example(columbus)
coords <- coordinates(columbus)
rn <- sapply(slot(columbus, "polygons"), function(x) slot(x, "ID"))
knn1 <- knearneigh(coords, 1)
knn2 <- knearneigh(coords, 2)
nb1 <- knn2nb(knn1, row.names=rn)
nb2 <- knn2nb(knn2, row.names=rn)
diffs <- diffnb(nb2, nb1)
plot(columbus, border="grey")
plot(nb1, coords, add=TRUE)
plot(diffs, coords, add=TRUE, col="red", lty=2)
title(main="Plot of first (black) and second (red)\nnearest neighbours")</pre>
```

28 dnearneigh

dnearneigh	Neighbourhood contiguity by distance	
------------	--------------------------------------	--

# **Description**

The function identifies neighbours of region points by Euclidean distance between lower (greater than) and upper (less than or equal to) bounds, or with longlat = TRUE, by Great Circle distance in kilometers.

# Usage

```
dnearneigh(x, d1, d2, row.names = NULL, longlat = NULL)
```

### **Arguments**

X	matrix of point coordinates or a SpatialPoints object
d1	lower distance bound
d2	upper distance bound
row.names	character vector of region ids to be added to the neighbours list as attribute region.id, default $seq(1, nrow(x))$
longlat	TRUE if point coordinates are longitude-latitude decimal degrees, in which case distances are measured in kilometers; if x is a SpatialPoints object, the value is

taken from the object itself

#### Value

The function returns a list of integer vectors giving the region id numbers for neighbours satisfying the distance criteria. See card for details of "nb" objects.

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

### See Also

knearneigh, card

```
example(columbus)
coords <- coordinates(columbus)
rn <- sapply(slot(columbus, "polygons"), function(x) slot(x, "ID"))
k1 <- knn2nb(knearneigh(coords))
all.linked <- max(unlist(nbdists(k1, coords)))
col.nb.0.all <- dnearneigh(coords, 0, all.linked, row.names=rn)
summary(col.nb.0.all, coords)
plot(columbus, border="grey")</pre>
```

```
plot(col.nb.0.all, coords, add=TRUE)
title(main=paste("Distance based neighbours 0-", format(all.linked),
 " distance units", sep=""))
data(state)
us48.fipsno <- read.geoda(system.file("etc/weights/us48.txt",</pre>
package="spdep")[1])
if (as.numeric(paste(version$major, version$minor, sep="")) < 19) {</pre>
m50.48 <- match(us48.fipsno$"State.name", state.name)</pre>
} else {
m50.48 <- match(us48.fipsno$"State_name", state.name)</pre>
xy <- as.matrix(as.data.frame(state.center))[m50.48,]</pre>
11k1 <- knn2nb(knearneigh(xy, k=1, longlat=FALSE))</pre>
all.linked <- max(unlist(nbdists(llk1, xy, longlat=FALSE)))</pre>
11.nb <- dnearneigh(xy, 0, all.linked, longlat=FALSE)</pre>
summary(ll.nb, xy, longlat=TRUE, scale=0.5)
gck1 <- knn2nb(knearneigh(xy, k=1, longlat=TRUE))</pre>
all.linked <- max(unlist(nbdists(gck1, xy, longlat=TRUE)))</pre>
gc.nb <- dnearneigh(xy, 0, all.linked, longlat=TRUE)</pre>
summary(gc.nb, xy, longlat=TRUE, scale=0.5)
plot(ll.nb, xy)
plot(diffnb(ll.nb, gc.nb), xy, add=TRUE, col="red", lty=2)
title(main="Differences between Euclidean and Great Circle neighbours")
xy1 <- SpatialPoints((as.data.frame(state.center))[m50.48,],</pre>
 proj4string=CRS("+proj=longlat"))
gck1a <- knn2nb(knearneigh(xy1, k=1))
all.linked <- max(unlist(nbdists(gck1a, xy1)))</pre>
gc.nb <- dnearneigh(xy1, 0, all.linked)</pre>
summary(gc.nb, xy1, scale=0.5)
```

do\_ldet

Spatial regression model Jacobian computations

### **Description**

These functions are made available in the package namespace for other developers, and are not intended for users. They provide a shared infrastructure for setting up data for Jacobian computation, and then for caclulating the Jacobian, either exactly or approximately, in maximum likelihood fitting of spatial regression models. The techniques used are the exact eigenvalue, Cholesky decompositions (Matrix, spam), and LU ones, with Chebyshev and Monte Carlo approximations; moments use the methods due to Martin and Smirnov/Anselin.

```
do_ldet(coef, env, which=1)
cheb_setup(env, q=5, which=1)
mcdet_setup(env, p=16, m=30, which=1)
eigen_setup(env, which=1)
```

```
eigen_pre_setup(env, pre_eig, which=1)
spam_setup(env, pivot="MMD", which=1)
spam_update_setup(env, in_coef=0.1, pivot="MMD", which=1)
Matrix_setup(env, Imult, super=as.logical(NA), which=1)
Matrix_J_setup(env, super=FALSE, which=1)
LU_setup(env, which=1)
LU_prepermutate_setup(env, coef=0.1, order=FALSE, which=1)
moments_setup(env, trs=NULL, m, p, type="MC", correct=TRUE, trunc=TRUE, eq7=TRUE, which=1)
SE_classic_setup(env, SE_method="LU", p=16, m=30, nrho=200, interpn=2000, interval=c(-1,0.999), SEIndet=NULL, which=1)
SE_whichMin_setup(env, SE_method="LU", p=16, m=30, nrho=200, interpn=2000, interval=c(-1,0.999), SEIndet=NULL, which=1)
SE_interp_setup(env, SE_method="LU", p=16, m=30, nrho=200, interval=c(-1,0.999), which=1)
```

#### **Arguments**

coef	spatial coefficient value
env	environment containing pre-computed objects, fixed after assignment in setup functions
which	default 1; if 2, use second listw object
pre_eig	pre-computed eigenvalues of length n
q	Chebyshev approximation order; default in calling spdep functions is 5, here it cannot be missing and does not have a default
р	Monte Carlo approximation number of random normal variables; default calling spdep functions is 16, here it cannot be missing and does not have a default
m	Monte Carlo approximation number of series terms; default in calling spdep functions is 30, here it cannot be missing and does not have a default; m serves the same purpose in the moments method
pivot	default "MMD", may also be "RCM" for Cholesky decompisition using spam
in_coef	fill-in initiation coefficient value, default 0.1
Imult	see Cholesky; numeric scalar which defaults to zero. The matrix that is decomposed is A+m*I where m is the value of Imult and I is the identity matrix of order ncol(A). Default in calling spdep functions is 2, here it cannot be missing and does not have a default, but is rescaled for binary weights matrices in proportion to the maximim row sum in those calling functions
super	see Cholesky; logical scalar indicating is a supernodal decomposition should be created. The alternative is a simplicial decomposition. Default in calling spdep functions is FALSE for "Matrix_J" and as.logical(NA) for "Matrix". Setting it to NA leaves the choice to a CHOLMOD-internal heuristic
order	default FALSE; used in LU_prepermutate, note warnings given for 1u method
trs	A numeric vector of m traces, as from trW
type	moments trace type, see trW
correct	default TRUE: use Smirnov correction term, see trW

trunc	default TRUE: truncate Smirnov correction term, see trW
eq7	default TRUEuse equation 7 in Smirnov and Anselin (2009), if FALSE no unit root correction
SE_method	default "LU", alternatively "MC"; underlying Indet method to use for generating SE toolbox emulation grid
nrho	default 200, number of Indet values in first stage SE toolbox emulation grid
interval	default $c(-1,0.999)$ if interval argument NULL, bounds for SE toolbox emulation grid
interpn	default 2000, number of Indet values to interpolate in second stage SE toolbox emulation grid
SElndet	default NULL, used to pass a pre-computed two-column matrix of coefficient values and corresponding interpolated Indet values

#### **Details**

Since environments are containers in the R workspace passed by reference rather than by value, they are useful for passing objects to functions called in numerical optimisation, here for the maximum likelihood estimation of spatial regression models. This technique can save a little time on each function call, balanced against the need to access the objects in the environment inside the function. The environment should contain a family string object either "SAR", "CAR" or "SMA" (used in do\_ldet to choose spatial moving average in spautolm, and these specific objects before calling the set-up functions:

```
eigen Classical Ord eigenvalue computations - either:
```

```
listw A listw spatial weights object
     can.sim logical scalar: can the spatial weights be made symmetric by similarity
     verbose logical scalar: legacy report print control, for historical reasons only
     pre_eig pre-computed eigenvalues
     and assigns to the environment:
     eig a vector of eigenvalues
     eig.range the search interval for the spatial coefficient
     method string: "eigen"
Matrix Sparse matrix pre-computed Cholesky decomposition with fast updating:
     listw A listw spatial weights object
     can.sim logical scalar: can the spatial weights be made symmetric by similarity
     and assigns to the environment:
     csrw sparse spatial weights matrix
     nW negative sparse spatial weights matrix
     pChol a "CHMfactor" from factorising csrw with Cholesky
     nChol a "CHMfactor" from factorising nW with Cholesky
     method string: "Matrix"
Matrix_J Standard Cholesky decomposition without updating:
```

listw A listw spatial weights objectcan.sim logical scalar: can the spatial weights be made symmetric by similarity

**n** number of spatial objects and assigns to the environment:

**csrw** sparse spatial weights matrix

I sparse identity matrix

super the value of the super argument

method string: "Matrix\_J"

spam Standard Cholesky decomposition without updating:

listw A listw spatial weights object

can.sim logical scalar: can the spatial weights be made symmetric by similarity

**n** number of spatial objects

and assigns to the environment:

csrw sparse spatial weights matrix

I sparse identity matrix

pivot string — pivot method

method string: "spam"

**spam\_update** Pre-computed Cholesky decomposition with updating:

listw A listw spatial weights object

can.sim logical scalar: can the spatial weights be made symmetric by similarity

n number of spatial objects

and assigns to the environment:

csrw sparse spatial weights matrix

I sparse identity matrix

csrwchol A Cholesky decomposition for updating

method string: "spam"

LU Standard LU decomposition without updating:

**listw** A listw spatial weights object

**n** number of spatial objects

and assigns to the environment:

W sparse spatial weights matrix

I sparse identity matrix

method string: "LU"

**LU\_prepermutate** Standard LU decomposition with updating (pre-computed fill-reducing permutation):

listw A listw spatial weights object

n number of spatial objects

and assigns to the environment:

W sparse spatial weights matrix

lu\_order order argument to lu

pq 2-column matrix for row and column permutation for fill-reduction I sparse identity matrix method string: "LU" MC Monte Carlo approximation: listw A listw spatial weights object and assigns to the environment: clx list of Monte Carlo approximation terms (the first two simulated traces are replaced by their analytical equivalents) W sparse spatial weights matrix method string: "MC" cheb Chebyshev approximation: **listw** A listw spatial weights object and assigns to the environment: trT vector of Chebyshev approximation terms W sparse spatial weights matrix method string: "Chebyshev" moments moments approximation: **listw** A listw spatial weights object can.sim logical scalar: can the spatial weights be made symmetric by similarity and assigns to the environment: trs vector of traces, possibly approximated q12 integer vector of length 2, unit roots terms, ignored until 0.5-52 eq7 logical scalar: use equation 7 correct logical scalar: use Smirnov correction term trunc logical scalar: truncate Smirnov correction term method string: "moments" SE classic: listw A listw spatial weights object **n** number of spatial objects and assigns to the environment: detval two column matrix of Indet grid values method string: "SE\_classic" **SE\_method** string: "LU" or "MC" SE\_whichMin: listw A listw spatial weights object **n** number of spatial objects and assigns to the environment: detval two column matrix of Indet grid values method string: "SE\_whichMin" **SE\_method** string: "LU" or "MC"

### **SE\_interp**:

```
listw A listw spatial weights object

n number of spatial objects

and assigns to the environment:

fit fitted spline object from which to predict Indet values

method string: "SE_interp"

SE method string: "LU" or "MC"
```

Some set-up functions may also assign similar to the environment if the weights were made symmetric by similarity.

Three set-up functions emulate the behaviour of the Spatial Econometrics toolbox (March 2010) maximum likelihood Indet grid performance. The toolbox Indet functions compute a smaller number of Indet values for a grid of coefficient values (spacing 0.01), and then interpolate to a finer grid of values (spacing 0.001). "SE\_classic", which is an implementation of the SE toolbox code, for example in f\_sar.m, appears to have selected a row in the grid matrix one below the correct row when the candidate coefficient value was between 0.005 and 0.01-fuzz, always rounding the row index down. A possible alternative is to choose the index that is closest to the candidate coefficient value ("SE\_whichMin"). Another alternative is to fit a spline model to the first stage coarser grid, and pass this fitted model to the log likelihood function to make a point prediction using the candidate coefficient value, rather than finding the grid index ("SE\_interp").

#### Value

do\_ldet returns the value of the Jacobian for the calculation method recorded in the environment argument, and for the Monte Carlo approximation, returns a measure of the spread of the approximation as an "sd" attribute; the remaining functions modify the environment in place as a side effect and return nothing.

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

LeSage J and RK Pace (2009) Introduction to Spatial Econometrics. CRC Press, Boca Raton, pp. 77–110

# See Also

```
spautolm, lagsarlm, errorsarlm, Cholesky
```

```
data(boston)
lw <- nb2listw(boston.soi)
can.sim <- spdep:::can.be.simmed(lw)
env <- new.env(parent=globalenv())
assign("listw", lw, envir=env)
assign("can.sim", can.sim, envir=env)</pre>
```

```
assign("similar", FALSE, envir=env)
assign("verbose", FALSE, envir=env)
assign("family", "SAR", envir=env)
eigen_setup(env)
get("similar", envir=env)
do_ldet(0.5, env)
rm(env)
env <- new.env(parent=globalenv())</pre>
assign("listw", lw, envir=env)
assign("can.sim", can.sim, envir=env)
assign("similar", FALSE, envir=env)
assign("verbose", FALSE, envir=env)
assign("family", "SAR", envir=env)
assign("n", length(boston.soi), envir=env)
eigen_pre_setup(env, pre_eig=eigenw(similar.listw(lw)))
do_ldet(0.5, env)
rm(env)
env <- new.env(parent=globalenv())</pre>
assign("listw", lw, envir=env)
assign("can.sim", can.sim, envir=env)
assign("similar", FALSE, envir=env)
assign("family", "SAR", envir=env)
assign("n", length(boston.soi), envir=env)
Matrix_setup(env, Imult=2, super=FALSE)
get("similar", envir=env)
do_ldet(0.5, env)
rm(env)
env <- new.env(parent=globalenv())</pre>
assign("listw", lw, envir=env)
assign("n", length(boston.soi), envir=env)
assign("can.sim", can.sim, envir=env)
assign("similar", FALSE, envir=env)
assign("family", "SAR", envir=env)
spam_setup(env)
get("similar", envir=env)
do_ldet(0.5, env)
rm(env)
env <- new.env(parent=globalenv())</pre>
assign("listw", lw, envir=env)
assign("n", length(boston.soi), envir=env)
assign("similar", FALSE, envir=env)
assign("family", "SAR", envir=env)
LU_setup(env)
get("similar", envir=env)
do_ldet(0.5, env)
rm(env)
env <- new.env(parent=globalenv())</pre>
assign("listw", lw, envir=env)
assign("n", length(boston.soi), envir=env)
assign("similar", FALSE, envir=env)
assign("family", "SAR", envir=env)
LU_prepermutate_setup(env)
get("similar", envir=env)
```

36 droplinks

```
do_ldet(0.5, env)
rm(env)
env <- new.env(parent=globalenv())</pre>
assign("listw", lw, envir=env)
assign("similar", FALSE, envir=env)
assign("family", "SAR", envir=env)
cheb_setup(env, q=5)
get("similar", envir=env)
do_ldet(0.5, env)
rm(env)
env <- new.env(parent=globalenv())</pre>
assign("listw", lw, envir=env)
assign("n", length(boston.soi), envir=env)
assign("similar", FALSE, envir=env)
assign("family", "SAR", envir=env)
set.seed(12345)
mcdet_setup(env, p=16, m=30)
get("similar", envir=env)
do_ldet(0.5, env)
rm(env)
```

droplinks

Drop links in a neighbours list

# Description

Drops links to and from or just to a region from a neighbours list. The example corresponds to Fingleton's Table 1, p. 6, for lattices 5 to 19.

### Usage

```
droplinks(nb, drop, sym=TRUE)
```

# **Arguments**

nb a neighbours list object of class nb

drop either a logical vector the length of nb, or a character vector of named regions corresponding to nb's region.id attribute, or an integer vector of region numbers sym TRUE for removal of both "row" and "column" links, FALSE for only "row"

links

### Value

The function returns an object of class nb with a list of integer vectors containing neighbour region number ids.

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

EBest 37

### References

B. Fingleton (1999) Spurious spatial regression: some Monte Carlo results with a spatial unit root and spatial cointegration, Journal of Regional Science 39, pp. 1–19.

#### See Also

```
is.symmetric.nb
```

## **Examples**

```
rho <- c(0.2, 0.5, 0.95, 0.999, 1.0)
ns <- c(5, 7, 9, 11, 13, 15, 17, 19)
mns <- matrix(0, nrow=length(ns), ncol=length(rho))</pre>
rownames(mns) <- ns</pre>
colnames(mns) <- rho</pre>
mxs <- matrix(0, nrow=length(ns), ncol=length(rho))</pre>
rownames(mxs) <- ns</pre>
colnames(mxs) <- rho</pre>
for (i in 1:length(ns)) {
  nblist <- cell2nb(ns[i], ns[i])</pre>
  nbdropped <- droplinks(nblist, ((ns[i]*ns[i])+1)/2, sym=FALSE)</pre>
  listw <- nb2listw(nbdropped, style="W", zero.policy=TRUE)</pre>
  wmat <- listw2mat(listw)</pre>
  for (j in 1:length(rho)) {
    mat <- diag(ns[i]*ns[i]) - rho[j] * wmat</pre>
    res <- diag(solve(t(mat) %*% mat))</pre>
    mns[i,j] <- mean(res)</pre>
    mxs[i,j] <- max(res)</pre>
}
print(mns)
print(mxs)
```

**EBest** 

Global Empirical Bayes estimator

### **Description**

The function computes global empirical Bayes estimates for rates "shrunk" to the overall mean.

## Usage

```
EBest(n, x, family="poisson")
```

38 EBest

## Arguments

n a numeric vector of counts of cases

x a numeric vector of populations at risk

family either "poisson" for rare conditions or "binomial" for non-rare conditions

## **Details**

Details of the implementation for the "poisson" family are to be found in Marshall, p. 284–5, and Bailey and Gatrell p. 303–306 and exercise 8.2, pp. 328–330. For the "binomial" family, see Martuzzi and Elliott (implementation by Olaf Berke).

#### Value

A data frame with two columns:

raw a numerical vector of raw (crude) rates

estmm a numerical vector of empirical Bayes estimates

and a parameters attribute list with components:

a global method of moments phi value
m global method of moments gamma value

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no> and Olaf Berke, Population Medicine, OVC, University of Guelph, CANADA

### References

Marshall R M (1991) Mapping disease and mortality rates using Empirical Bayes Estimators, Applied Statistics, 40, 283–294; Bailey T, Gatrell A (1995) Interactive Spatial Data Analysis, Harlow: Longman, pp. 303–306, Martuzzi M, Elliott P (1996) Empirical Bayes estimation of small area prevalence of non-rare conditions, Statistics in Medicine 15, 1867–1873.

### See Also

```
EBlocal, probmap, EBImoran.mc
```

```
example(auckland)
res <- EBest(auckland$M77_85, 9*auckland$Und5_81)
attr(res, "parameters")
cols <- grey(6:2/7)
brks <- c(-Inf,2,2.5,3,3.5,Inf)
plot(auckland, col=cols[findInterval(res$estmm*1000, brks, all.inside=TRUE)])
legend("bottomleft", fill=cols, legend=leglabs(brks), bty="n")
title(main="Global moment estimator of infant mortality per 1000 per year")
data(huddersfield)</pre>
```

EBImoran.mc 39

```
res <- EBest(huddersfield$cases, huddersfield$total, family="binomial") round(res[,1:2],4)\times100
```

EBImoran.mc

Permutation test for empirical Bayes index

## **Description**

An empirical Bayes index modification of Moran's I for testing for spatial autocorrelation in a rate, typically the number of observed cases in a population at risk. The index value is tested by using nsim random permutations of the index for the given spatial weighting scheme, to establish the rank of the observed statistic in relation to the nsim simulated values.

## Usage

```
EBImoran.mc(n, x, listw, nsim, zero.policy = NULL,
   alternative = "greater", spChk=NULL, return_boot=FALSE)
```

### **Arguments**

n a numeric vector of counts of cases the same length as the neighbou listw	
x	a numeric vector of populations at risk the same length as the neighbours list in listw
listw	a listw object created for example by nb2listw
nsim	number of permutations
zero.policy default NULL, use global option value; if TRUE assign zero to the of zones without neighbours, if FALSE assign NA	
alternative	a character string specifying the alternative hypothesis, must be one of "greater" (default), or "less"
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
return_boot	return an object of class boot from the equivalent permutation bootstrap rather than an object of class htest

### **Details**

The statistic used is (m is the number of observations):

$$EBI = \frac{m}{\sum_{i=1}^{m} \sum_{j=1}^{m} w_{ij}} \frac{\sum_{i=1}^{m} \sum_{j=1}^{m} w_{ij} z_{i} z_{j}}{\sum_{i=1}^{m} (z_{i} - \bar{z})^{2}}$$

where:

$$z_i = \frac{p_i - b}{\sqrt{v_i}}$$

and:

$$p_{i} = n_{i}/x_{i}$$

$$v_{i} = a + (b/x_{i})$$

$$b = \sum_{i=1}^{m} n_{i}/\sum_{i=1}^{m} x_{i}$$

$$a = s^{2} - b/(\sum_{i=1}^{m} x_{i}/m)$$

$$s^{2} = \sum_{i=1}^{m} x_{i}(p_{i} - b)^{2}/\sum_{i=1}^{m} x_{i}$$

#### Value

A list with class htest and mc.sim containing the following components:

statistic the value of the observed Moran's I.

p.value the pseudo p-value of the test.

alternative a character string describing the alternative hypothesis.

method a character string giving the method used.

data.name a character string giving the name(s) of the data, and the number of simulations.

res nsim simulated values of statistic, final value is observed statistic

z a numerical vector of Empirical Bayes indices as z above

### Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

#### References

Assunção RM, Reis EA 1999 A new proposal to adjust Moran's I for population density. Statistics in Medicine 18, pp. 2147–2162

## See Also

```
moran, moran.mc, EBest
```

```
example(nc.sids)
EBImoran.mc(nc.sids$SID74, nc.sids$BIR74,
   nb2listw(ncCC89_nb, style="B", zero.policy=TRUE), nsim=999, zero.policy=TRUE)
sids.p <- nc.sids$SID74 / nc.sids$BIR74
moran.mc(sids.p, nb2listw(ncCC89_nb, style="B", zero.policy=TRUE),
   nsim=999, zero.policy=TRUE)</pre>
```

EBlocal 41

|--|

## Description

The function computes local empirical Bayes estimates for rates "shrunk" to a neighbourhood mean for neighbourhoods given by the nb neighbourhood list.

## Usage

```
EBlocal(ri, ni, nb, zero.policy = NULL, spChk = NULL, geoda=FALSE)
```

## **Arguments**

ri	a numeric vector of counts of cases the same length as the neighbours list in nb
ni	a numeric vector of populations at risk the same length as the neighbours list in nb
nb	a nb object of neighbour relationships
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
geoda	default=FALSE, following Marshall's algorithm as interpreted by Bailey and Gatrell, pp. 305-307, and exercise 8.2, pp. 328-330 for the definition of phi; TRUE for the definition of phi used in GeoDa (see discussion on OpenSpace mailing list June 2003: http://agec221.agecon.uiuc.edu/pipermail/openspace/2003-June/thread.html)

## **Details**

Details of the implementation are to be found in Marshall, p. 286, and Bailey and Gatrell p. 307 and exercise 8.2, pp. 328–330. The example results do not fully correspond to the sources because of slightly differing neighbourhoods, but are generally close.

## Value

A data frame with two columns:

raw a numerical vector of raw (crude) rates
est a numerical vector of local empirical Bayes estimates
and a parameters attribute list with components:

a a numerical vector of local phi values
m a numerical vector of local gamma values

42 edit.nb

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>, based on contributions by Marilia Carvalho

#### References

Marshall R M (1991) Mapping disease and mortality rates using Empirical Bayes Estimators, Applied Statistics, 40, 283–294; Bailey T, Gatrell A (1995) Interactive Spatial Data Analysis, Harlow: Longman, pp. 303–306.

#### See Also

```
EBest, probmap
```

## **Examples**

```
example(auckland)
res <- EBlocal(auckland$M77_85, 9*auckland$Und5_81, auckland.nb)
brks <- c(-Inf,2,2.5,3,3.5,Inf)
cols <- grey(6:2/7)
plot(auckland, col=cols[findInterval(res$est*1000, brks, all.inside=TRUE)])
legend("bottomleft", fill=cols, legend=leglabs(brks), bty="n")
title(main="Local moment estimator of infant mortality per 1000 per year")</pre>
```

edit.nb

Interactive editing of neighbours lists

### **Description**

The function provides simple interactive editing of neighbours lists to allow unneeded links to be deleted, and missing links to be inserted. It uses identify to pick the endpoints of the link to be deleted or added, and asks for confirmation before committing. If the result is not assigned to a new object, the editing will be lost - as in edit.

## Usage

```
## S3 method for class 'nb'
edit(name, coords, polys=NULL, ..., use_region.id=FALSE)
```

# **Arguments**

name	an object of class nb
coords	matrix of region point coordinates; if missing and polys= inherits from SpatialPolygons, the label points of that object are used
polys	if polygon boundaries supplied, will be used as background; must inherit from SpatialPolygons
	further arguments passed to or from other methods
use_region.id	default FALSE, in identify use 1-based observation numbers, otherwise use the nb region.id attribute values

eigenw 43

### Value

The function returns an object of class nb with the edited list of integer vectors containing neighbour region number ids, with added attributes tallying the added and deleted links.

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

```
summary.nb, plot.nb
```

## **Examples**

```
## Not run:
data(columbus)
class(polys)
nnb <- edit.nb(col.gal.nb, coords, polys)
example(columbus)
class(columbus)
nnb1 <- edit.nb(col.gal.nb, polys=columbus)
## End(Not run)</pre>
```

eigenw

Spatial weights matrix eigenvalues

### **Description**

The eigenw function returns a numeric vector of eigenvalues of the weights matrix generated from the spatial weights object listw. The eigenvalues are used to speed the computation of the Jacobian in spatial model estimation:

$$\log(\det[I - \rho W]) = \sum_{i=1}^{n} \log(1 - \rho \lambda_i)$$

where W is the n by n spatial weights matrix, and  $\lambda_i$  are the eigenvalues of W.

### Usage

```
eigenw(listw, quiet=NULL)
griffith_sone(P, Q, type="rook")
subgraph_eigenw(nb, glist=NULL, style="W", zero.policy=NULL, quiet=NULL)
```

44 eigenw

#### **Arguments**

listw a listw object created for example by nb2listw		a listw object created for example by nb2listw
	quiet	default NULL, use global !verbose option value; set to FALSE for short summary
	Р	number of columns in the grid (number of units in a horizontal axis direction)
	Q	number of rows in the grid (number of units in a vertical axis direction.)
	type	"rook" or "queen"
	nb	an object of class nb
	glist	list of general weights corresponding to neighbours

style style can take values "W", "B", "C", "U", "minmax" and "S"

zero.policy default NULL, use global option value; if FALSE stop with error for any empty

neighbour sets, if TRUE permit the weights list to be formed with zero-length

weights vectors

#### **Details**

The eigenw function computes the eigenvalues of a single spatial weights object. The griffith\_sone function function may be used, following Ord and Gasim (for references see Griffith and Sone (1995)), to calculate analytical eigenvalues for binary rook or queen contiguous neighbours where the data are arranged as a regular P times Q grid. The subgraph\_eigenw function may be used when there are multiple graph components, of which the largest may be handled as a dense matrix. Here the eigenvalues are computed for each subgraph in turn, and catenated to reconstruct the complete set. The functions may be used to provide pre-computed eigenvalues for spatial regression functions.

### Value

a numeric or complex vector of eigenvalues of the weights matrix generated from the spatial weights object.

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 155; Ord, J. K. 1975 Estimation methods for models of spatial interaction, Journal of the American Statistical Association, 70, 120-126.; Griffith, D. A. and Sone, A. (1995). Trade-offs associated with normalizing constant computational simplifications for estimating spatial statistical models. Journal of Statistical Computation and Simulation, 51, 165-183.

#### See Also

eigen

eigenw 45

```
data(oldcol)
W.eig <- eigenw(nb2listw(COL.nb, style="W"))</pre>
1/range(W.eig)
S.eig <- eigenw(nb2listw(COL.nb, style="S"))</pre>
1/range(S.eig)
B.eig <- eigenw(nb2listw(COL.nb, style="B"))</pre>
1/range(B.eig)
# cases for intrinsically asymmetric weights
crds <- cbind(COL.OLD$X, COL.OLD$Y)</pre>
k3 <- knn2nb(knearneigh(crds, k=3))
is.symmetric.nb(k3)
k3eig <- eigenw(nb2listw(k3, style="W"))</pre>
is.complex(k3eig)
rho <- 0.5
Jc \leftarrow sum(log(1 - rho * k3eig))
# complex eigenvalue Jacobian
Jc
# subgraphs
nc <- n.comp.nb(k3)
nc$nc
table(nc$comp.id)
k3eigSG <- subgraph_eigenw(k3, style="W")</pre>
all.equal(sort(k3eig), k3eigSG)
\label{eq:wave_problem} W <- \ as(as\_dgRMatrix\_listw(nb2listw(k3, \ style="W")), \ "CsparseMatrix")
I <- diag(length(k3))</pre>
Jl <- sum(log(abs(diag(slot(lu(I - rho * W), "U")))))</pre>
# LU Jacobian equals complex eigenvalue Jacobian
Jl
all.equal(Re(Jc), Jl)
# wrong value if only real part used
Jr \leftarrow sum(log(1 - rho * Re(k3eig)))
Jr
all.equal(Jr, Jl)
# construction of Jacobian from complex conjugate pairs (Jan Hauke)
Rev <- Re(k3eig)[which(Im(k3eig) == 0)]
# real eigenvalues
Cev <- k3eig[which(Im(k3eig) != 0)]</pre>
pCev \leftarrow Cev[Im(Cev) > 0]
# separate complex conjugate pairs
RpCev <- Re(pCev)
IpCev <- Im(pCev)</pre>
# reassemble Jacobian
Jc1 <- sum(log(1 - rho*Rev)) + sum(log((1 - rho * RpCev)^2 + (rho^2)*(IpCev^2)))
all.equal(Re(Jc), Jc1)
# impact of omitted complex part term in real part only Jacobian
Jc2 \leftarrow sum(log(1 - rho*Rev)) + sum(log((1 - rho * RpCev)^2))
all.equal(Jr, Jc2)
# trace of asymmetric (WW) and crossprod of complex eigenvalues for APLE
sum(diag(W %*% W))
crossprod(k3eig)
# analytical regular grid eigenvalues
```

46 eire

```
rg <- cell2nb(ncol=7, nrow=7, type="rook")
rg_eig <- eigenw(nb2listw(rg, style="B"))
rg_GS <- griffith_sone(P=7, Q=7, type="rook")
all.equal(rg_eig, rg_GS)</pre>
```

eire

Eire data sets

#### **Description**

The Eire data set has been converted to shapefile format and placed in the etc/shapes directory. The initial data objects are now stored as a SpatialPolygonsDataFrame object, from which the contiguity neighbour list is recreated. For purposes of record, the original data set is retained.

The eire.df data frame has 26 rows and 9 columns. In addition, polygons of the 26 counties are provided as a multipart polylist in eire.polys.utm (coordinates in km, projection UTM zone 30). Their centroids are in eire.coords.utm. The original Cliff and Ord binary contiguities are in eire.nb.

## Usage

data(eire)

#### **Format**

This data frame contains the following columns:

A Percentage of sample with blood group A

towns Towns/unit area

**pale** Beyond the Pale 0, within the Pale 1

size number of blood type samples

**ROADACC** arterial road network accessibility in 1961

**OWNCONS** percentage in value terms of gross agricultural output of each county consumed by itself

**POPCHG** 1961 population as percentage of 1926

**RETSALE** value of retail sales £000 **INCOME** total personal income £000

names County names

### Source

Upton and Fingleton 1985, - Bailey and Gatrell 1995, ch. 1 for blood group data, Cliff and Ord (1973), p. 107 for remaining variables (also after O'Sullivan, 1968). Polygon borders and Irish data sourced from Michael Tiefelsdorf's SPSS Saddlepoint bundle, originally hosted at: http://geog-www.sbs.ohio-state.edu/faculty/tiefelsdorf/GeoStat.htm.

eire 47

```
require(maptools)
eire <- readShapePoly(system.file("etc/shapes/eire.shp", package="spdep")[1],</pre>
 ID="names", proj4string=CRS("+proj=utm +zone=30 +units=km"))
eire.nb <- poly2nb(eire)
#data(eire)
summary(eire$A)
brks <- round(fivenum(eire$A), digits=2)</pre>
cols <- rev(heat.colors(4))</pre>
plot(eire, col=cols[findInterval(eire$A, brks, all.inside=TRUE)])
title(main="Percentage with blood group A in Eire")
legend(x=c(-50, 70), y=c(6120, 6050), leglabs(brks), fill=cols, bty="n")
plot(eire)
plot(eire.nb, coordinates(eire), add=TRUE)
1A <- lag.listw(nb2listw(eire.nb), eire$A)</pre>
summary(1A)
moran.test(eire$A, nb2listw(eire.nb))
geary.test(eire$A, nb2listw(eire.nb))
cor(lA, eire$A)
moran.plot(eire$A, nb2listw(eire.nb),
labels=eire$names)
A.lm <- lm(A ~ towns + pale, data=eire)
summary(A.lm)
res <- residuals(A.lm)</pre>
brks <- c(min(res), -2, -1, 0, 1, 2, max(res))
cols <- rev(cm.colors(6))</pre>
plot(eire, col=cols[findInterval(res, brks, all.inside=TRUE)])
title(main="Regression residuals")
legend(x=c(-50, 70), y=c(6120, 6050), legend=leglabs(brks), fill=cols,
 bty="n")
lm.morantest(A.lm, nb2listw(eire.nb))
lm.morantest.sad(A.lm, nb2listw(eire.nb))
lm.LMtests(A.lm, nb2listw(eire.nb), test="LMerr")
brks <- round(fivenum(eire$OWNCONS), digits=2)</pre>
cols <- grey(4:1/5)
plot(eire, col=cols[findInterval(eire$OWNCONS, brks, all.inside=TRUE)])
title(main="Percentage own consumption of agricultural produce")
legend(x=c(-50, 70), y=c(6120, 6050), legend=leglabs(brks),
 fill=cols, bty="n")
moran.plot(eire$OWNCONS, nb2listw(eire.nb))
moran.test(eire$OWNCONS, nb2listw(eire.nb))
e.lm <- lm(OWNCONS ~ ROADACC, data=eire)</pre>
res <- residuals(e.lm)</pre>
brks <- c(min(res), -2, -1, 0, 1, 2, max(res))
cols <- rev(cm.colors(6))</pre>
plot(eire, col=cols[findInterval(res, brks, all.inside=TRUE)])
title(main="Regression residuals")
legend(x=c(-50, 70), y=c(6120, 6050), legend=leglabs(brks), fill=cm.colors(6),
 bty="n")
lm.morantest(e.lm, nb2listw(eire.nb))
```

48 elect80

```
lm.morantest.sad(e.lm, nb2listw(eire.nb))
lm.LMtests(e.lm, nb2listw(eire.nb), test="LMerr")
print(localmoran.sad(e.lm, eire.nb, select=1:length(slot(eire, "polygons"))))
```

elect80

1980 Presidential election results

# Description

A data set for 1980 Presidential election results covering 3,107 US counties using geographical coordinates. In addition, three spatial neighbour objects, k4 not using Great Circle distances, d11 using Great Circle distances, and e80\_queen of Queen contiguities for equivalent County polygons taken from file co1980p020.tar.gz on the USGS National Atlas site, and a spatial weights object imported from elect.ford - a 4-nearest neighbour non-GC row-standardised object, but with coercion to symmetry.

# Usage

data(elect80)

#### **Format**

A SpatialPointsDataFrame with 3107 observations on the following 7 variables.

FIPS a factor of county FIPS codes

long a numeric vector of longitude values

lat a numeric vector of latitude values

pc\_turnout Votes cast as proportion of population over age 19 eligible to vote

pc\_college Population with college degrees as proportion of population over age 19 eligible to vote

pc\_homeownership Homeownership as proportion of population over age 19 eligible to vote pc\_income Income per capita of population over age 19 eligible to vote

## Source

Pace, R. Kelley and Ronald Barry. 1997. "Quick Computation of Spatial Autoregressive Estimators", in Geographical Analysis; sourced from the data folder in the Spatial Econometrics Toolbox for Matlab, http://www.spatial-econometrics.com/html/jplv7.zip, files elect.dat and elect.ford (with the final line dropped).

# **Examples**

data(elect80)

errorsarlm	Spatial simultaneous autoregressive error model estimation	

#### **Description**

Maximum likelihood estimation of spatial simultaneous autoregressive error models of the form:

$$y = X\beta + u, u = \lambda Wu + \varepsilon$$

where  $\lambda$  is found by optimize() first, and  $\beta$  and other parameters by generalized least squares subsequently. With one of the sparse matrix methods, larger numbers of observations can be handled, but the interval= argument may need be set when the weights are not row-standardised.

### Usage

```
errorsarlm(formula, data=list(), listw, na.action, etype="error",
method="eigen", quiet=NULL, zero.policy=NULL,
interval = NULL, tol.solve=1.0e-10, trs=NULL, control=list())
```

### **Arguments**

formula a symbolic	description of the mo	del to be fit. The d	letails of model specification
--------------------	-----------------------	----------------------	--------------------------------

are given for lm()

data an optional data frame containing the variables in the model. By default the

variables are taken from the environment which the function is called.

listw a listw object created for example by nb2listw

na.action a function (default options("na.action")), can also be na.omit or na.exclude

with consequences for residuals and fitted values - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create no-neighbour observations. Note that only weights lists created without using the glist argument to

nb2listw may be subsetted.

etype default "error", may be set to "emixed" to include the spatially lagged inde-

pendent variables added to X; when "emixed", the lagged intercept is dropped for spatial weights style "W", that is row-standardised weights, but otherwise

included

method "eigen" (default) - the Jacobian is computed as the product of (1 - rho\*eigenvalue)

using eigenw, and "spam" or "Matrix\_J" for strictly symmetric weights lists of styles "B" and "C", or made symmetric by similarity (Ord, 1975, Appendix C) if possible for styles "W" and "S", using code from the spam package or Matrix package to calculate the determinant; "Matrix" and "spam\_update" provide updating Cholesky decomposition methods; "LU" provides an alternative sparse matrix decomposition approach. In addition, there are "Chebyshev" and Monte Carlo "MC" approximate log-determinant methods; the Smirnov/Anselin (2009) trace approximation is available as "moments". Three methods: "SE\_classic",

"SE\_whichMin", and "SE\_interp" are provided experimentally, the first to attempt to emulate the behaviour of Spatial Econometrics toolbox ML fitting functions. All use grids of log determinant values, and the latter two attempt

to ameliorate some features of "SE\_classic".

quiet default NULL, use !verbose global option value; if FALSE, reports function

values during optimization.

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged value

of zones without neighbours, if FALSE assign NA - causing errorsarlm() to

terminate with an error

interval default is NULL, search interval for autoregressive parameter

tol.solve the tolerance for detecting linear dependencies in the columns of matrices to be

inverted - passed to solve() (default=1.0e-10). This may be used if necessary to extract coefficient standard errors (for instance lowering to 1e-12), but errors in solve() may constitute indications of poorly scaled variables: if the variables have scales differing much from the autoregressive coefficient, the values in this matrix may be very different in scale, and inverting such a matrix is analytically possible by definition, but numerically unstable; rescaling the RHS variables

alleviates this better than setting tol.solve to a very small value

trs default NULL, if given, a vector of powered spatial weights matrix traces output

by trW; when given, insert the asymptotic analytical values into the numerical Hessian instead of the approximated values; may be used to get around some problems raised when the numerical Hessian is poorly conditioned, generating NaNs in subsequent operations. When using the numerical Hessian to get the standard error of lambda, it is very strongly advised that trs be given, as the parts of fdHess corresponding to the regression coefficients are badly approximated, affecting the standard error of lambda; the coefficient correlation matrix

is unusable

control list of extra control arguments - see section below

### **Details**

The asymptotic standard error of  $\lambda$  is only computed when method=eigen, because the full matrix operations involved would be costly for large n typically associated with the choice of method="spam" or "Matrix". The same applies to the coefficient covariance matrix. Taken as the asymptotic matrix from the literature, it is typically badly scaled, being block-diagonal, and with the elements involving  $\lambda$  being very small, while other parts of the matrix can be very large (often many orders of magnitude in difference). It often happens that the tol.solve argument needs to be set to a smaller value than the default, or the RHS variables can be centred or reduced in range.

Note that the fitted() function for the output object assumes that the response variable may be reconstructed as the sum of the trend, the signal, and the noise (residuals). Since the values of the response variable are known, their spatial lags are used to calculate signal components (Cressie 1993, p. 564). This differs from other software, including GeoDa, which does not use knowledge of the response variable in making predictions for the fitting data.

## Value

A list object of class sarlm

type "error"

lambda simultaneous autoregressive error coefficient

coefficients GLS coefficient estimates

rest.se GLS coefficient standard errors (are equal to asymptotic standard errors)

LL log likelihood value at computed optimum

SSE GLS residual variance
sum of squared GLS errors

parameters number of parameters estimated

logLik\_lm.model

Log likelihood of the linear model for  $\lambda = 0$ 

AIC\_lm.model AIC of the linear model for  $\lambda = 0$ 

 $coef_{lm.model}$  coefficients of the linear model for  $\lambda = 0$ 

tarX model matrix of the GLS model tary response of the GLS model

y response of the linear model for  $\lambda=0$  X model matrix of the linear model for  $\lambda=0$  method the method used to calculate the Jacobian

call the call used to create this object

residuals GLS residuals

opt object returned from numerical optimisation

ase TRUE if method=eigen

se.fit Not used yet

lambda.se if ase=TRUE, the asymptotic standard error of  $\lambda$ 

LMtest NULL for this model

aliased if not NULL, details of aliased variables LLNullLlm Log-likelihood of the null linear model

Hcov Spatial DGP covariance matrix for Hausman test if available

interval line search interval fdHess finite difference Hessian optimHess optim or fdHess used

insert logical; is TRUE, asymptotic values inserted in fdHess where feasible

timings processing timings

f\_calls number of calls to the log likelihood function during optimization

hf\_calls number of calls to the log likelihood function during numerical Hessian compu-

tation

intern\_classic a data frame of detval matrix row choices used by the SE toolbox classic method

zero.policy zero.policy for this model

na.action (possibly) named vector of excluded or omitted observations if non-default na.action

argument used

The internal sar.error.\* functions return the value of the log likelihood function at  $\lambda$ .

### Control arguments

**tol.opt:** the desired accuracy of the optimization - passed to optimize() (default=square root of double precision machine tolerance, a larger root may be used needed, see help(boston) for an example)

returnHcov: default TRUE, return the Vo matrix for a spatial Hausman test

**pWOrder:** default 250, if returnHcov=TRUE and the method is not "eigen", pass this order to powerWeights as the power series maximum limit

**fdHess:** default NULL, then set to (method != "eigen") internally; use fdHess to compute an approximate Hessian using finite differences when using sparse matrix methods; used to make a coefficient covariance matrix when the number of observations is large; may be turned off to save resources if need be

**optimHess:** default FALSE, use fdHess from **nlme**, if TRUE, use optim to calculate Hessian at optimum

optimHessMethod: default "optimHess", may be "nlm" or one of the optim methods

LAPACK: default FALSE; logical value passed to qr in the SSE log likelihood function

**compiled\_sse:** default FALSE; logical value used in the log likelihood function to choose compiled code for computing SSE

**Imult:** default 2; used for preparing the Cholesky decompositions for updating in the Jacobian function

super: if NULL (default), set to FALSE to use a simplicial decomposition for the sparse Cholesky decomposition and method "Matrix\_J", set to as.logical(NA) for method "Matrix", if TRUE, use a supernodal decomposition

cheb\_q: default 5; highest power of the approximating polynomial for the Chebyshev approximation

MC\_p: default 16; number of random variates

MC\_m: default 30; number of products of random variates matrix and spatial weights matrix

spamPivot: default "MMD", alternative "RCM"

in\_coef default 0.1, coefficient value for initial Cholesky decomposition in "spam\_update"

**type** default "MC", used with method "moments"; alternatives "mult" and "moments", for use if trs is missing, trW

correct default TRUE, used with method "moments" to compute the Smirnov/Anselin correction term

**trunc** default TRUE, used with method "moments" to truncate the Smirnov/Anselin correction term

SE\_method default "LU", may be "MC"

**nrho** default 200, as in SE toolbox; the size of the first stage Indet grid; it may be reduced to for example 40

interpn default 2000, as in SE toolbox; the size of the second stage Indet grid

small\_asy default TRUE; if the method is not "eigen", use asymmetric covariances rather than numerical Hessian ones if n <= small</pre>

**small** default 1500; threshold number of observations for asymmetric covariances when the method is not "eigen"

**SEIndet** default NULL, may be used to pass a pre-computed SE toolbox style matrix of coefficients and their Indet values to the "SE classic" and "SE whichMin" methods

**LU\_order** default FALSE; used in "LU\_prepermutate", note warnings given for 1u method **pre\_eig** default NULL; may be used to pass a pre-computed vector of eigenvalues

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion; Ord, J. K. 1975 Estimation methods for models of spatial interaction, Journal of the American Statistical Association, 70, 120-126; Anselin, L. 1988 Spatial econometrics: methods and models. (Dordrecht: Kluwer); Anselin, L. 1995 SpaceStat, a software program for the analysis of spatial data, version 1.80. Regional Research Institute, West Virginia University, Morgantown, WV; Anselin L, Bera AK (1998) Spatial dependence in linear regression models with an introduction to spatial econometrics. In: Ullah A, Giles DEA (eds) Handbook of applied economic statistics. Marcel Dekker, New York, pp. 237-289; Cressie, N. A. C. 1993 Statistics for spatial data, Wiley, New York; LeSage J and RK Pace (2009) Introduction to Spatial Econometrics. CRC Press, Boca Raton.

## See Also

lm, lagsarlm, similar.listw, predict.sarlm, residuals.sarlm, do\_ldet

```
data(oldcol)
lw <- nb2listw(COL.nb, style="W")
COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    lw, method="eigen", quiet=FALSE)
summary(COL.errW.eig, correlation=TRUE)
ev <- eigenw(similar.listw(lw))
COL.errW.eig_ev <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    lw, method="eigen", control=list(pre_eig=ev))
all.equal(coefficients(COL.errW.eig), coefficients(COL.errW.eig_ev))
COL.errB.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    nb2listw(COL.nb, style="B"), method="eigen", quiet=FALSE)
summary(COL.errB.eig, correlation=TRUE)
W <- as(as_dgRMatrix_listw(nb2listw(COL.nb)), "CsparseMatrix")
trMatc <- trW(W, type="mult")
COL.errW.M <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,</pre>
```

54 geary

```
lw, method="Matrix", quiet=FALSE, trs=trMatc)
summary(COL.errW.M, correlation=TRUE)
NA.COL.OLD <- COL.OLD
NA.COL.OLD$CRIME[20:25] <- NA
COL.err.NA <- errorsarlm(CRIME ~ INC + HOVAL, data=NA.COL.OLD,
nb2listw(COL.nb), na.action=na.exclude)
COL.err.NA$na.action
COL.err.NA
resid(COL.err.NA)
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="eigen"))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="eigen", control=list(LAPACK=TRUE)))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="eigen", control=list(compiled_sse=TRUE)))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,</pre>
lw, method="Matrix_J", control=list(super=TRUE)))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="Matrix_J", control=list(super=FALSE)))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="Matrix_J", control=list(super=as.logical(NA))))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="Matrix", control=list(super=TRUE)))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="Matrix", control=list(super=FALSE)))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,</pre>
lw, method="Matrix", control=list(super=as.logical(NA))))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="spam", control=list(spamPivot="MMD")))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="spam", control=list(spamPivot="RCM")))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="spam_update", control=list(spamPivot="MMD")))
system.time(COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="spam_update", control=list(spamPivot="RCM")))
COL.merrW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
lw, method="eigen", etype="emixed")
summary(COL.merrW.eig, correlation=TRUE)
```

geary

Compute Geary's C

#### **Description**

A simple function to compute Geary's C, called by geary.test and geary.mc;

$$C = \frac{(n-1)}{2\sum_{i=1}^{n} \sum_{j=1}^{n} w_{ij}} \frac{\sum_{i=1}^{n} \sum_{j=1}^{n} w_{ij} (x_i - x_j)^2}{\sum_{i=1}^{n} (x_i - \bar{x})^2}$$

geary. intern is an internal function used to vary the similarity criterion.

geary 55

### Usage

```
geary(x, listw, n, n1, S0, zero.policy=NULL)
```

## **Arguments**

x a numeric vector the same length as the neighbours list in listw

listw a listw object created for example by nb2listw

n number of zones

n1 n - 1

S0 global sum of weights

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged value

of zones without neighbours, if FALSE assign NA

### Value

a list with

C Geary's C

K sample kurtosis of x

# Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

```
Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 17.
```

## See Also

```
geary.test, geary.mc, sp.mantel.mc
```

56 geary.mc

geary.mc Permutation test for Geary's C statistic	geary.mc	Permutation test for Geary's C statistic	
---	----------	--	--

# Description

A permutation test for Geary's C statistic calculated by using nsim random permutations of x for the given spatial weighting scheme, to establish the rank of the observed statistic in relation to the nsim simulated values.

# Usage

```
geary.mc(x, listw, nsim, zero.policy=NULL, alternative="greater",
    spChk=NULL, adjust.n=TRUE, return_boot=FALSE)
```

# Arguments

x	a numeric vector the same length as the neighbours list in listw
listw	a listw object created for example by nb2listw
nsim number of permutations	
zero.policy default NULL, use global option value; if TRUE assign zero to the la of zones without neighbours, if FALSE assign NA	
alternative	a character string specifying the alternative hypothesis, must be one of "greater" (default), or "less"; this reversal corresponds to that on geary.test described in the section on the output statistic value, based on Cliff and Ord 1973, p. 21 (changed 2011-04-11, thanks to Daniel Garavito).
spChk should the data vector names be checked against the spatial object integrity, TRUE, or FALSE, default NULL to use get.spChkOptic	
adjust.n	default TRUE, if FALSE the number of observations is not adjusted for noneighbour observations, if TRUE, the number of observations is adjusted
return_boot	return an object of class boot from the equivalent permutation bootstrap rather than an object of class htest

## Value

A list with class htest and mc.sim containing the following components:

statistic	the value of the observed Geary's C.
parameter	the rank of the observed Geary's C.
p.value	the pseudo p-value of the test.
alternative	a character string describing the alternative hypothesis.
method	a character string giving the method used.
data.name	a character string giving the name(s) of the data, and the number of simulations.
res	nsim simulated values of statistic, final value is observed statistic

geary.test 57

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 63-5.

#### See Also

```
geary, geary.test
```

## **Examples**

```
data(oldcol)
sim1 <- geary.mc(COL.OLD$CRIME, nb2listw(COL.nb, style="W"),</pre>
nsim=99, alternative="less")
sim1
mean(sim1$res)
var(sim1$res)
summary(sim1$res)
colold.lags <- nblag(COL.nb, 3)</pre>
sim2 <- geary.mc(COL.OLD$CRIME, nb2listw(colold.lags[[2]],</pre>
 style="W"), nsim=99)
sim2
summary(sim2$res)
sim3 <- geary.mc(COL.OLD$CRIME, nb2listw(colold.lags[[3]],</pre>
style="W"), nsim=99)
sim3
summary(sim3$res)
```

geary.test

Geary's C test for spatial autocorrelation

## **Description**

Geary's test for spatial autocorrelation using a spatial weights matrix in weights list form. The assumptions underlying the test are sensitive to the form of the graph of neighbour relationships and other factors, and results may be checked against those of geary.mc permutations.

# Usage

58 geary.test

### **Arguments**

x a numeric vector the same length as the neighbours list in listw

 ${\tt listw} \qquad \qquad {\tt a \ listw} \ {\tt object} \ {\tt created} \ {\tt for} \ {\tt example} \ {\tt by} \ {\tt nb2listw}$ 

randomisation variance of I calculated under the assumption of randomisation, if FALSE nor-

mality

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged value

of zones without neighbours, if FALSE assign NA

alternative a character string specifying the alternative hypothesis, must be one of "greater"

(default), "less" or "two.sided".

spChk should the data vector names be checked against the spatial objects for identity

integrity, TRUE, or FALSE, default NULL to use get.spChkOption()

adjust.n default TRUE, if FALSE the number of observations is not adjusted for no-

neighbour observations, if TRUE, the number of observations is adjusted

#### Value

A list with class htest containing the following components:

statistic the value of the standard deviate of Geary's C, in the order given in Cliff and

Ord 1973, p. 21, which is (EC - C) / sqrt(VC), that is with the sign reversed with respect to the more usual (C - EC) / sqrt(VC); this means that the "greater" alternative for the Geary C test corresponds to the "greater" alternative for Moran's

I test.

p.value the p-value of the test.

estimate the value of the observed Geary's C, its expectation and variance under the

method assumption.

alternative a character string describing the alternative hypothesis.

method a character string giving the assumption used for calculating the standard devi-

ate.

data.name a character string giving the name(s) of the data.

### Note

The derivation of the test (Cliff and Ord, 1981, p. 18) assumes that the weights matrix is symmetric. For inherently non-symmetric matrices, such as k-nearest neighbour matrices, listw2U() can be used to make the matrix symmetric. In non-symmetric weights matrix cases, the variance of the test statistic may be negative (thanks to Franz Munoz I for a well documented bug report). Geary's C is affected by non-symmetric weights under normality much more than Moran's I. From 0.4-35, the sign of the standard deviate of C is changed to match Cliff and Ord (1973, p. 21).

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

getisord 59

### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 21, Cliff, A. D., Ord, J. K. 1973 Spatial Autocorrelation, Pion, pp. 15-16, 21.

### See Also

```
geary, geary.mc, listw2U
```

## **Examples**

```
data(oldcol)
geary.test(COL.OLD$CRIME, nb2listw(COL.nb, style="W"))
geary.test(COL.OLD$CRIME, nb2listw(COL.nb, style="W"),
randomisation=FALSE)
colold.lags <- nblag(COL.nb, 3)</pre>
geary.test(COL.OLD$CRIME, nb2listw(colold.lags[[2]],
style="W"))
geary.test(COL.OLD$CRIME, nb2listw(colold.lags[[3]],
style="W"), alternative="greater")
print(is.symmetric.nb(COL.nb))
coords.OLD <- cbind(COL.OLD$X, COL.OLD$Y)</pre>
COL.k4.nb <- knn2nb(knearneigh(coords.OLD, 4))</pre>
print(is.symmetric.nb(COL.k4.nb))
geary.test(COL.OLD$CRIME, nb2listw(COL.k4.nb, style="W"))
geary.test(COL.OLD$CRIME, nb2listw(COL.k4.nb, style="W"),
randomisation=FALSE)
cat("Note non-symmetric weights matrix - use listw2U()\n")
geary.test(COL.OLD$CRIME, listw2U(nb2listw(COL.k4.nb,
style="W")))
geary.test(COL.OLD$CRIME, listw2U(nb2listw(COL.k4.nb,
style="W")), randomisation=FALSE)
```

getisord

Getis-Ord remote sensing example data

## **Description**

The xyz data frame has 256 rows and 3 columns. Vectors x and y are of length 16 and give the centres of the grid rows and columns, 30m apart. The data start from the bottom left, Getis and Ord start from the top left - so their 136th grid cell is our 120th.

## Usage

```
data(getisord)
```

globalG.test

### **Format**

This data frame contains the following columns:

```
x grid eastingsy grid northingsval remote sensing values
```

#### Source

Getis, A. and Ord, J. K. 1996 Local spatial statistics: an overview. In P. Longley and M. Batty (eds) *Spatial analysis: modelling in a GIS environment* (Cambridge: Geoinformation International), 266.

### **Examples**

```
data(getisord)
image(x, y, t(matrix(xyz$val, nrow=16, ncol=16, byrow=TRUE)), asp=1)
text(xyz$x, xyz$y, xyz$val, cex=0.7)
polygon(c(195,225,225,195), c(195,195,225,225), lwd=2)
title(main="Getis-Ord 1996 remote sensing data")
```

globalG.test

Global G test for spatial autocorrelation

## **Description**

The global G statistic for spatial autocorrelation, complementing the local Gi LISA measures: localG.

## Usage

```
globalG.test(x, listw, zero.policy=NULL, alternative="greater",
    spChk=NULL, adjust.n=TRUE)
```

## **Arguments**

X	a numeric vector the same length as the neighbours list in listw
listw	a listw object created for example by nb2listw; if a sequence of distance bands is to be used, it is recommended that the weights style be binary (one of $c("B", "C", "U")$ ).
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
alternative	a character string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
adjust.n	default TRUE, if FALSE the number of observations is not adjusted for no- neighbour observations, if TRUE, the number of observations is adjusted

#### Value

A list with class htest containing the following components:

statistic the value of the standard deviate of Moran's I.

p.value the p-value of the test.

estimate the value of the observed statistic, its expectation and variance.

alternative a character string describing the alternative hypothesis.
data.name a character string giving the name(s) of the data.

#### Author(s)

Hisaji ONO <hi-ono@mn.xdsl.ne.jp> and Roger Bivand <Roger.Bivand@nhh.no>

#### References

Getis. A, Ord, J. K. 1992 The analysis of spatial association by use of distance statistics, *Geographical Analysis*, 24, p. 195; see also Getis. A, Ord, J. K. 1993 Erratum, *Geographical Analysis*, 25, p. 276.

#### See Also

localG

# **Examples**

```
example(nc.sids)
sidsrate79 <- (1000*nc.sids$SID79)/nc.sids$BIR79
dists <- c(10, 20, 30, 33, 40, 50, 60, 70, 80, 90, 100)
ndists <- length(dists)
ZG <- numeric(length=ndists)
milesxy <- cbind(nc.sids$east, nc.sids$north)
for (i in 1:ndists) {
   thisnb <- dnearneigh(milesxy, 0, dists[i])
   thislw <- nb2listw(thisnb, style="B", zero.policy=TRUE)
   ZG[i] <- globalG.test(sidsrate79, thislw, zero.policy=TRUE)$statistic
}
cbind(dists, ZG)</pre>
```

GMerrorsar

Spatial simultaneous autoregressive error model estimation by GMM

### **Description**

An implementation of Kelejian and Prucha's generalised moments estimator for the autoregressive parameter in a spatial model.

#### Usage

```
GMerrorsar(formula, data = list(), listw, na.action = na.fail,
  zero.policy = NULL, method="nlminb", arnoldWied=FALSE,
  control = list(), pars, scaleU=FALSE, verbose=NULL, legacy=FALSE,
  se.lambda=TRUE, returnHcov=FALSE, pWOrder=250, tol.Hcov=1.0e-10)
## S3 method for class 'gmsar'
summary(object, correlation = FALSE, Hausman=FALSE, ...)
GMargminImage(obj, lambdaseq, s2seq)
```

### **Arguments**

formula a symbolic description of the model to be fit. The details of model specification

are given for lm()

data an optional data frame containing the variables in the model. By default the

variables are taken from the environment which the function is called.

listw a listw object created for example by nb2listw

na.action a function (default na.fail), can also be na.omit or na.exclude with conse-

quences for residuals and fitted values - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create no-neighbour observations. Note that only weights lists created without using the glist argument to nb2listw may be

subsetted.

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged

value of zones without neighbours, if FALSE (default) assign NA - causing

GMerrorsar() to terminate with an error

method default "nlminb", or optionally a method passed to optim to use an alternative

optimizer

arnoldWied default FALSE

control A list of control parameters. See details in optim or nlminb.

pars starting values for  $\lambda$  and  $\sigma^2$  for GMM optimisation, if missing (default), approx-

imated from initial OLS model as the autocorrelation coefficient corrected for

weights style and model sigma squared

scaleU Default FALSE: scale the OLS residuals before computing the moment matri-

ces; only used if the pars argument is missing

verbose default NULL, use global option value; if TRUE, reports function values during

optimization.

legacy default FALSE - compute using the unfiltered values of the response and right

hand side variables; if TRUE - compute the fitted value and residuals from the

spatially filtered model using the spatial error parameter

se.lambda default TRUE, use the analytical method described in http://econweb.umd.

edu/~prucha/STATPROG/OLS/desols.pdf

returnHcov default FALSE, return the Vo matrix for a spatial Hausman test

to1. Hoov the tolerance for computing the Vo matrix (default=1.0e-10)

pWOrder default 250, if returnHcov=TRUE, pass this order to powerWeights as the power

series maximum limit

object, obj gmsar object from GMerrorsar

correlation logical; (default=FALSE), TRUE not available

Hausman if TRUE, the results of the Hausman test for error models are reported

... summary arguments passed through

lambdaseq if given, an increasing sequence of lambda values for gridding

s2seq if given, an increasing sequence of sigma squared values for gridding

#### **Details**

When the control list is set with care, the function will converge to values close to the ML estimator without requiring computation of the Jacobian, the most resource-intensive part of ML estimation.

Note that the fitted() function for the output object assumes that the response variable may be reconstructed as the sum of the trend, the signal, and the noise (residuals). Since the values of the response variable are known, their spatial lags are used to calculate signal components (Cressie 1993, p. 564). This differs from other software, including GeoDa, which does not use knowledge of the response variable in making predictions for the fitting data.

The GMargminImage may be used to visualize the shape of the surface of the argmin function used to find lambda.

### Value

A list object of class gmsar

type "ERROR"

lambda simultaneous autoregressive error coefficient

coefficients GMM coefficient estimates

rest.se GMM coefficient standard errors

SSE Sum of squared GMM errors
parameters number of parameters estimated

1m. model the 1m object returned when estimating for  $\lambda = 0$ 

call the call used to create this object

residuals GMM residuals

lm.target the lm object returned for the GMM fit

fitted.values Difference between residuals and response variable

formula model formula

aliased if not NULL, details of aliased variables

zero.policy zero.policy for this model

vv list of internal bigG and litg components for testing optimisation surface

optres object returned by optimizer

pars start parameter values for optimisation

Hcov Spatial DGP covariance matrix for Hausman test if available

na.action (possibly) named vector of excluded or omitted observations if non-default na.action

argument used

#### Author(s)

Luc Anselin and Roger Bivand

#### References

Kelejian, H. H., and Prucha, I. R., 1999. A Generalized Moments Estimator for the Autoregressive Parameter in a Spatial Model. International Economic Review, 40, pp. 509–533; Cressie, N. A. C. 1993 *Statistics for spatial data*, Wiley, New York.

### See Also

```
optim, nlminb, errorsarlm
```

```
data(oldcol)
COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
nb2listw(COL.nb, style="W"), method="eigen")
summary(COL.errW.eig, Hausman=TRUE)
COL.errW.GM <- GMerrorsar(CRIME ~ INC + HOVAL, data=COL.OLD,</pre>
nb2listw(COL.nb, style="W"), returnHcov=TRUE)
summary(COL.errW.GM, Hausman=TRUE)
aa <- GMargminImage(COL.errW.GM)</pre>
levs \leftarrow quantile(aa$z, seq(0, 1, 1/12))
image(aa, breaks=levs, xlab="lambda", ylab="s2")
points(COL.errW.GM$lambda, COL.errW.GM$s2, pch=3, lwd=2)
contour(aa, levels=signif(levs, 4), add=TRUE)
COL.errW.GM1 <- GMerrorsar(CRIME ~ INC + HOVAL, data=COL.OLD,
nb2listw(COL.nb, style="W"))
summary(COL.errW.GM1)
example(NY_data)
esar1f <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, family="SAR", method="eigen")
summary(esar1f)
esar1gm <- GMerrorsar(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,
data=nydata, listw=listw_NY)
summary(esar1gm)
esar1gm1 <- GMerrorsar(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,
data=nydata, listw=listw_NY, method="Nelder-Mead")
summary(esar1gm1)
```

Graph Components 65

**Graph Components** 

Depth First Search on Neighbor Lists

# Description

n.comp.nb() finds the number of disjoint connected subgraphs in the graph depicted by nb.obj - a spatial neighbours list object.

## Usage

```
n.comp.nb(nb.obj)
```

# Arguments

nb.obj a neighbours list object of class nb

# Value

A list of:

nc number of disjoint connected subgraphs

comp.id vector with the indices of the disjoint connected subgraphs that the nodes in

nb.obj belong to

## Author(s)

Nicholas Lewin-Koh <nikko@hailmail.net>

## See Also

```
plot.nb
```

```
example(columbus)
coords <- coordinates(columbus)
plot(col.gal.nb, coords, col="grey")
col2 <- droplinks(col.gal.nb, 21)
plot(col2, coords, add=TRUE)
res <- n.comp.nb(col2)
table(res$comp.id)
points(coords, col=res$comp.id, pch=16)</pre>
```

graphneigh graphneigh

Graph based spatial weights

## **Description**

Functions return a graph object containing a list with the vertex coordinates and the to and from indices defining the edges. The helper function graph2nb converts a graph object into a neighbour list. The plot functions plot the graph objects.

# Usage

```
gabrielneigh(coords, nnmult=3)
relativeneigh(coords, nnmult=3)
soi.graph(tri.nb, coords)
graph2nb(gob, row.names=NULL,sym=FALSE)
## S3 method for class 'Gabriel'
plot(x, show.points=FALSE, add=FALSE, linecol=par(col), ...)
## S3 method for class 'relative'
plot(x, show.points=FALSE, add=FALSE, linecol=par(col),...)
```

### **Arguments**

coords	matrix of region point coordinates
nnmult	scaling factor for memory allocation, default 3; if higher values are required, the function will exit with an error; example below thanks to Dan Putler
tri.nb	a neighbor list created from tri2nb
gob	a graph object created from any of the graph funtions
row.names	character vector of region ids to be added to the neighbours list as attribute region.id, default $seq(1, nrow(x))$
sym	a logical argument indicating whether or not neighbors should be symetric (if $i$ -> $j$ then $j$ -> $i$ )
X	object to be plotted
show.points	(logical) add points to plot
add	(logical) add to existing plot
linecol	edge plotting colour
	further graphical parameters as in par()

## **Details**

The graph functions produce graphs on a 2d point set that

are all subgraphs of the Delaunay triangulation. The relative neighbor graph is defined by the relation, x and y are neighbors if

graphneigh 67

$$d(x,y) \le min(max(d(x,z),d(y,z))|z \in S)$$

where d() is the distance, S is the set of points and z is an arbitrary point in S. The Gabriel graph is a subgraph of the delaunay triangulation and has the relative neighbor graph as a sub-graph. The relative neighbor graph is defined by the relation x and y are Gabriel neighbors if

$$d(x,y) \le \min((d(x,z)^2 + d(y,z)^2)^{1/2} | z \in S)$$

where x,y,z and S are as before. The sphere of influence graph is defined for a finite point set S, let  $r_x$  be the distance from point x to its nearest neighbor in S, and  $C_x$  is the circle centered on x. Then x and y are SOI neighbors iff  $C_x$  and  $C_y$  intersect in at least 2 places.

See card for details of "nb" objects.

#### Value

A list of class Graph withte following elements

np number of input points

from array of origin ids

to array of destination ids

nedges number of edges in graph

x input x coordinates

y input y coordinates

The helper functions return an nb object with a list of integer vectors containing neighbour region number ids.

## Author(s)

Nicholas Lewin-Koh <nikko@hailmail.net>

### References

Matula, D. W. and Sokal R. R. 1980, Properties of Gabriel graphs relevant to geographic variation research and the clustering of points in the plane, Geographic Analysis, 12(3), pp. 205-222.

Toussaint, G. T. 1980, The relative neighborhood graph of a finite planar set, Pattern Recognition, 12(4), pp. 261-268.

Kirkpatrick, D. G. and Radke, J. D. 1985, A framework for computational morphology. In Computational Geometry, Ed. G. T. Toussaint, North Holland.

#### See Also

knearneigh, dnearneigh, knn2nb, card

68 gstsls

### **Examples**

```
example(columbus)
coords <- coordinates(columbus)</pre>
par(mfrow=c(2,2))
col.tri.nb<-tri2nb(coords)</pre>
col.gab.nb<-graph2nb(gabrielneigh(coords), sym=TRUE)</pre>
col.rel.nb<- graph2nb(relativeneigh(coords), sym=TRUE)</pre>
col.soi.nb<- graph2nb(soi.graph(col.tri.nb,coords), sym=TRUE)</pre>
plot(columbus, border="grey")
plot(col.tri.nb,coords,add=TRUE)
title(main="Delaunay Triangulation")
plot(columbus, border="grey")
plot(col.gab.nb, coords, add=TRUE)
title(main="Gabriel Graph")
plot(columbus, border="grey")
plot(col.rel.nb, coords, add=TRUE)
title(main="Relative Neighbor Graph")
plot(columbus, border="grey")
plot(col.soi.nb, coords, add=TRUE)
title(main="Sphere of Influence Graph")
par(mfrow=c(1,1))
dx < - rep(0.25*0:4,5)
dy <- c(rep(0,5), rep(0.25,5), rep(0.5,5), rep(0.75,5), rep(1,5))
m \leftarrow cbind(c(dx, dx, 3+dx, 3+dx), c(dy, 3+dy, dy, 3+dy))
try(res <- gabrielneigh(m))</pre>
res <- gabrielneigh(m, nnmult=4)
summary(graph2nb(res))
```

gstsls

Spatial simultaneous autoregressive SAC model estimation by GMM

## Description

An implementation of Kelejian and Prucha's generalised moments estimator for the autoregressive parameter in a spatial model with a spatially lagged dependent variable.

### Usage

```
gstsls(formula, data = list(), listw, listw2 = NULL, na.action = na.fail,
  zero.policy = NULL, pars, scaleU=FALSE, control = list(),
  verbose=NULL, method="nlminb", robust=FALSE, legacy=FALSE, W2X=TRUE)
```

## **Arguments**

formula a symbolic description of the model to be fit. The details of model specification

are given for lm()

data an optional data frame containing the variables in the model. By default the

variables are taken from the environment which the function is called.

gstsls 69

listw a listw object created for example by nb2listw

listw2 a listw object created for example by nb2listw, if not given, set to the same

spatial weights as the listw argument

na.action a function (default na.fail), can also be na.omit or na.exclude with conse-

quences for residuals and fitted values - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create no-neighbour observations. Note that only weights lists created without using the glist argument to nb2listw may be

subsetted.

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged

value of zones without neighbours, if FALSE (default) assign NA - causing

GMerrorsar() to terminate with an error

pars starting values for  $\lambda$  and  $\sigma^2$  for GMM optimisation, if missing (default), ap-

proximated from initial 2sls model as the autocorrelation coefficient corrected

for weights style and model sigma squared

scaleU Default FALSE: scale the OLS residuals before computing the moment matri-

ces; only used if the pars argument is missing

control A list of control parameters. See details in optim or nlminb

verbose default NULL, use global option value; if TRUE, reports function values during

optimization.

method default nlminb, or optionally a method passed to optim to use an alternative

optimizer

robust see stsls
legacy see stsls
W2X see stsls

#### Details

When the control list is set with care, the function will converge to values close to the ML estimator without requiring computation of the Jacobian, the most resource-intensive part of ML estimation.

### Value

A list object of class gmsar

lambda simultaneous autoregressive error coefficient

coefficients GMM coefficient estimates (including the spatial autocorrelation coefficient)

rest.se GMM coefficient standard errors

SSE Sum of squared GMM errors
parameters number of parameters estimated

call the call used to create this object

70 gstsls

residuals GMM residuals

lm.target NULL

formula model formula

aliased NULL

zero.policy zero.policy for this model

LL NULL

vv list of internal bigG and litg components for testing optimisation surface

optres object returned by optimizer

pars start parameter values for optimisation

Hcov NULL

na.action (possibly) named vector of excluded or omitted observations if non-default na.action

argument used

## Author(s)

Gianfranco Piras and Roger Bivand

### References

Kelejian, H. H., and Prucha, I. R., 1999. A Generalized Moments Estimator for the Autoregressive Parameter in a Spatial Model. International Economic Review, 40, pp. 509–533; Cressie, N. A. C. 1993 *Statistics for spatial data*, Wiley, New York.

## See Also

optim, nlminb, GMerrorsar, GMargminImage

```
data(oldcol)
COL.errW.GM <- gstsls(CRIME ~ INC + HOVAL, data=COL.OLD, nb2listw(COL.nb, style="W"))
summary(COL.errW.GM)
aa <- GMargminImage(COL.errW.GM)
levs <- quantile(aa$z, seq(0, 1, 1/12))
image(aa, breaks=levs, xlab="lambda", ylab="s2")
points(COL.errW.GM$lambda, COL.errW.GM$s2, pch=3, lwd=2)
contour(aa, levels=signif(levs, 4), add=TRUE)
COL.errW.GM <- gstsls(CRIME ~ INC + HOVAL, data=COL.OLD, nb2listw(COL.nb, style="W"), scaleU=TRUE)
summary(COL.errW.GM)
listw <- nb2listw(COL.nb)
W <- as(as_dgRMatrix_listw(listw), "CsparseMatrix")
trMat <- trW(W, type="mult")
impacts(COL.errW.GM, tr=trMat)</pre>
```

hopkins 71

hopkins

Hopkins burnt savanna herb remains

## **Description**

A 20m square is divided into 40 by 40 0.5m quadrats. Observations are in tens of grams of herb remains, 0 being from 0g to less than 10g, and so on. Analysis was mostly conducted using the interior 32 by 32 grid.

# Usage

```
data(hopkins)
```

### **Format**

The format is: num [1:40, 1:40] 0 0 0 0 0 0 0 0 0 1 ...

#### Source

Upton, G., Fingleton, B. 1985 Spatial data analysis by example: point pattern and quatitative data, Wiley, pp. 38–39.

### References

Hopkins, B., 1965 Observations on savanna burning in the Olokemeji Forest Reserve, Nigeria. Journal of Applied Ecology, 2, 367–381.

# **Examples**

```
data(hopkins)
image(1:32, 1:32, hopkins[5:36,36:5], breaks=c(-0.5, 3.5, 20),
  col=c("white", "black"))
box()
```

house

Lucas county OH housing

## **Description**

Data on 25,357 single family homes sold in Lucas County, Ohio, 1993-1998 from the county auditor, together with an nb neighbour object constructed as a sphere of influence graph from projected coordinates.

#### Usage

```
data(house)
```

72 house

### **Format**

The format is: Formal class 'SpatialPointsDataFrame' [package "sp"] with 5 slots. The data slot is a data frame with 25357 observations on the following 24 variables.

```
price a numeric vector
yrbuilt a numeric vector
stories a factor with levels one bilevel multilvl one+half two two+half three
TLA a numeric vector
wall a factor with levels stucdrvt ccbtile metlvnyl brick stone wood partbrk
beds a numeric vector
baths a numeric vector
halfbaths a numeric vector
frontage a numeric vector
depth a numeric vector
garage a factor with levels no garage basement attached detached carport
garagesqft a numeric vector
rooms a numeric vector
lotsize a numeric vector
sdate a numeric vector
avalue a numeric vector
s1993 a numeric vector
s1994 a numeric vector
s1995 a numeric vector
s1996 a numeric vector
s1997 a numeric vector
s1998 a numeric vector
syear a factor with levels 1993 1994 1995 1996 1997 1998
age a numeric vector
Its projection is CRS(+init=epsg: 2834), the Ohio North State Plane.
```

### Source

Dataset included in the Spatial Econometrics toolbox for Matlab, http://www.spatial-econometrics.com/html/jplv7.zip.

huddersfield 73

## **Examples**

```
## Not run:
house <- read.table("house.dat", header=FALSE)</pre>
names(house) <- c("price", "yrbuilt", "stories", "TLA", "wall", "beds",</pre>
  "baths", "halfbaths", "frontage", "depth", "garage", "garagesqft", "rooms",
  "lotsize", "sdate", "avalue", "long", "lat", "s1993", "s1994", "s1995",
  "s1996", "s1997", "s1998")
house$syear <- 1992 + house$s1993 + 2*house$s1994 + 3*house$s1995 +
4*house$s1996 + 5*house$s1997 + 6*house$s1998
house$syear <- factor(house$syear)</pre>
house$age <- (1999 - house$yrbuilt)/100</pre>
house$stories <- factor(house$stories, levels=1:7, labels=c("one",
 "bilevel", "multilvl", "one+half", "two", "two+half", "three"))
house$wall <- factor(house$wall, levels=1:7, labels=c("stucdrvt",
 "ccbtile", "metlvnyl", "brick", "stone", "wood", "partbrk"))
house$garage <- factor(house$garage, levels=0:4, labels=c("no garage",
 "basement", "attached", "detached", "carport"))
library(sp)
coordinates(house) <- c("long", "lat")</pre>
proj4string(house) <- CRS("+proj=longlat")</pre>
library(rgdal)
house <- spTransform(house, CRS("+init=epsg:2834"))</pre>
library(spdep)
LO_nb <- graph2nb(soi.graph(tri2nb(coordinates(house)), coordinates(house)))
W <- as(as_dgRMatrix_listw(nb2listw(LO_nb)), "CsparseMatrix")</pre>
trMat <- trW(W, type="mult")</pre>
## End(Not run)
data(house)
## maybe str(house) ; plot(house) ...
```

huddersfield

Prevalence of respiratory symptoms

## **Description**

Prevalence of respiratory symptoms in 71 school catchment areas in Huddersfield, Northern England

#### Usage

```
data(huddersfield)
```

### **Format**

A data frame with 71 observations on the following 2 variables.

cases Prevalence of at least mild conditions

total Number of questionnaires returned

### **Source**

Martuzzi M, Elliott P (1996) Empirical Bayes estimation of small area prevalence of non-rare conditions, Statistics in Medicine 15, 1867–1873, pp. 1870–1871.

## **Examples**

```
data(huddersfield)
str(huddersfield)
```

impacts

Impacts in spatial lag models

## **Description**

The calculation of impacts for spatial lag and spatial Durbin models is needed in order to interpret the regression coefficients correctly, because of the spillovers between the terms in these data generation processes (unlike the spatial error model).

## Usage

```
## S3 method for class 'sarlm'
impacts(obj, ..., tr, R = NULL, listw = NULL, useHESS = NULL,
tol = 1e-06, empirical = FALSE, Q=NULL)
## S3 method for class 'stsls'
impacts(obj, ..., tr, R = NULL, listw = NULL,
 tol = 1e-06, empirical = FALSE, Q=NULL)
## S3 method for class 'gmsar'
impacts(obj, ..., tr, R = NULL, listw = NULL,
tol = 1e-06, empirical = FALSE, Q=NULL)
## S3 method for class 'lagImpact'
plot(x, ..., choice="direct", trace=FALSE, density=TRUE)
## S3 method for class 'lagImpact'
print(x, ..., reportQ=NULL)
## S3 method for class 'lagImpact'
summary(object, ..., zstats=FALSE, short=FALSE, reportQ=NULL)
## S3 method for class 'lagImpact'
HPDinterval(obj, prob = 0.95, ..., choice="direct")
```

## Arguments

obj	A sarIm spatial regression object created by lagsarIm; in HPDinterval.lagImpact a lagImpact object
	Arguments passed through to methods in the coda package
tr	A vector of traces of powers of the spatial weights matrix created using trW, for approximate impact measures; if not given, listw must be given for exact measures (for small to moderate spatial weights matrices); the traces must be for the same spatial weights as were used in fitting the spatial regression

R	If given, simulations are used to compute distributions for the impact measures, returned as mcmc objects; the objects are used for convenience but are not output by an MCMC process
listw	If tr is not given, a spatial weights object as created by nb2listw; they must be the same spatial weights as were used in fitting the spatial regression
useHESS	Use the Hessian approximation (if available) even if the asymptotic coefficient covariance matrix is available; used for comparing methods
tol	Argument passed to myrnorm: tolerance (relative to largest variance) for numerical lack of positive-definiteness in the coefficient covariance matrix
empirical	Argument passed to myrnorm (default FALSE): if true, the coefficients and their covariance matrix specify the empirical not population mean and covariance matrix
Q	default NULL, else an integer number of cumulative power series impacts to calculate if tr is given
reportQ	default NULL; if TRUE and $\ensuremath{\mathtt{Q}}$ given as an argument to impacts, report impact components
x, object	lagImpact objects created by impacts methods
zstats	default FALSE, if TRUE, also return z-values and p-values for the impacts based on the simulations
short	default FALSE, if TRUE passed to the print summary method to omit printing of the mcmc summaries
choice	One of three impacts: direct, indirect, or total
trace	Argument passed to plot.mcmc: plot trace plots
density	Argument passed to plot.mcmc: plot density plots
prob	Argument passed to HPDinterval.mcmc: a numeric scalar in the interval $(0,1)$ giving the target probability content of the intervals

## **Details**

If called without R being set, the method returns the direct, indirect and total impacts for the variables in the model, for the variables themselves in tha spatial lag model case, for the variables and their spatial lags in the spatial Durbin (mixed) model case. The spatial lag impact measures are computed using eq. 2.46 (LeSage and Pace, 2009, p. 38), either using the exact dense matrix (when listw is given), or traces of powers of the weights matrix (when tr is given). When the traces are created by powering sparse matrices, the exact and the trace methods should give very similar results, unless the number of powers used is very small.

If R is given, simulations will be used to create distributions for the impact measures, provided that the fitted model object contains a coefficient covariance matrix. The simulations are made using myrnorm with the coefficients and their covariance matrix from the fitted model.

The simulations are stored as mcmc objects as defined in the **coda** package; the objects are used for convenience but are not output by an MCMC process. The simulated values of the coefficients are checked to see that the spatial coefficient remains within its valid interval — draws outside the interval are discarded.

When Q and tr are given, addition impact component results are provided for each step in the traces of powers of the weights matrix up to and including the Q'th power. This increases computing time because the output object is substantially increased in size in proportion to the size of Q.

The method for gmsar objects is only for those of type SARAR output by gstsls, and assume that the spatial error coefficient is fixed, and thus omitted from the coefficients and covariance matrix used for simulation.

#### Value

An object of class lagImpact.

If no simulation is carried out, the object returned is a list with:

direct numeric vector indirect numeric vector total numeric vector

and a matching Qres list attribute if Q was given.

If simulation is carried out, the object returned is a list with:

res a list with three components as for the non-simulation case, with a matching

Qres list attribute if Q was given

sres a list with three mcmc matrices, for the direct, indirect and total impacts with a

matching Qmcmc list attribute if Q was given

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

LeSage J and RK Pace (2009) Introduction to Spatial Econometrics. CRC Press, Boca Raton, pp. 33–42, 114–115; LeSage J and MM Fischer (2008) Spatial growth regressions: model specification, estimation and interpretation. Spatial Economic Analysis 3 (3), pp. 275–304.

## See Also

```
trW, lagsarlm, nb2listw, mvrnorm, plot.mcmc, summary.mcmc, HPDinterval
```

```
example(columbus)
listw <- nb2listw(col.gal.nb)
lobj <- lagsarlm(CRIME ~ INC + HOVAL, columbus, listw)
summary(lobj)
mobj <- lagsarlm(CRIME ~ INC + HOVAL, columbus, listw, type="mixed")
summary(mobj)
W <- as(as_dgRMatrix_listw(listw), "CsparseMatrix")
trMatc <- trW(W, type="mult")
trMC <- trW(W, type="MC")</pre>
```

```
impacts(lobj, listw=listw)
impacts(lobj, tr=trMatc)
impacts(lobj, tr=trMC)
lobj1 <- stsls(CRIME ~ INC + HOVAL, columbus, listw)</pre>
loobj1 <- impacts(lobj1, tr=trMatc, R=200)</pre>
summary(loobj1, zstats=TRUE, short=TRUE)
lobj1r <- stsls(CRIME ~ INC + HOVAL, columbus, listw, robust=TRUE)</pre>
loobj1r <- impacts(lobj1r, tr=trMatc, R=200)</pre>
summary(loobj1r, zstats=TRUE, short=TRUE)
lobjIQ5 <- impacts(lobj, tr=trMatc, R=200, Q=5)</pre>
summary(lobjIQ5, zstats=TRUE, short=TRUE)
summary(lobjIQ5, zstats=TRUE, short=TRUE, reportQ=TRUE)
impacts(mobj, listw=listw)
impacts(mobj, tr=trMatc)
impacts(mobj, tr=trMC)
summary(impacts(mobj, tr=trMatc, R=200), zstats=TRUE)
## Not run:
mobj1 <- lagsarlm(CRIME ~ INC + HOVAL, columbus, listw, type="mixed",</pre>
method="Matrix", fdHess=TRUE)
summary(mobj1)
summary(impacts(mobj1, tr=trMatc, R=1000), zstats=TRUE, short=TRUE)
summary(impacts(mobj, tr=trMatc, R=1000), zstats=TRUE, short=TRUE)
mobj2 <- lagsarlm(CRIME ~ INC + HOVAL, columbus, listw, type="mixed",</pre>
method="Matrix", fdHess=TRUE, optimHess=TRUE)
summary(impacts(mobj2, tr=trMatc, R=1000), zstats=TRUE, short=TRUE)
\dontrun{
mobj3 <- lagsarlm(CRIME ~ INC + HOVAL, columbus, listw, type="mixed",</pre>
method="spam", fdHess=TRUE)
summary(impacts(mobj3, tr=trMatc, R=1000), zstats=TRUE, short=TRUE)
}
data(boston)
Wb <- as(as_dgRMatrix_listw(nb2listw(boston.soi)), "CsparseMatrix")</pre>
trMatb <- trW(Wb, type="mult")</pre>
gp2mMi <- lagsarlm(log(CMEDV) ~ CRIM + ZN + INDUS + CHAS + I(NOX^2) +</pre>
I(RM^2) + AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
data=boston.c, nb2listw(boston.soi), type="mixed", method="Matrix",
fdHess=TRUE, trs=trMatb)
summary(gp2mMi)
summary(impacts(gp2mMi, tr=trMatb, R=1000), zstats=TRUE, short=TRUE)
data(house)
lw <- nb2listw(LO_nb)</pre>
form <- formula(log(price) ~ age + I(age^2) + I(age^3) + log(lotsize) +</pre>
   rooms + log(TLA) + beds + syear)
lobj <- lagsarlm(form, house, lw, method="Matrix",</pre>
fdHess=TRUE, trs=trMat)
summary(lobj)
loobj <- impacts(lobj, tr=trMat, R=1000)</pre>
summary(loobj, zstats=TRUE, short=TRUE)
lobj1 <- stsls(form, house, lw)</pre>
loobj1 <- impacts(lobj1, tr=trMat, R=1000)</pre>
summary(loobj1, zstats=TRUE, short=TRUE)
mobj <- lagsarlm(form, house, lw, type="mixed",</pre>
method="Matrix", fdHess=TRUE, trs=trMat)
```

78 include.self

```
summary(mobj)
moobj <- impacts(mobj, tr=trMat, R=1000)
summary(moobj, zstats=TRUE, short=TRUE)
## End(Not run)</pre>
```

include.self

Include self in neighbours list

## **Description**

The function includes the region itself in its own list of neighbours, and sets attribute "self.included" to TRUE.

# Usage

```
include.self(nb)
```

## **Arguments**

nb

input neighbours list of class nb

# Value

The function returns an object of class nb with a list of integer vectors containing neighbour region number ids; attribute "self.included" is set to TRUE.

## Author(s)

```
Roger Bivand < Roger . Bivand@nhh . no >
```

# See Also

```
summary.nb
```

```
example(columbus)
coords <- coordinates(columbus)
summary(col.gal.nb, coords)
summary(include.self(col.gal.nb), coords)</pre>
```

invIrM 79

invIrM

Compute SAR generating operator

## **Description**

Computes the matrix used for generating simultaneous autoregressive random variables, for a given value of rho, a neighbours list object, a chosen coding scheme style, and optionally a list of general weights corresponding to neighbours.

## Usage

```
invIrM(neighbours, rho, glist=NULL, style="W", method="solve",
  feasible=NULL)
invIrW(listw, rho, method="solve", feasible=NULL)
powerWeights(W, rho, order=250, X, tol=.Machine$double.eps^(3/5))
```

## **Arguments**

neighbours	an object of class nb
rho	autoregressive parameter
glist	list of general weights corresponding to neighbours
style	style can take values W, B, C, and S
method	default solve, can also take value chol
feasible	if NULL, the given value of rho is checked to see if it lies within its feasible range, if TRUE, the test is not conducted
listw	a listw object from for example nb2listw
W	A spatial weights matrix (either a dense matrix or a CsparseMatrix)
order	Power series maximum limit
Χ	A numerical matrix
tol	Tolerance for convergence of power series

## **Details**

The invIrW function generates the full weights matrix V, checks that rho lies in its feasible range between  $1/\min(eigen(V))$  and  $1/\max(eigen(V))$ , and returns the nxn inverted matrix

$$(I - \rho V)^{-1}$$

. With method="chol", Cholesky decomposition is used, thanks to contributed code by Markus Reder and Werner Mueller.

The powerWeights function uses power series summation to cumulate the product

$$(I - \rho V)^{-1}\% * \%X$$

from

$$(I + \rho V + (\rho V)^2 + \ldots)\% * \%X$$

, which can be done by storing only sparse V and several matrices of the same dimensions as X. This makes it possible to handle larger spatial weights matrices.

80 invIrM

#### Value

An nxn matrix with a "call" attribute; the powerWeights function returns a matrix of the same dimensions as X which has been multipled by the power series equivalent of the dense matrix

$$(I - \rho V)^{-1}$$

.

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Tiefelsdorf, M., Griffith, D. A., Boots, B. 1999 A variance-stabilizing coding scheme for spatial link matrices, Environment and Planning A, 31, pp. 165-180; Tiefelsdorf, M. 2000 Modelling spatial processes, Lecture notes in earth sciences, Springer, p. 76; Haining, R. 1990 Spatial data analysis in the social and environmental sciences, Cambridge University Press, p. 117; Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 152; Reder, M. and Mueller, W. (2007) An Improvement of the invIrM Routine of the Geostatistical R-package spdep by Cholesky Inversion, Statistical Projects, LV No: 238.205, SS 2006, Department of Applied Statistics, Johannes Kepler University, Linz

## See Also

nb2listw

```
nb7rt <- cell2nb(7, 7, torus=TRUE)</pre>
set.seed(1)
x <- matrix(rnorm(500*length(nb7rt)), nrow=length(nb7rt))</pre>
res0 <- apply(invIrM(nb7rt, rho=0.0, method="chol".
feasible=TRUE) %*% x, 2, function(x) var(x)/length(x))
res2 <- apply(invIrM(nb7rt, rho=0.2, method="chol",
feasible=TRUE) %*% x, 2, function(x) var(x)/length(x))
res4 <- apply(invIrM(nb7rt, rho=0.4, method="chol",
feasible=TRUE) %*% x, 2, function(x) var(x)/length(x))
res6 <- apply(invIrM(nb7rt, rho=0.6, method="chol",
feasible=TRUE) %*% x, 2, function(x) var(x)/length(x))
res8 <- apply(invIrM(nb7rt, rho=0.8, method="chol",
feasible=TRUE) %*% x, 2, function(x) var(x)/length(x))
res9 <- apply(invIrM(nb7rt, rho=0.9, method="chol",
feasible=TRUE) %*% x, 2, function(x) var(x)/length(x))
plot(density(res9), col="red", xlim=c(-0.01, max(density(res9)$x)),
 ylim=range(density(res0)$y),
 xlab="estimated variance of the mean",
 main=expression(paste("Effects of spatial autocorrelation for different ",
    rho, " values")))
lines(density(res0), col="black")
lines(density(res2), col="brown")
```

is.symmetric.nb 81

```
lines(density(res4), col="green")
lines(density(res6), col="orange")
lines(density(res8), col="pink")
legend(c(-0.02, 0.01), c(7, 25),
legend=c("0.0", "0.2", "0.4", "0.6", "0.8", "0.9"),
 col=c("black", "brown", "green", "orange", "pink", "red"), lty=1, bty="n")
## Not run:
x <- matrix(rnorm(length(nb7rt)), ncol=1)</pre>
system.time(e <- invIrM(nb7rt, rho=0.9, method="chol", feasible=TRUE) %*% x)
system.time(e <- invIrM(nb7rt, rho=0.9, method="chol", feasible=NULL) %*% x)</pre>
system.time(e <- invIrM(nb7rt, rho=0.9, method="solve", feasible=TRUE) %*% x)</pre>
system.time(e <- invIrM(nb7rt, rho=0.9, method="solve", feasible=NULL) %*% x)
W <- as(as_dgRMatrix_listw(nb2listw(nb7rt)), "CsparseMatrix")</pre>
system.time(ee <- powerWeights(W, rho=0.9, X=x))</pre>
all.equal(e, as(ee, "matrix"), check.attributes=FALSE)
nb60rt <- cell2nb(60, 60, torus=TRUE)</pre>
W <- as(as_dgRMatrix_listw(nb2listw(nb60rt)), "CsparseMatrix")</pre>
set.seed(1)
x <- matrix(rnorm(dim(W)[1]), ncol=1)</pre>
system.time(ee <- powerWeights(W, rho=0.3, X=x))</pre>
str(as(ee, "matrix"))
obj <- errorsarlm(as(ee, "matrix")[,1] ~ 1, listw=nb2listw(nb60rt), method="Matrix")
coefficients(obj)
## End(Not run)
```

is.symmetric.nb

Test a neighbours list for symmetry

## **Description**

Checks a neighbours list for symmetry/transitivity (if i is a neighbour of j, then j is a neighbour of i). This holds for distance and contiguity based neighbours, but not for k-nearest neighbours. The helper function sym.attr.nb() calls is.symmetric.nb() to set the sym attribute if needed, and make.sym.nb makes a non-symmetric list symmetric by adding neighbors. is.symmetric.glist checks a list of general weights corresponding to neighbours for symmetry for symmetric neighbours.

## Usage

```
is.symmetric.nb(nb, verbose = NULL, force = FALSE)
sym.attr.nb(nb)
make.sym.nb(nb)
old.make.sym.nb(nb)
is.symmetric.glist(nb, glist)
```

82 joincount.mc

## Arguments

nb	an object of class nb with a list of integer vectors containing neighbour region number ids.
verbose	default NULL, use global option value; if TRUE prints non-matching pairs
force	do not respect a neighbours list sym attribute and test anyway
glist	list of general weights corresponding to neighbours

## Value

TRUE if symmetric, FALSE if not; is.symmetric.glist returns a value with an attribute, "d", indicating for failed symmetry the largest failing value.

#### Note

A new version of make.sym.nb by Bjarke Christensen is now included. The older version has been renamed old.make.sym.nb, and their comparison constitutes a nice demonstration of vectorising speedup using sapply and lapply rather than loops. When any no-neighbour observations are present, old.make.sym.nb is used.

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## See Also

```
read.gal
```

## **Examples**

```
example(columbus)
coords <- coordinates(columbus)
ind <- sapply(slot(columbus, "polygons"), function(x) slot(x, "ID"))
print(is.symmetric.nb(col.gal.nb, verbose=TRUE, force=TRUE))
k4 <- knn2nb(knearneigh(coords, k=4), row.names=ind)
k4 <- sym.attr.nb(k4)
print(is.symmetric.nb(k4))
k4.sym <- make.sym.nb(k4)
print(is.symmetric.nb(k4.sym))</pre>
```

joincount.mc

Permutation test for same colour join count statistics

# Description

A permutation test for same colour join count statistics calculated by using nsim random permutations of fx for the given spatial weighting scheme, to establish the ranks of the observed statistics (for each colour) in relation to the nsim simulated values.

joincount.mc 83

## Usage

```
joincount.mc(fx, listw, nsim, zero.policy=FALSE, alternative="greater",
    spChk=NULL)
```

### **Arguments**

fx a factor of the same length as the neighbours and weights objects in listw

listw a listw object created for example by nb2listw

nsim number of permutations

zero.policy if TRUE assign zero to the lagged value of zones without neighbours, if FALSE

assign NA

alternative a character string specifying the alternative hypothesis, must be one of "greater"

(default), or "less".

spChk should the data vector names be checked against the spatial objects for identity

integrity, TRUE, or FALSE, default NULL to use get.spChkOption()

#### Value

A list with class jclist of lists with class htest and mc.sim for each of the k colours containing the following components:

statistic the value of the observed statistic.

parameter the rank of the observed statistic.

method a character string giving the method used.

data. name a character string giving the name(s) of the data.

p. value the pseudo p-value of the test.

alternative a character string describing the alternative hypothesis.

estimate the mean and variance of the simulated distribution.

res nsim simulated values of statistic, the final element is the observed statistic

# Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 63-5.

## See Also

```
joincount.test
```

84 joincount.multi

## **Examples**

```
data(oldcol)
HICRIME <- cut(COL.OLD$CRIME, breaks=c(0,35,80), labels=c("low","high"))
names(HICRIME) <- rownames(COL.OLD)
joincount.mc(HICRIME, nb2listw(COL.nb, style="B"), nsim=99)
joincount.test(HICRIME, nb2listw(COL.nb, style="B"))</pre>
```

joincount.multi

BB, BW and Jtot join count statistic for k-coloured factors

# Description

A function for tallying join counts between same-colour and different colour spatial objects, where neighbour relations are defined by a weights list. Given the global counts in each colour, expected counts and variances are calculated under non-free sampling, and a z-value reported. Since multiple tests are reported, no p-values are given, allowing the user to adjust the significance level applied. Jtot is the count of all different-colour joins.

## Usage

```
joincount.multi(fx, listw, zero.policy = FALSE,
  spChk = NULL, adjust.n=TRUE)
## S3 method for class 'jcmulti'
print(x, ...)
```

## **Arguments**

fx	a factor of the same length as the neighbours and weights objects in listw
listw	a listw object created for example by nb2listw
zero.policy	if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign $\ensuremath{NA}$
adjust.n	default TRUE, if FALSE the number of observations is not adjusted for noneighbour observations, if TRUE, the number of observations is adjusted consistently (up to and including spdep 0.3-28 the adjustment was inconsistent thanks to Tomoki NAKAYA for a careful bug report)
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
x	object to be printed
	arguments to be passed through for printing

#### Value

A matrix with class jcmulti with row and column names for observed and expected counts, variance, and z-value.

joincount.multi 85

#### Note

The derivation of the test (Cliff and Ord, 1981, p. 18) assumes that the weights matrix is symmetric. For inherently non-symmetric matrices, such as k-nearest neighbour matrices, listw2U() can be used to make the matrix symmetric. In non-symmetric weights matrix cases, the variance of the test statistic may be negative.

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 20; Upton, G., Fingleton, B. 1985 Spatial data analysis by example: point pattern and quatitative data, Wiley, pp. 158–170.

#### See Also

```
joincount.test
```

```
data(oldcol)
HICRIME <- cut(COL.OLD$CRIME, breaks=c(0,35,80), labels=c("low","high"))</pre>
names(HICRIME) <- rownames(COL.OLD)</pre>
joincount.multi(HICRIME, nb2listw(COL.nb, style="B"))
## Not run:
data(hopkins)
image(1:32, 1:32, hopkins[5:36,36:5], breaks=c(-0.5, 3.5, 20),
col=c("white", "black"))
box()
hopkins.rook.nb <- cell2nb(32, 32, type="rook")
unlist(spweights.constants(nb2listw(hopkins.rook.nb, style="B")))
hopkins.queen.nb <- cell2nb(32, 32, type="queen")</pre>
hopkins.bishop.nb <- diffnb(hopkins.rook.nb, hopkins.queen.nb, verbose=FALSE)
hopkins4 <- hopkins[5:36,36:5]
hopkins4[which(hopkins4 > 3, arr.ind=TRUE)] <- 4
hopkins4.f <- factor(hopkins4)</pre>
table(hopkins4.f)
joincount.multi(hopkins4.f, nb2listw(hopkins.rook.nb, style="B"))
cat("replicates Upton & Fingleton table 3.4 (p. 166)\n")
joincount.multi(hopkins4.f, nb2listw(hopkins.bishop.nb, style="B"))
cat("replicates Upton & Fingleton table 3.6 (p. 168)\n")
joincount.multi(hopkins4.f, nb2listw(hopkins.queen.nb, style="B"))
cat("replicates Upton & Fingleton table 3.7 (p. 169)\n")
## End(Not run)
```

86 joincount.test

joincount.test BB join count statistic for k-coloured factors

## **Description**

The BB join count test for spatial autocorrelation using a spatial weights matrix in weights list form for testing whether same-colour joins occur more frequently than would be expected if the zones were labelled in a spatially random way. The assumptions underlying the test are sensitive to the form of the graph of neighbour relationships and other factors, and results may be checked against those of joincount.mc permutations.

## Usage

```
joincount.test(fx, listw, zero.policy=NULL, alternative="greater", spChk=NULL, adjust.n=TRUE) ## S3 method for class 'jclist' print(x, ...)
```

## **Arguments**

fx	a factor of the same length as the neighbours and weights objects in listw
listw	a listw object created for example by nb2listw
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
alternative	a character string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".
adjust.n	default TRUE, if FALSE the number of observations is not adjusted for noneighbour observations, if TRUE, the number of observations is adjusted consistently (up to and including spdep 0.3-28 the adjustment was inconsistent - thanks to Tomoki NAKAYA for a careful bug report)
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
X	object to be printed
	arguments to be passed through for printing

## Value

A list with class jclist of lists with class htest for each of the k colours containing the following components:

statistic the value of the standard deviate of the join count statistic.

p. value the p-value of the test.

estimate the value of the observed statistic, its expectation and variance under non-free

sampling.

knearneigh 87

alternative a character string describing the alternative hypothesis.

method a character string giving the method used.

data.name a character string giving the name(s) of the data.

#### Note

The derivation of the test (Cliff and Ord, 1981, p. 18) assumes that the weights matrix is symmetric. For inherently non-symmetric matrices, such as k-nearest neighbour matrices, listw2U() can be used to make the matrix symmetric. In non-symmetric weights matrix cases, the variance of the test statistic may be negative.

### Author(s)

```
Roger Bivand < Roger . Bivand@nhh . no>
```

#### References

```
Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 20.
```

#### See Also

```
joincount.mc, joincount.multi, listw2U
```

### **Examples**

```
data(oldcol)
HICRIME <- cut(COL.OLD$CRIME, breaks=c(0,35,80), labels=c("low","high"))
names(HICRIME) <- rownames(COL.OLD)
joincount.test(HICRIME, nb2listw(COL.nb, style="B"))
joincount.test(HICRIME, nb2listw(COL.nb, style="C"))
joincount.test(HICRIME, nb2listw(COL.nb, style="S"))
joincount.test(HICRIME, nb2listw(COL.nb, style="S"))
joincount.test(HICRIME, nb2listw(COL.nb, style="W"))
by(card(COL.nb), HICRIME, summary)
print(is.symmetric.nb(COL.nb))
coords.OLD <- cbind(COL.OLD$X, COL.OLD$Y)
COL.k4.nb <- knn2nb(knearneigh(coords.OLD, 4))
print(is.symmetric.nb(COL.k4.nb))
joincount.test(HICRIME, nb2listw(COL.k4.nb, style="B"))
cat("Note non-symmetric weights matrix - use listw2U()\n")
joincount.test(HICRIME, listw2U(nb2listw(COL.k4.nb, style="B")))</pre>
```

knearneigh

K nearest neighbours for spatial weights

## **Description**

The function returns a matrix with the indices of regions belonging to the set of the k nearest neighbours of each other. If longlat = TRUE, Great Circle distances are used.

88 knearneigh

## Usage

```
knearneigh(x, k=1, longlat = NULL, RANN=TRUE)
```

## **Arguments**

x matrix of region point coordinates or a SpatialPoints object

k number of nearest neighbours to be returned

longlat TRUE if point coordinates are longitude-latitude decimal degrees, in which case

distances are measured in kilometers; if x is a SpatialPoints object, the value is

taken from the object itself

RANN logical value, if the RANN package is available, use for finding k nearest neigh-

bours when longlat is FALSE

## **Details**

The underlying C code is based on the knn function in the class package in the VR bundle.

#### Value

A list of class knn

nn integer matrix of region number ids

np number of input points

k input required k

dimension number of columns of x

x input coordinates

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## See Also

knn, dnearneigh, knn2nb, nn2

```
example(columbus)
coords <- coordinates(columbus)
col.knn <- knearneigh(coords, k=4)
plot(columbus, border="grey")
plot(knn2nb(col.knn), coords, add=TRUE)
title(main="K nearest neighbours, k = 4")
data(state)
us48.fipsno <- read.geoda(system.file("etc/weights/us48.txt",
    package="spdep")[1])
if (as.numeric(paste(version$major, version$minor, sep="")) < 19) {
    m50.48 <- match(us48.fipsno$"State.name", state.name)</pre>
```

knn2nb 89

```
} else {
    m50.48 <- match(us48.fipsno$"State_name", state.name)
}

xy <- as.matrix(as.data.frame(state.center))[m50.48,]

llk4.nb <- knn2nb(knearneigh(xy, k=4, longlat=FALSE))

gck4.nb <- knn2nb(knearneigh(xy, k=4, longlat=TRUE))

plot(llk4.nb, xy)

plot(diffnb(llk4.nb, gck4.nb), xy, add=TRUE, col="red", lty=2)

title(main="Differences between Euclidean and Great Circle k=4 neighbours")

summary(llk4.nb, xy, longlat=TRUE)

xy1 <- SpatialPoints((as.data.frame(state.center))[m50.48,],
    proj4string=CRS("+proj=longlat"))

gck4a.nb <- knn2nb(knearneigh(xy1, k=4))

summary(gck4a.nb, xy1)</pre>
```

knn2nb

Neighbours list from knn object

## **Description**

The function converts a knn object returned by knearneigh into a neighbours list of class nb with a list of integer vectors containing neighbour region number ids.

## Usage

```
knn2nb(knn, row.names = NULL, sym = FALSE)
```

## **Arguments**

knn A knn object returned by knearneigh

row.names character vector of region ids to be added to the neighbours list as attribute

region.id, default seq(1, nrow(x))

sym force the output neighbours list to symmetry

## Value

The function returns an object of class nb with a list of integer vectors containing neighbour region number ids. See card for details of "nb" objects.

## Author(s)

Roger Bivand < Roger . Bivand @nhh . no >

#### See Also

knearneigh, card

90 lag.listw

## **Examples**

```
example(columbus)
coords <- coordinates(columbus)
col.knn <- knearneigh(coords, k=4)
plot(columbus, border="grey")
plot(knn2nb(col.knn), coords, add=TRUE)
title(main="K nearest neighbours, k = 4")</pre>
```

lag.listw

Spatial lag of a numeric vector

## **Description**

Using a listw sparse representation of a spatial weights matrix, compute the lag vector Vx

### Usage

```
## S3 method for class 'listw'
lag(x, var, zero.policy=NULL, NAOK=FALSE, ...)
```

## **Arguments**

x a listw object created for example by nb2listw

var a numeric vector the same length as the neighbours list in listw

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged value

of zones without neighbours, if FALSE assign NA

NAOK If 'FALSE', the presence of 'NA' values is regarded as an error; if 'TRUE' then

any 'NA' or 'NaN' or 'Inf' values in var are represented as an NA lagged value.

... additional arguments

### Value

a numeric vector the same length as var

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

nb2listw

lagmess 91

## **Examples**

```
data(oldcol)
Vx <- lag.listw(nb2listw(COL.nb, style="W"), COL.OLD$CRIME)
plot(Vx, COL.OLD$CRIME)
plot(ecdf(COL.OLD$CRIME))
plot(ecdf(Vx), add=TRUE, col.points="red", col.hor="red")
is.na(COL.OLD$CRIME[5]) <- TRUE
VxNA <- lag.listw(nb2listw(COL.nb, style="W"), COL.OLD$CRIME, NAOK=TRUE)</pre>
```

lagmess

Matrix exponential spatial lag model

## **Description**

The function fits a matrix exponential spatial lag model, using optim to find the value of alpha, the spatial coefficient.

## Usage

```
lagmess(formula, data = list(), listw, zero.policy = NULL, na.action = na.fail,
q = 10, start = -2.5, control=list(), method="BFGS", verbose=NULL)
## S3 method for class 'lagmess'
summary(object, ...)
## S3 method for class 'lagmess'
print(x, ...)
## S3 method for class 'summary.lagmess'
print(x, digits = max(5, .0ptions digits - 3),
    signif.stars = FALSE, ...)
## S3 method for class 'lagmess'
residuals(object, ...)
## S3 method for class 'lagmess'
deviance(object, ...)
## S3 method for class 'lagmess'
coef(object, ...)
## S3 method for class 'lagmess'
fitted(object, ...)
## S3 method for class 'lagmess'
logLik(object, ...)
```

## **Arguments**

formula	a symbolic description of the model to be fit. The details of model specification are given for lm()
data	an optional data frame containing the variables in the model. By default the variables are taken from the environment which the function is called.
listw	a listw object created for example by nb2listw

92 lagmess

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged value

of zones without neighbours, if FALSE assign NA - causing lagmess() to ter-

minate with an error

na.action a function (default options ("na.action")), can also be na.omit or na.exclude

with consequences for residuals and fitted values - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create no-neighbour observations. Note that only weights lists created without using the glist argument to

nb2listw may be subsetted.

q default 10; number of powers of the spatial weights to use

start starting value for numerical optimization, should be a small negative number

control control parameters passed to optim
method default BFGS, method passed to optim

verbose default NULL, use global option value; if TRUE report function values during

optimization

x, object Objects of classes lagmess or summary. lagmess to be passed to methods

digits the number of significant digits to use when printing

signif.stars logical. If TRUE, "significance stars" are printed for each coefficient.

.. further arguments passed to or from other methods

#### **Details**

The underlying spatial lag model:

$$y = \rho W y + X \beta + \varepsilon$$

where  $\rho$  is the spatial parameter may be fitted by maximum likelihood. In that case, the log likelihood function includes the logartithm of cumbersome Jacobian term  $|I - \rho W|$ . If we rewrite the model as:

$$Sy = X\beta + \varepsilon$$

we see that in the ML case  $Sy=(I-\rho W)y$ . If W is row-stochastic, S may be expressed as a linear combination of row-stochastic matrices. By pre-computing the matrix  $[yWy,W^2y,...,W^{q-1}y]$ , the term  $Sy(\alpha)$  can readily be found by numerical optimization using the matrix exponential approach.  $\alpha$  and  $\rho$  are related as  $\rho=1-\exp\alpha$ , conditional on the number of matrix power terms taken q.

#### Value

The function returns an object of class lagmess with components:

lmobj the lm object returned after fitting alpha

alpha the spatial coefficient

alphase the standard error of the spatial coefficient using the numerical Hessian

rho the value of rho implied by alpha

lagmess 93

bestmess	the object returned by optim
q	the number of powers of the spatial weights used
start	the starting value for numerical optimization used
na.action	(possibly) named vector of excluded or omitted observations if non-default na. action argument used $$
nullLL	the log likelihood of the aspatial model for the same data

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no > and Eric Blankmeyer

#### References

J. P. LeSage and R. K. Pace (2007) A matrix exponential specification. Journal of Econometrics, 140, 190-214; J. P. LeSage and R. K. Pace (2009) Introduction to Spatial Econometrics. CRC Press, Chapter 9.

#### See Also

lagsarlm, optim

```
data(baltimore)
baltimore$AGE <- ifelse(baltimore$AGE < 1, 1, baltimore$AGE)</pre>
lw <- nb2listw(knn2nb(knearneigh(cbind(baltimore$X, baltimore$Y), k=7)))</pre>
obj1 \leftarrow lm(log(PRICE) \sim PATIO + log(AGE) + log(SQFT) + lag(lw, log(AGE)),
data=baltimore)
lm.morantest(obj1, lw)
lm.LMtests(obj1, lw, test="all")
obj2 <- lagmess(log(PRICE) ~ PATIO + log(AGE) + log(SQFT) +
lag(lw, log(AGE)), data=baltimore, listw=lw)
summary(obj2)
obj3 <- lagsarlm(log(PRICE) ~ PATIO + log(AGE) + log(SQFT) +
lag(lw, log(AGE)), data=baltimore, listw=lw)
summary(obj3)
data(boston)
lw <- nb2listw(boston.soi)</pre>
gp2 \leftarrow lagsarlm(log(CMEDV) \sim CRIM + ZN + INDUS + CHAS + I(NOX^2) + I(RM^2)
+ AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
 data=boston.c, lw, method="Matrix")
summary(gp2)
gp2a \leftarrow lagmess(CMEDV \sim CRIM + ZN + INDUS + CHAS + I(NOX^2) + I(RM^2)
 + AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
 data=boston.c, lw)
summary(gp2a)
```

lagsarlm Spatial simultaneous autoregressive lag model estimation
---

## **Description**

Maximum likelihood estimation of spatial simultaneous autoregressive lag and spatial Durbin (mixed) models of the form:

$$y = \rho W y + X \beta + \varepsilon$$

where  $\rho$  is found by optimize() first, and  $\beta$  and other parameters by generalized least squares subsequently (one-dimensional search using optim performs badly on some platforms). In the spatial Durbin (mixed) model, the spatially lagged independent variables are added to X. Note that interpretation of the fitted coefficients should use impact measures, because of the feedback loops induced by the data generation process for this model. With one of the sparse matrix methods, larger numbers of observations can be handled, but the interval= argument may need be set when the weights are not row-standardised.

## Usage

```
lagsarlm(formula, data = list(), listw,
na.action, type="lag", method="eigen", quiet=NULL,
zero.policy=NULL, interval=NULL, tol.solve=1.0e-10, trs=NULL,
control=list())
```

## **Arguments**

formula	a symbolic description of the model to be fit. The details of model specification are given for $lm()$
data	an optional data frame containing the variables in the model. By default the variables are taken from the environment which the function is called.
listw	a listw object created for example by nb2listw
na.action	a function (default options ("na.action")), can also be na.omit or na.exclude with consequences for residuals and fitted values - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create no-neighbour observations. Note that only weights lists created without using the glist argument to nb2listw may be subsetted.
type	default "lag", may be set to "mixed"; when "mixed", the lagged intercept is dropped for spatial weights style "W", that is row-standardised weights, but otherwise included
method	"eigen" (default) - the Jacobian is computed as the product of (1 - rho*eigenvalue) using eigenw, and "spam" or "Matrix_J" for strictly symmetric weights lists of styles "B" and "C", or made symmetric by similarity (Ord, 1975, Appendix C) if possible for styles "W" and "S", using code from the spam or Matrix

> packages to calculate the determinant; "Matrix" and "spam\_update" provide updating Cholesky decomposition methods; "LU" provides an alternative sparse matrix decomposition approach. In addition, there are "Chebyshev" and Monte Carlo "MC" approximate log-determinant methods; the Smirnov/Anselin (2009) trace approximation is available as "moments". Three methods: "SE\_classic", "SE\_whichMin", and "SE\_interp" are provided experimentally, the first to attempt to emulate the behaviour of Spatial Econometrics toolbox ML fitting functions. All use grids of log determinant values, and the latter two attempt to ameliorate some features of "SE classic".

quiet default NULL, use !verbose global option value; if FALSE, reports function

values during optimization.

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged

value of zones without neighbours, if FALSE (default) assign NA - causing

lagsarlm() to terminate with an error

interval default is NULL, search interval for autoregressive parameter

tol.solve the tolerance for detecting linear dependencies in the columns of matrices to be

> inverted - passed to solve() (default=1.0e-10). This may be used if necessary to extract coefficient standard errors (for instance lowering to 1e-12), but errors in solve() may constitute indications of poorly scaled variables: if the variables have scales differing much from the autoregressive coefficient, the values in this matrix may be very different in scale, and inverting such a matrix is analytically possible by definition, but numerically unstable; rescaling the RHS variables

alleviates this better than setting tol.solve to a very small value

default NULL, if given, a vector of powered spatial weights matrix traces output trs by trW; when given, insert the asymptotic analytical values into the numerical

Hessian instead of the approximated values; may be used to get around some problems raised when the numerical Hessian is poorly conditioned, generating

NaNs in subsequent operations; the use of trs is recommended

list of extra control arguments - see section below control

### **Details**

The asymptotic standard error of  $\rho$  is only computed when method=eigen, because the full matrix operations involved would be costly for large n typically associated with the choice of method="spam" or "Matrix". The same applies to the coefficient covariance matrix. Taken as the asymptotic matrix from the literature, it is typically badly scaled, and with the elements involving  $\rho$  being very small, while other parts of the matrix can be very large (often many orders of magnitude in difference). It often happens that the tol.solve argument needs to be set to a smaller value than the default, or the RHS variables can be centred or reduced in range.

Versions of the package from 0.4-38 include numerical Hessian values where asymptotic standard errors are not available. This change has been introduced to permit the simulation of distributions for impact measures. The warnings made above with regard to variable scaling also apply in this case.

Note that the fitted() function for the output object assumes that the response variable may be reconstructed as the sum of the trend, the signal, and the noise (residuals). Since the values of the response variable are known, their spatial lags are used to calculate signal components (Cressie

1993, p. 564). This differs from other software, including GeoDa, which does not use knowledge of the response variable in making predictions for the fitting data.

## Value

A list object of class sarlm

type "lag" or "mixed"

rho simultaneous autoregressive lag coefficient

coefficients GLS coefficient estimates

rest.se asymptotic standard errors if ase=TRUE, otherwise approximate numeriacal Hessian-

based values

LL log likelihood value at computed optimum

SSE GLS residual variance
SSE sum of squared GLS errors
parameters number of parameters estimated

logLik\_lm.model

Log likelihood of the linear model for  $\rho = 0$ 

AIC\_lm.model AIC of the linear model for  $\rho = 0$ 

method the method used to calculate the Jacobian

call the call used to create this object

residuals GLS residuals

tarX model matrix of the GLS model tary response of the GLS model

y response of the linear model for  $\rho=0$  X model matrix of the linear model for  $\rho=0$ 

opt object returned from numerical optimisation

se.fit Not used yet

ase TRUE if method=eigen

rho. se if ase=TRUE, the asymptotic standard error of  $\rho$ , otherwise approximate nume-

riacal Hessian-based value

LMtest if ase=TRUE, the Lagrange Multiplier test for the absence of spatial autocorre-

lation in the lag model residuals

resvar the asymptotic coefficient covariance matrix for (s2, rho, B)

zero.policy zero.policy for this model

aliased the aliased explanatory variables (if any) listw\_style the style of the spatial weights used interval the line search interval used to find  $\rho$ 

	fdHess	the numerical Hessian-based coefficient covariance matrix for (rho, B) if computed
	optimHess	if TRUE and fdHess returned, optim used to calculate Hessian at optimum
		if TRUE and fdHess returned, the asymptotic analytical values are inserted into the numerical Hessian instead of the approximated values, and its size increased to include the first row/column for sigma2
	LLNullLlm	Log-likelihood of the null linear model
	timings	processing timings
	f_calls	number of calls to the log likelihood function during optimization
	hf_calls	number of calls to the log likelihood function during numerical Hessian computation
	intern_classic	a data frame of detval matrix row choices used by the SE toolbox classic method
	na.action	(possibly) named vector of excluded or omitted observations if non-default na. action argument used $$

The internal sar.lag.mixed.\* functions return the value of the log likelihood function at  $\rho$ .

## Control arguments

**tol.opt:** the desired accuracy of the optimization - passed to optimize() (default=square root of double precision machine tolerance, a larger root may be used needed, see help(boston) for an example)

**fdHess:** default NULL, then set to (method != "eigen") internally; use fdHess to compute an approximate Hessian using finite differences when using sparse matrix methods; used to make a coefficient covariance matrix when the number of observations is large; may be turned off to save resources if need be

**optimHess:** default FALSE, use fdHess from **nlme**, if TRUE, use optim to calculate Hessian at optimum

optimHessMethod: default "optimHess", may be "nlm" or one of the optim methods

**compiled\_sse:** default FALSE; logical value used in the log likelihood function to choose compiled code for computing SSE

**Imult:** default 2; used for preparing the Cholesky decompositions for updating in the Jacobian function

super: if NULL (default), set to FALSE to use a simplicial decomposition for the sparse Cholesky decomposition and method "Matrix\_J", set to as.logical(NA) for method "Matrix", if TRUE, use a supernodal decomposition

cheb\_q: default 5; highest power of the approximating polynomial for the Chebyshev approximation

**MC\_p:** default 16; number of random variates

MC\_m: default 30; number of products of random variates matrix and spatial weights matrix

spamPivot: default "MMD", alternative "RCM"

in\_coef default 0.1, coefficient value for initial Cholesky decomposition in "spam\_update"

**type** default "MC", used with method "moments"; alternatives "mult" and "moments", for use if trs is missing, trW

correct default TRUE, used with method "moments" to compute the Smirnov/Anselin correction term

trunc default TRUE, used with method "moments" to truncate the Smirnov/Anselin correction term

**SE\_method** default "LU", may be "MC"

**nrho** default 200, as in SE toolbox; the size of the first stage lndet grid; it may be reduced to for example 40

interpn default 2000, as in SE toolbox; the size of the second stage Indet grid

small\_asy default TRUE; if the method is not "eigen", use asymmetric covariances rather than numerical Hessian ones if n <= small</pre>

**small** default 1500; threshold number of observations for asymmetric covariances when the method is not "eigen"

**SEIndet** default NULL, may be used to pass a pre-computed SE toolbox style matrix of coefficients and their Indet values to the "SE\_classic" and "SE\_whichMin" methods

**LU\_order** default FALSE; used in "LU\_prepermutate", note warnings given for 1u method **pre\_eig** default NULL; may be used to pass a pre-computed vector of eigenvalues

### Author(s)

Roger Bivand <Roger.Bivand@nhh.no>, with thanks to Andrew Bernat for contributions to the asymptotic standard error code.

## References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion; Ord, J. K. 1975 Estimation methods for models of spatial interaction, Journal of the American Statistical Association, 70, 120-126; Anselin, L. 1988 Spatial econometrics: methods and models. (Dordrecht: Kluwer); Anselin, L. 1995 SpaceStat, a software program for the analysis of spatial data, version 1.80. Regional Research Institute, West Virginia University, Morgantown, WV; Anselin L, Bera AK (1998) Spatial dependence in linear regression models with an introduction to spatial econometrics. In: Ullah A, Giles DEA (eds) Handbook of applied economic statistics. Marcel Dekker, New York, pp. 237-289; Cressie, N. A. C. 1993 Statistics for spatial data, Wiley, New York; LeSage J and RK Pace (2009) Introduction to Spatial Econometrics. CRC Press, Boca Raton.

#### See Also

lm, errorsarlm, eigenw, predict.sarlm, impacts.sarlm, residuals.sarlm, do\_ldet

```
data(oldcol)
COL.lag.eig <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    nb2listw(COL.nb, style="W"), method="eigen", quiet=FALSE)
summary(COL.lag.eig, correlation=TRUE)
COL.lag.eig$fdHess</pre>
```

listw2sn 99

```
COL.lag.eig$resvar
W <- as(as_dgRMatrix_listw(nb2listw(COL.nb)), "CsparseMatrix")</pre>
trMatc <- trW(W, type="mult")</pre>
COL.lag.eig1 <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
nb2listw(COL.nb, style="W"), control=list(fdHess=TRUE), trs=trMatc)
COL.lag.eig1$fdHess
system.time(COL.lag.M <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,</pre>
nb2listw(COL.nb), method="Matrix", quiet=FALSE))
summary(COL.lag.M)
impacts(COL.lag.M, listw=nb2listw(COL.nb))
## Not run:
system.time(COL.lag.sp <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,</pre>
nb2listw(COL.nb), method="spam", quiet=FALSE))
summary(COL.lag.sp)
## End(Not run)
COL.lag.B <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
nb2listw(COL.nb, style="B"))
summary(COL.lag.B, correlation=TRUE)
COL.mixed.B <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
nb2listw(COL.nb, style="B"), type="mixed", tol.solve=1e-9)
summary(COL.mixed.B, correlation=TRUE)
COL.mixed.W <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
nb2listw(COL.nb, style="W"), type="mixed")
summary(COL.mixed.W, correlation=TRUE)
NA.COL.OLD <- COL.OLD
NA.COL.OLD$CRIME[20:25] <- NA
COL.lag.NA <- lagsarlm(CRIME ~ INC + HOVAL, data=NA.COL.OLD,
nb2listw(COL.nb), na.action=na.exclude,
control=list(tol.opt=.Machine$double.eps^0.4))
COL.lag.NA$na.action
COL.lag.NA
resid(COL.lag.NA)
## Not run:
data(boston)
gp2mM <- lagsarlm(log(CMEDV) ~ CRIM + ZN + INDUS + CHAS + I(NOX^2) +</pre>
I(RM^2) + AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
data=boston.c, nb2listw(boston.soi), type="mixed", method="Matrix")
summary(gp2mM)
W <- as(as_dgRMatrix_listw(nb2listw(boston.soi)), "CsparseMatrix")</pre>
trMatb <- trW(W, type="mult")</pre>
gp2mMi <- lagsarlm(log(CMEDV) ~ CRIM + ZN + INDUS + CHAS + I(NOX^2) +</pre>
I(RM^2) + AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
data=boston.c, nb2listw(boston.soi), type="mixed", method="Matrix",
trs=trMatb)
summary(gp2mMi)
## End(Not run)
```

100 listw2sn

## **Description**

The function makes a "spatial neighbour" object representation (similar to the S-PLUS spatial statististics module representation of a "listw" spatial weights object. sn2listw() is the inverse function to listw2sn(), creating a "listw" object from a "spatial neighbour" object. The as.spam.listw method converts a "listw" object to a sparse matrix as defined in the **spam** package, using listw2sn().

### Usage

```
listw2sn(listw)
sn2listw(sn)
as.spam.listw(listw)
```

## **Arguments**

listw a listw object from for example nb2listw

sn a spatial.neighbour object

#### Value

listw2sn()returns a data frame with three columns, and with class spatial.neighbour:

from region number id for the start of the link (S-PLUS row.id) to region number id for the end of the link (S-PLUS col.id)

weights weight for this link

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

### See Also

```
nb2listw
```

```
example(columbus)
col.listw <- nb2listw(col.gal.nb)
col.listw$neighbours[[1]]
col.listw$weights[[1]]
col.sn <- listw2sn(col.listw)
str(col.sn)
## Not run:
col.sp <- as.spam.listw(col.listw)
str(col.sp)
## End(Not run)</pre>
```

Im.LMtests

lm.LMtests	Lagrange Multiplier diagnostics for spatial dependence in linear models

## **Description**

The function reports the estimates of tests chosen among five statistics for testing for spatial dependence in linear models. The statistics are the simple LM test for error dependence (LMerr), the simple LM test for a missing spatially lagged dependent variable (LMlag), variants of these robust to the presence of the other (RLMerr, RLMlag - RLMerr tests for error dependence in the possible presence of a missing lagged dependent variable, RLMlag the other way round), and a portmanteau test (SARMA, in fact LMerr + RLMlag). Note: from spdep 0.3-32, the value of the weights matrix trace term is returned correctly for both underlying symmetric and asymmetric neighbour lists, before 0.3-32, the value was wrong for listw objects based on asymmetric neighbour lists, such as k-nearest neighbours (thanks to Luc Anselin for finding the bug).

## Usage

```
lm.LMtests(model, listw, zero.policy=NULL, test="LMerr", spChk=NULL, naSubset=TRUE)
## S3 method for class 'LMtestlist'
print(x, ...)
## S3 method for class 'LMtestlist'
summary(object, p.adjust.method="none", ...)
## S3 method for class 'LMtestlist.summary'
print(x, digits=max(3, getOption("digits") - 2), ...)
```

#### **Arguments**

model	an object of class 1m returned by 1m, or optionally a vector of externally calculated residuals (run though na.omit if any NAs present) for use when only "LMerr" is chosen; weights and offsets should not be used in the 1m object
listw	a listw object created for example by ${\tt nb2listw}$ , expected to be row-standardised (W-style)
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
test	a character vector of tests requested chosen from LMerr, LMlag, RLMerr, RLMlag, SARMA; test="all" computes all the tests.
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
naSubset	default TRUE to subset listw object for omitted observations in model object (this is a change from earlier behaviour, when the model\$na.action component was ignored, and the listw object had to be subsetted by hand)
x, object	object to be printed

lm.LMtests

p.adjust.method

a character string specifying the probability value adjustment (see p.adjust)

for multiple tests, default "none"

digits minimum number of significant digits to be used for most numbers

... printing arguments to be passed through

## **Details**

The two types of dependence are for spatial lag  $\rho$  and spatial error  $\lambda$ :

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \rho \mathbf{W_{(1)}} \mathbf{y} + \mathbf{u},$$

$$\mathbf{u} = \lambda \mathbf{W}_{(2)} \mathbf{u} + \mathbf{e}$$

where e is a well-behaved, uncorrelated error term. Tests for a missing spatially lagged dependent variable test that  $\rho=0$ , tests for spatial autocorrelation of the error u test whether  $\lambda=0$ . W is a spatial weights matrix; for the tests used here they are identical.

#### Value

A list of class LMtestlist of htest objects, each with:

statistic the value of the Lagrange Multiplier test.

parameter number of degrees of freedom

p.value the p-value of the test.

method a character string giving the method used.

data. name a character string giving the name(s) of the data.

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no > and Andrew Bernat

## References

Anselin, L. 1988 Spatial econometrics: methods and models. (Dordrecht: Kluwer); Anselin, L., Bera, A. K., Florax, R. and Yoon, M. J. 1996 Simple diagnostic tests for spatial dependence. Regional Science and Urban Economics, 26, 77–104.

## See Also

1m

Im.morantest

## **Examples**

lm.morantest

Moran's I test for residual spatial autocorrelation

# Description

Moran's I test for spatial autocorrelation in residuals from an estimated linear model (lm()). The helper function listw2U() constructs a weights list object corresponding to the sparse matrix  $\frac{1}{2}(\mathbf{W} + \mathbf{W}')$ 

## Usage

```
lm.morantest(model, listw, zero.policy=NULL, alternative = "greater",
    spChk=NULL, resfun=weighted.residuals, naSubset=TRUE)
listw2U(listw)
```

## **Arguments**

model	an object of class $1m$ returned by $1m$ ; weights may be specified in the $1m$ fit, but offsets should not be used
listw	a listw object created for example by nb2listw
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA $$
alternative	a character string specifying the alternative hypothesis, must be one of "greater" (default), "less" or "two.sided".
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
resfun	default: weighted.residuals; the function to be used to extract residuals from the lm object, may be residuals, weighted.residuals, rstandard, or rstudent
naSubset	default TRUE to subset listw object for omitted observations in model object (this is a change from earlier behaviour, when the model\$na.action component was ignored, and the listw object had to be subsetted by hand)

104 Im.morantest

## Value

A list with class htest containing the following components:

statistic the value of the standard deviate of Moran's I.

p. value the p-value of the test.

estimate the value of the observed Moran's I, its expectation and variance under the

method assumption.

alternative a character string describing the alternative hypothesis.

method a character string giving the method used.

data.name a character string giving the name(s) of the data.

## Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

## References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 203,

#### See Also

```
lm.LMtests.lm
```

```
data(oldcol)
oldcrime1.lm <- lm(CRIME ~ 1, data = COL.OLD)
oldcrime.lm <- lm(CRIME ~ HOVAL + INC, data = COL.OLD)
lm.morantest(oldcrime.lm, nb2listw(COL.nb, style="W"))
lm.LMtests(oldcrime.lm, nb2listw(COL.nb, style="W"))
lm.morantest(oldcrime.lm, nb2listw(COL.nb, style="S"))
lm.morantest(oldcrime1.lm, nb2listw(COL.nb, style="W"))
moran.test(COL.OLD$CRIME, nb2listw(COL.nb, style="W"),
    randomisation=FALSE)
oldcrime.wlm <- lm(CRIME ~ HOVAL + INC, data = COL.OLD,
    weights = I(1/AREA))
lm.morantest(oldcrime.wlm, nb2listw(COL.nb, style="W"),
    resfun=weighted.residuals)
lm.morantest(oldcrime.wlm, nb2listw(COL.nb, style="W"),
    resfun=rstudent)</pre>
```

Im.morantest.exact 105

# Description

The function implements Tiefelsdorf's exact global Moran's I test.

## Usage

```
lm.morantest.exact(model, listw, zero.policy = NULL, alternative = "greater",
   spChk = NULL, resfun = weighted.residuals, zero.tol = 1e-07, Omega=NULL,
   save.M=NULL, save.U=NULL, useTP=FALSE, truncErr=1e-6, zeroTreat=0.1)
## S3 method for class 'moranex'
print(x, ...)
```

# Arguments

model	an object of class $1m$ returned by $1m$ ; weights may be specified in the $1m$ fit, but offsets should not be used
listw	a listw object created for example by nb2listw
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
alternative	a character string specifying the alternative hypothesis, must be one of greater (default), less or two.sided.
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
resfun	default: weighted.residuals; the function to be used to extract residuals from the lm object, may be residuals, weighted.residuals, rstandard, or rstudent
zero.tol	tolerance used to find eigenvalues close to absolute zero
Omega	A SAR process matrix may be passed in to test an alternative hypothesis, for example Omega <- invIrW(listw, rho=0.1); Omega <- tcrossprod(Omega), chol() is taken internally
save.M	$return \ the \ full \ M \ matrix \ for \ use \ in \ spdep:::exactMoranAlt$
save.U	return the full U matrix for use in spdep:::exactMoranAlt
useTP	default FALSE, if TRUE, use truncation point in integration rather than upper=Inf, see Tiefelsdorf (2000), eq. 6.7, p.69
truncErr	when useTP=TRUE, pass truncation error to truncation point function
zeroTreat	when useTP=TRUE, pass zero adjustment to truncation point function
x	a moranex object
	arguments to be passed through

106 Im.morantest.sad

#### Value

A list of class moranex with the following components:

statistic the value of the saddlepoint approximation of the standard deviate of global

Moran's I.

p. value the p-value of the test.

estimate the value of the observed global Moran's I.
method a character string giving the method used.

alternative a character string describing the alternative hypothesis.

gamma eigenvalues (excluding zero values)

oType usually set to "E"

data. name a character string giving the name(s) of the data.

df degrees of freedom

## Author(s)

Markus Reder and Roger Bivand

## See Also

```
lm.morantest.sad
```

## **Examples**

```
require(maptools)
eire <- readShapePoly(system.file("etc/shapes/eire.shp", package="spdep")[1],
    ID="names", proj4string=CRS("+proj=utm +zone=30 +units=km"))
eire.nb <- poly2nb(eire)
#data(eire)
e.lm <- lm(OWNCONS ~ ROADACC, data=eire)
lm.morantest(e.lm, nb2listw(eire.nb))
lm.morantest.sad(e.lm, nb2listw(eire.nb))
lm.morantest.exact(e.lm, nb2listw(eire.nb))
lm.morantest.exact(e.lm, nb2listw(eire.nb)), useTP=TRUE)</pre>
```

lm.morantest.sad

Saddlepoint approximation of global Moran's I test

## **Description**

The function implements Tiefelsdorf's application of the Saddlepoint approximation to global Moran's I's reference distribution.

Im.morantest.sad 107

## Usage

```
lm.morantest.sad(model, listw, zero.policy=NULL, alternative="greater",
    spChk=NULL, resfun=weighted.residuals, tol=.Machine$double.eps^0.5,
    maxiter=1000, tol.bounds=0.0001, zero.tol = 1e-07, Omega=NULL,
    save.M=NULL, save.U=NULL)
## S3 method for class 'moransad'
print(x, ...)
## S3 method for class 'moransad'
summary(object, ...)
## S3 method for class 'summary.moransad'
print(x, ...)
```

## **Arguments**

model	an object of class $lm$ returned by $lm$ ; weights may be specified in the $lm$ fit, but offsets should not be used
listw	a listw object created for example by nb2listw
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
alternative	a character string specifying the alternative hypothesis, must be one of greater (default), less or two.sided.
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
resfun	default: weighted.residuals; the function to be used to extract residuals from the lm object, may be residuals, weighted.residuals, rstandard, or rstudent
tol	the desired accuracy (convergence tolerance) for uniroot
maxiter	the maximum number of iterations for uniroot
tol.bounds	offset from bounds for uniroot
zero.tol	tolerance used to find eigenvalues close to absolute zero
_	6
Omega	A SAR process matrix may be passed in to test an alternative hypothesis, for example Omega <- invIrW(listw, rho=0.1); Omega <- tcrossprod(Omega), chol() is taken internally
Omega save.M	A SAR process matrix may be passed in to test an alternative hypothesis, for example Omega <- invIrW(listw, rho=0.1); Omega <- tcrossprod(Omega),
g	A SAR process matrix may be passed in to test an alternative hypothesis, for example Omega <- invIrW(listw, rho=0.1); Omega <- tcrossprod(Omega), chol() is taken internally
save.M	A SAR process matrix may be passed in to test an alternative hypothesis, for example Omega <- invIrW(listw, rho=0.1); Omega <- tcrossprod(Omega), chol() is taken internally return the full M matrix for use in spdep:::exactMoranAlt
save.M save.U	A SAR process matrix may be passed in to test an alternative hypothesis, for example Omega <- invIrW(listw, rho=0.1); Omega <- tcrossprod(Omega), chol() is taken internally return the full M matrix for use in spdep:::exactMoranAlt return the full U matrix for use in spdep:::exactMoranAlt

## **Details**

The function involves finding the eigenvalues of an n by n matrix, and numerically finding the root for the Saddlepoint approximation, and should therefore only be used with care when n is large.

108 lm.morantest.sad

#### Value

A list of class moransad with the following components:

statistic the value of the saddlepoint approximation of the standard deviate of global

Moran's I.

p.value the p-value of the test.

estimate the value of the observed global Moran's I.

alternative a character string describing the alternative hypothesis.

method a character string giving the method used.

data. name a character string giving the name(s) of the data.

internal1 Saddlepoint omega, r and u

internal2 f.root, iter and estim.prec from uniroot

df degrees of freedom

tau eigenvalues (excluding zero values)

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

### References

Tiefelsdorf, M. 2002 The Saddlepoint approximation of Moran's I and local Moran's Ii reference distributions and their numerical evaluation. Geographical Analysis, 34, pp. 187–206.

## See Also

lm.morantest

```
require(maptools)
eire <- readShapePoly(system.file("etc/shapes/eire.shp", package="spdep")[1],
    ID="names", proj4string=CRS("+proj=utm +zone=30 +units=km"))
eire.nb <- poly2nb(eire)
#data(eire)
e.lm <- lm(OWNCONS ~ ROADACC, data=eire)
lm.morantest(e.lm, nb2listw(eire.nb))
lm.morantest.sad(e.lm, nb2listw(eire.nb))
summary(lm.morantest.sad(e.lm, nb2listw(eire.nb)))
e.wlm <- lm(OWNCONS ~ ROADACC, data=eire, weights=RETSALE)
lm.morantest(e.wlm, nb2listw(eire.nb), resfun=rstudent)
lm.morantest.sad(e.wlm, nb2listw(eire.nb), resfun=rstudent)</pre>
```

localG 109

localG	G and Gstar local spatial statistics	

## **Description**

The local spatial statistic G is calculated for each zone based on the spatial weights object used. The value returned is a Z-value, and may be used as a diagnostic tool. High positive values indicate the posibility of a local cluster of high values of the variable being analysed, very low relative values a similar cluster of low values. For inference, a Bonferroni-type test is suggested in the references, where tables of critical values may be found (see also details below).

## Usage

```
localG(x, listw, zero.policy=NULL, spChk=NULL)
```

## **Arguments**

X	a numeric vector the same length as the neighbours list in listw
listw	a listw object created for example by nb2listw
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
spChk	should the data vector names be checked against the spatial objects for identity

integrity, TRUE, or FALSE, default NULL to use get.spChkOption()

### **Details**

If the neighbours member of listw has a "self.included" attribute set to TRUE, the Gstar variant, including the self-weight  $w_{ii}>0$ , is calculated and returned. The returned vector will have a "gstari" attribute set to TRUE. Self-weights can be included by using the include.self function in the spweights package before converting the neighbour list to a spatial weights list with nb2listw as shown below in the example.

The critical values of the statistic under assumptions given in the references for the 95th percentile are for n=1: 1.645, n=50: 3.083, n=100: 3.289, n=1000: 3.886.

#### Value

A vector of G or Gstar values, with attributes "gstari" set to TRUE or FALSE, "call" set to the function call, and class "localG".

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

110 localmoran

#### References

Ord, J. K. and Getis, A. 1995 Local spatial autocorrelation statistics: distributional issues and an application. *Geographical Analysis*, 27, 286–306; Getis, A. and Ord, J. K. 1996 Local spatial statistics: an overview. In P. Longley and M. Batty (eds) *Spatial analysis: modelling in a GIS environment* (Cambridge: Geoinformation International), 261–277.

```
data(getisord)
xycoords <- cbind(xyz$x, xyz$y)</pre>
nb30 <- dnearneigh(xycoords, 0, 30)</pre>
G30 <- localG(xyz$val, nb2listw(nb30, style="B"))
G30[length(xyz$val)-136]
nb60 <- dnearneigh(xycoords, 0, 60)</pre>
G60 <- localG(xyz$val, nb2listw(nb60, style="B"))
G60[length(xyz$val)-136]
nb90 <- dnearneigh(xycoords, 0, 90)</pre>
G90 <- localG(xyz$val, nb2listw(nb90, style="B"))
G90[length(xyz$val)-136]
nb120 <- dnearneigh(xycoords, 0, 120)</pre>
G120 <- localG(xyz$val, nb2listw(nb120, style="B"))</pre>
G120[length(xyz$val)-136]
nb150 <- dnearneigh(xycoords, 0, 150)</pre>
G150 <- localG(xyz$val, nb2listw(nb150, style="B"))
G150[length(xyz$val)-136]
brks <- seq(-5,5,1)
cm.col <- cm.colors(length(brks)-1)</pre>
image(x, y, t(matrix(G30, nrow=16, ncol=16, byrow=TRUE)),
  breaks=brks, col=cm.col, asp=1)
text(xyz$x, xyz$y, round(G30, digits=1), cex=0.7)
polygon(c(195,225,225,195), c(195,195,225,225), lwd=2)
title(main=expression(paste("Values of the ", G[i], " statistic")))
G30s <- localG(xyz$val, nb2listw(include.self(nb30),
 style="B"))
cat("value according to Getis and Ord's eq. 14.2, p. 263 (1996)\n")
G30s[length(xyz$val)-136]
cat(paste("value given by Getis and Ord (1996), p. 267",
  "(division by n-1 rather than n \n in variance)\n"))
G30s[length(xyz$val)-136] *
  (sqrt(sum(scale(xyz$val, scale=FALSE)^2)/length(xyz$val)) /
  sqrt(var(xyz$val)))
image(x, y, t(matrix(G30s, nrow=16, ncol=16, byrow=TRUE)),
  breaks=brks, col=cm.col, asp=1)
text(xyz$x, xyz$y, round(G30s, digits=1), cex=0.7)
polygon(c(195,225,225,195), c(195,195,225,225), lwd=2)
title(main=expression(paste("Values of the ", G[i]^"*", " statistic")))
```

localmoran 111

### **Description**

The local spatial statistic Moran's I is calculated for each zone based on the spatial weights object used. The values returned include a Z-value, and may be used as a diagnostic tool. The statistic is:

$$I_i = \frac{(x_i - \bar{x})}{\sum_{k=1}^n (x_k - \bar{x})^2 / (n-1)} \sum_{j=1}^n w_{ij} (x_j - \bar{x})$$

, and its expectation and variance are given in Anselin (1995).

## Usage

## **Arguments**

x a numeric vector the same length as the neighbours list in listw

listw a listw object created for example by nb2listw

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged value

of zones without neighbours, if FALSE assign NA

na.action a function (default na.fail), can also be na.omit or na.exclude - in these

cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create noneighbour observations. Note that only weights lists created without using the glist argument to nb2listw may be subsetted. If na. pass is used, zero is substituted for NA values in calculating the spatial lag. (Note that na.exclude will only work properly starting from R 1.9.0, na.omit and na.exclude assign the wrong

classes in 1.8.\*)

alternative a character string specifying the alternative hypothesis, must be one of greater

(default), less or two.sided.

p.adjust.method

a character string specifying the probability value adjustment for multiple tests, default "none"; see p.adjustSP. Note that the number of multiple tests for each region is only taken as the number of neighbours + 1 for each region, rather than

the total number of regions.

mlvar default TRUE: values of local Moran's I are reported using the variance of the

variable of interest (sum of squared deviances over n), but can be reported as the sample variance, dividing by (n-1) instead; both are used in other implementa-

tions

spChk should the data vector names be checked against the spatial objects for identity

integrity, TRUE, or FALSE, default NULL to use get.spChkOption()

## Value

Ii local moran statistic

112 localmoran

E.Ii	expectation of local moran statistic
Var.Ii	variance of local moran statistic
Z.Ii	standard deviate of local moran statistic
Pr()	p-value of local moran statistic

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Anselin, L. 1995. Local indicators of spatial association, Geographical Analysis, 27, 93–115; Getis, A. and Ord, J. K. 1996 Local spatial statistics: an overview. In P. Longley and M. Batty (eds) *Spatial analysis: modelling in a GIS environment* (Cambridge: Geoinformation International), 261–277.

### See Also

localG

```
data(afcon)
oid <- order(afcon$id)</pre>
resI <- localmoran(afcon$totcon, nb2listw(paper.nb))</pre>
printCoefmat(data.frame(resI[oid,], row.names=afcon$name[oid]),
check.names=FALSE)
hist(resI[,5])
resI <- localmoran(afcon$totcon, nb2listw(paper.nb),</pre>
 p.adjust.method="bonferroni")
printCoefmat(data.frame(resI[oid,], row.names=afcon$name[oid]),
 check.names=FALSE)
hist(resI[,5])
totcon <-afcon$totcon
is.na(totcon) <- sample(1:length(totcon), 5)</pre>
totcon
resI.na <- localmoran(totcon, nb2listw(paper.nb), na.action=na.exclude,</pre>
zero.policy=TRUE)
if (class(attr(resI.na, "na.action")) == "exclude") {
print(data.frame(resI.na[oid,], row.names=afcon$name[oid]), digits=2)
} else print(resI.na, digits=2)
resG <- localG(afcon$totcon, nb2listw(include.self(paper.nb)))</pre>
print(data.frame(resG[oid], row.names=afcon$name[oid]), digits=2)
```

localmoran.exact 113

localmoran.exact Exact local Moran's Ii tests	localmoran.exact	calmoran.exact Exact local Moran's Ii tests

## Description

localmoran.exact provides exact local Moran's Ii tests under the null hypothesis, while localmoran.exact.alt provides exact local Moran's Ii tests under the alternative hypothesis. In this case, the model may be a fitted model returned by errorsarlm from which the covariance matrix is retrieved, or the covariance matrix can be passed through the Omega= argument.

## Usage

```
localmoran.exact(model, select, nb, glist = NULL, style = "W",
  zero.policy = NULL, alternative = "greater", spChk = NULL,
  resfun = weighted.residuals, save.Vi = FALSE, useTP=FALSE, truncErr=1e-6,
  zeroTreat=0.1)
localmoran.exact.alt(model, select, nb, glist = NULL, style = "W",
  zero.policy = NULL, alternative = "greater", spChk = NULL,
  resfun = weighted.residuals, Omega = NULL, save.Vi = FALSE,
  save.M = FALSE, useTP=FALSE, truncErr=1e-6, zeroTreat=0.1)
## S3 method for class 'localmoranex'
  print(x, ...)
## S3 method for class 'localmoranex'
as.data.frame(x, row.names=NULL, optional=FALSE, ...)
```

## Arguments

model	an object of class 1m returned by 1m (assuming no global spatial autocorrelation), or an object of class sar1m returned by a spatial simultaneous autoregressive model fit (assuming global spatial autocorrelation represented by the model spatial coefficient); weights may be specified in the 1m fit, but offsets should not be used
select	an integer vector of the id. numbers of zones to be tested; if missing, all zones
nb	a list of neighbours of class nb
glist	a list of general weights corresponding to neighbours
style	can take values W, B, C, and S
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
alternative	a character string specifying the alternative hypothesis, must be one of greater (default), less or two.sided.
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
resfun	default: weighted.residuals; the function to be used to extract residuals from the lm object, may be residuals, weighted.residuals, rstandard, or rstudent

114 localmoran.exact

Omega A SAR process matrix may be passed in to test an alternative hypothesis, for ex-

ample Omega <- invIrW(listw, rho=0.1); Omega <- tcrossprod(Omega),</pre>

chol() is taken internally

save. Vi if TRUE, return the star-shaped weights lists for each zone tested

save.M if TRUE, save a list of left and right M products

useTP default FALSE, if TRUE, use truncation point in integration rather than up-

per=Inf, see Tiefelsdorf (2000), eq. 6.7, p.69

truncErr when useTP=TRUE, pass truncation error to truncation point function zeroTreat when useTP=TRUE, pass zero adjustment to truncation point function

x object to be printed

row.names ignored argument to as.data.frame.localmoranex; row names assigned from

localmoranex object

optional ignored argument to as.data.frame.localmoranex; row names assigned from

localmoranex object

... arguments to be passed through

#### Value

A list with class localmoranex containing "select" lists, each with class moranex with the following components:

statistic the value of the saddlepoint approximation of the standard deviate of global

Moran's I.

p. value the p-value of the test.

estimate the value of the observed global Moran's I.
method a character string giving the method used.

alternative a character string describing the alternative hypothesis.

gamma eigenvalues (two extreme values for null, vector for alternative)

oType usually set to "E", but set to "N" if the integration leads to an out of domain

value for qnorm, when the Normal assumption is substituted. This only occurs

when the output p-value would be very close to zero

data.name a character string giving the name(s) of the data.

df degrees of freedom

i zone tested

Vi zone tested

When the alternative is being tested, a list of left and right M products in attribute M.

### Author(s)

Markus Reder and Roger Bivand

#### See Also

lm.morantest.exact, localmoran.sad

localmoran.sad 115

### **Examples**

```
eire <- readShapePoly(system.file("etc/shapes/eire.shp", package="spdep")[1],</pre>
 ID="names", proj4string=CRS("+proj=utm +zone=30 +units=km"))
eire.nb <- poly2nb(eire)</pre>
#data(eire)
e.lm <- lm(OWNCONS ~ ROADACC, data=eire)</pre>
localmoran.sad(e.lm, nb=eire.nb)
localmoran.exact(e.lm, nb=eire.nb)
localmoran.exact(e.lm, nb=eire.nb, useTP=TRUE)
e.errorsar <- errorsarlm(OWNCONS ~ ROADACC, data=eire,</pre>
listw=nb2listw(eire.nb))
lm.target <- lm(e.errorsar$tary ~ e.errorsar$tarX - 1)</pre>
localmoran.exact.alt(lm.target, nb=eire.nb)
Omega <- invIrW(nb2listw(eire.nb), rho=0.6)</pre>
Omega1 <- tcrossprod(Omega)</pre>
localmoran.exact.alt(lm.target, nb=eire.nb, Omega=Omega1)
localmoran.exact.alt(lm.target, nb=eire.nb, Omega=Omega1, useTP=TRUE)
```

localmoran.sad

Saddlepoint approximation of local Moran's Ii tests

## **Description**

The function implements Tiefelsdorf's application of the Saddlepoint approximation to local Moran's Ii's reference distribution. If the model object is of class "lm", global independence is assumed; if of class "sarlm", global dependence is assumed to be represented by the spatial parameter of that model. Tests are reported separately for each zone selected, and may be summarised using summary.localmoransad. Values of local Moran's Ii agree with those from localmoran(), but in that function, the standard deviate - here the Saddlepoint approximation - is based on the randomisation assumption.

## Usage

```
localmoran.sad(model, select, nb, glist=NULL, style="W",
  zero.policy=NULL, alternative="greater", spChk=NULL,
  resfun=weighted.residuals, save.Vi=FALSE,
  tol = .Machine$double.eps^0.5, maxiter = 1000, tol.bounds=0.0001,
  save.M=FALSE, Omega = NULL)

## S3 method for class 'localmoransad'
print(x, ...)
## S3 method for class 'localmoransad'
summary(object, ...)
## S3 method for class 'summary.localmoransad'
print(x, ...)
listw2star(listw, ireg, style, n, D, a, zero.policy=NULL)
```

localmoran.sad

# Arguments

`		
	model	an object of class 1m returned by 1m (assuming no global spatial autocorrelation), or an object of class sar1m returned by a spatial simultaneous autoregressive model fit (assuming global spatial autocorrelation represented by the model spatial coefficient); weights may be specified in the 1m fit, but offsets should not be used
	select	an integer vector of the id. numbers of zones to be tested; if missing, all zones
	nb	a list of neighbours of class nb
	glist	a list of general weights corresponding to neighbours
	style	can take values W, B, C, and S
	zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
	alternative	a character string specifying the alternative hypothesis, must be one of greater (default), less or two.sided.
	spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
	resfun	default: weighted.residuals; the function to be used to extract residuals from the lm object, may be residuals, weighted.residuals, rstandard, or rstudent
	save.Vi	if TRUE, return the star-shaped weights lists for each zone tested
	tol	the desired accuracy (convergence tolerance) for uniroot
	maxiter	the maximum number of iterations for uniroot
	tol.bounds	offset from bounds for uniroot
	save.M	if TRUE, save a list of left and right M products in a list for the conditional tests, or a list of the regression model matrix components
	Omega	A SAR process matrix may be passed in to test an alternative hypothesis, for example Omega <- invIrW(listw, rho=0.1); Omega <- tcrossprod(Omega), chol() is taken internally
	X	object to be printed
	object	object to be summarised
		arguments to be passed through
	listw	a listw object created for example by nb2listw
	ireg	a zone number
	n	internal value depending on listw and style
	D	internal value depending on listw and style
	а	internal value depending on listw and style

## **Details**

The function implements the analytical eigenvalue calculation together with trace shortcuts given or suggested in Tiefelsdorf (2002), partly following remarks by J. Keith Ord, and uses the Saddlepoint analytical solution from Tiefelsdorf's SPSS code.

localmoran.sad 117

If a histogram of the probability values of the saddlepoint estimate for the assumption of global independence is not approximately flat, the assumption is probably unjustified, and re-estimation with global dependence is recommended.

No n by n matrices are needed at any point for the test assuming no global dependence, the star-shaped weights matrices being handled as listw lists. When the test is made on residuals from a spatial regression, taking a global process into account. n by n matrices are necessary, and memory constraints may be reached for large lattices.

### Value

A list with class localmoransad containing "select" lists, each with class moransad with the following components:

statistic the value of the saddlepoint approximation of the standard deviate of local Moran's

Ii.

p. value the p-value of the test.

estimate the value of the observed local Moran's Ii.

alternative a character string describing the alternative hypothesis.

method a character string giving the method used.

data.name a character string giving the name(s) of the data.

internal1 Saddlepoint omega, r and u

df degrees of freedom

tau maximum and minimum analytical eigenvalues

i zone tested

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

### References

Tiefelsdorf, M. 2002 The Saddlepoint approximation of Moran's I and local Moran's Ii reference distributions and their numerical evaluation. Geographical Analysis, 34, pp. 187–206.

## See Also

localmoran, lm.morantest, lm.morantest.sad, errorsarlm

```
require(maptools)
eire <- readShapePoly(system.file("etc/shapes/eire.shp", package="spdep")[1],
    ID="names", proj4string=CRS("+proj=utm +zone=30 +units=km"))
eire.nb <- poly2nb(eire)
#data(eire)
e.lm <- lm(OWNCONS ~ ROADACC, data=eire)
e.locmor <- summary(localmoran.sad(e.lm, nb=eire.nb))</pre>
```

118 LR.sarlm

```
e.locmor
mean(e.locmor[,1])
lm.morantest(e.lm, nb2listw(eire.nb))
hist(e.locmor[,"Pr. (Sad)"])
e.wlm <- lm(OWNCONS ~ ROADACC, data=eire, weights=RETSALE)
e.locmorw1 <- summary(localmoran.sad(e.wlm, nb=eire.nb, resfun=weighted.residuals))
e.locmorw2 <- summary(localmoran.sad(e.wlm, nb=eire.nb, resfun=rstudent))
e.locmorw2
e.errorsar <- errorsarlm(OWNCONS ~ ROADACC, data=eire,
    listw=nb2listw(eire.nb))
e.errorsar
lm.target <- lm(e.errorsar$tary ~ e.errorsar$tarX - 1)
e.clocmor <- summary(localmoran.sad(lm.target, nb=eire.nb))
e.clocmor
hist(e.clocmor[,"Pr. (Sad)"])</pre>
```

LR.sarlm

Likelihood ratio test

## **Description**

The LR.sarlm() function provides a likelihood ratio test for objects for which a logLik() function exists for their class, or for objects of class logLik. LR1.sarlm() and Wald1.sarlm() are used internally in summary.sarlm(), but may be accessed directly; they report the values respectively of LR and Wald tests for the absence of spatial dependence in spatial lag or error models. The spatial Hausman test is available for models fitted with errorsarlm and GMerrorsar.

## Usage

```
LR.sarlm(x, y)
## S3 method for class 'sarlm'
logLik(object, ...)
LR1.sarlm(object)
Wald1.sarlm(object)
## S3 method for class 'sarlm'
Hausman.test(object, ..., tol=NULL)
## S3 method for class 'gmsar'
Hausman.test(object, ..., tol=NULL)
```

### **Arguments**

X	a logLik object or an object for which a logLik() function exists
у	a logLik object or an object for which a logLik() function exists
object	a sarlm object from lagsarlm() or errorsarlm()
	further arguments passed to or from other methods
tol	tol argument passed to solve, default NULL

LR.sarlm 119

## Value

The tests return objects of class htest with:

statistic value of statistic

parameter degrees of freedom

p.value Probability value

estimate varies with test

method description of test method

logLik.sarlm() returns an object of class logLik LR1.sarlm, Hausman.sarlm and Wald1.sarlm return objects of class htest

#### Note

The numbers of degrees of freedom returned by logLik.sarlm() include nuisance parameters, that is the number of regression coefficients, plus sigma, plus spatial parameter esitmate(s).

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## References

LeSage J and RK Pace (2009) Introduction to Spatial Econometrics. CRC Press, Boca Raton, pp. 61–63; Pace RK and LeSage J (2008) A spatial Hausman test. *Economics Letters* 101, 282–284.

#### See Also

```
logLik.lm, anova.sarlm
```

```
example(columbus)
mixed <- lagsarlm(CRIME ~ HOVAL + INC, data=columbus, nb2listw(col.gal.nb),
   type="mixed")
error <- errorsarlm(CRIME ~ HOVAL + INC, data=columbus, nb2listw(col.gal.nb))
LR.sarlm(mixed, error)
Hausman.test(error)</pre>
```

120 mat2listw

mat2listw	Convert a square spatial weights matrix to a weights list object

### **Description**

The function converts a square spatial weights matrix, optionally a sparse matrix to a weights list object, optionally adding region IDs from the row names of the matrix, as a sequence of numbers 1:nrow(x), or as given as an argument. The style can be imposed by rebuilting the weights list object internally.

## Usage

```
mat2listw(x, row.names = NULL, style="M")
```

## **Arguments**

x A square non-negative matrix with no NAs representing spatial weights; may be

a matrix of class "sparseMatrix"

row.names row names to use for region IDs

style default "M", unknown style; if not "M", passed to nb2listw to re-build the

object

### Value

A listw object with the following members:

style "M", meaning matrix style, underlying style unknown, or assigned style argu-

ment in rebuilt object

neighbours the derived neighbours list

weights the weights for the neighbours derived from the matrix

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

```
nb2listw, nb2mat
```

```
example(columbus)
coords <- coordinates(columbus)
col005 <- dnearneigh(coords, 0, 0.5, attr(col.gal.nb, "region.id"))
summary(col005)
col005.w.mat <- nb2mat(col005, zero.policy=TRUE)
col005.w.b <- mat2listw(col005.w.mat)</pre>
```

MCMCsamp 121

```
summary(col005.w.b$neighbours)
diffnb(col005, col005.w.b$neighbours)
col005.w.mat.3T <- kronecker(diag(3), col005.w.mat)
col005.w.b.3T <- mat2listw(col005.w.mat.3T, style="W")
summary(col005.w.b.3T$neighbours)
W <- as(as_dgRMatrix_listw(nb2listw(col005, style="W", zero.policy=TRUE)), "CsparseMatrix")
col005.spM <- mat2listw(W)
summary(col005.spM$neighbours)
diffnb(col005, col005.spM$neighbours)
IW <- kronecker(Diagonal(3), W)
col005.spM.3T <- mat2listw(IW, style="W")
summary(col005.spM.3T$neighbours)</pre>
```

**MCMCsamp** 

MCMC sample from fitted spatial regression

### **Description**

The MCMCsamp method uses rwmetrop, a random walk Metropolis algorithm, from **LearnBayes** to make MCMC samples from fitted maximum likelihood spatial regression models.

### Usage

```
MCMCsamp(object, mcmc = 1L, verbose = NULL, ...)
## S3 method for class 'spautolm'
MCMCsamp(object, mcmc = 1L, verbose = NULL, ...,
burnin = 0L, scale=1, listw, control = list())
## S3 method for class 'sarlm'
MCMCsamp(object, mcmc = 1L, verbose = NULL, ...,
burnin=0L, scale=1, listw, listw2=NULL, control=list())
```

### **Arguments**

object A spatial regression model object fitted by maximum likelihood with spautolm The number of MCMC iterations after burnin mcmc default NULL, use global option value; if TRUE, reports progress verbose Arguments passed through The number of burn-in iterations for the sampler burnin scale a positive scale parameter listw objects created for example by nb2listw; should be the same object(s) listw, listw2 used for fitting the model control list of extra control arguments - see spautolm

#### Value

An object of class "mcmc" suited to **coda**, with attributes: "accept" acceptance rate; "type" input ML fitted model type "SAR", "CAR", "SMA", "lag", "mixed", "error", "sac", "sacmixed"; "timings" run times

122 MCMCsamp

#### Note

If the acceptance rate is below 0.05, a warning will be issued; consider increasing mcmc.

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Jim Albert (2007) Bayesian Computation with R, Springer, New York, pp. 104-105.

#### See Also

rwmetrop, spautolm, lagsarlm, errorsarlm, sacsarlm

```
example(NY_data)
## Not run:
esar1f <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, family="SAR", method="eigen")
summary(esar1f)
res <- MCMCsamp(esar1f, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
ecar1f <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, family="CAR", method="eigen")
summary(ecar1f)
res <- MCMCsamp(ecar1f, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
esar1fw <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, weights=POP8, family="SAR", method="eigen")
summary(esar1fw)
res <- MCMCsamp(esar1fw, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
ecar1fw <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, weights=POP8, family="CAR", method="eigen")
summary(ecar1fw)
res <- MCMCsamp(ecar1fw, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
## End(Not run)
esar0 <- errorsarlm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY)
summary(esar0)
res <- MCMCsamp(esar0, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
## Not run:
esar1 <- errorsarlm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, etype="emixed")
summary(esar1)
res <- MCMCsamp(esar1, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
```

ME

```
lsar0 <- lagsarlm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY)
summary(lsar0)
res <- MCMCsamp(lsar0, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
lsar1 <- lagsarlm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, type="mixed")
summary(lsar1)
res <- MCMCsamp(lsar1, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
ssar0 <- sacsarlm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,</pre>
listw=listw_NY)
summary(ssar0)
res <- MCMCsamp(ssar0, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
ssar1 <- sacsarlm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, type="sacmixed")
summary(ssar1)
res <- MCMCsamp(ssar1, mcmc=5000, burnin=500, listw=listw_NY)
summary(res)
## End(Not run)
```

ME

Moran eigenvector GLM filtering

# Description

The Moran eigenvector filtering function is intended to remove spatial autocorrelation from the residuals of generalised linear models. It uses brute force eigenvector selection to reach a subset of such vectors to be added to the RHS of the GLM model to reduce residual autocorrelation to below the specified alpha value. Since eigenvector selection only works on symmetric weights, the weights are made symmetric before the eigenvectors are found (from spdep 0.5-50).

### Usage

```
ME(formula, data, family = gaussian, weights, offset, listw,
  alpha=0.05, nsim=99, verbose=NULL, stdev=FALSE)
```

## Arguments

formula	a symbolic description of the model to be fit
data	an optional data frame containing the variables in the model
family	a description of the error distribution and link function to be used in the model
weights	an optional vector of weights to be used in the fitting process
offset	this can be used to specify an a priori known component to be included in the linear predictor during fitting

124 ME

listw a listw object created for example by nb2listw

alpha used as a stopping rule to choose all eigenvectors up to and including the one

with a p-value exceeding alpha

nsim number of permutations for permutation bootstrap for finding p-values

verbose default NULL, use global option value; if TRUE report eigenvectors selected

stdev if TRUE, p-value calculated from bootstrap permutation standard deviate using

pnorm with alternative="greater", if FALSE the Hope-type p-value

#### **Details**

The eigenvectors for inclusion are chosen by calculating the empirical Moran's I values for the initial model plus each of the doubly centred symmetric spatial weights matrix eigenvectors in turn. Then the first eigenvector is chosen as that with the lowest Moran's I value. The procedure is repeated until the lowest remaining Moran's I value has a permutation-based probability value above alpha. The probability value is either Hope-type or based on using the mean and standard deviation of the permutations to calculate ZI based on the stdev argument.

#### Value

An object of class ME\_res:

selection a matrix summarising the selection of eigenvectors for inclusion, with columns:

Eigenvector number of selected eigenvector

**ZI** permutation-based standardized deviate of Moran's I if stdev=TRUE

pr(ZI) probability value: if stdev=TRUE of the permutation-based standardized deviate, if FALSE the Hope-type probability value, in both cases on-

sided

The first row is the value at the start of the search

vectors a matrix of the selected eigenvectors in order of selection

### Author(s)

Roger Bivand and Pedro Peres-Neto

#### References

Dray S, Legendre P and Peres-Neto PR (2005) Spatial modeling: a comprehensive framework for principle coordinate analysis of neigbbor matrices (PCNM), Ecological Modelling; Griffith DA and Peres-Neto PR (2006) Spatial modeling in ecology: the flexibility of eigenfunction spatial analyses.

## See Also

SpatialFiltering, glm

moran 125

```
## Not run:
example(columbus)
lmbase <- lm(CRIME ~ INC + HOVAL, data=columbus)</pre>
lagcol <- SpatialFiltering(CRIME ~ 1, ~ INC + HOVAL, data=columbus,</pre>
nb=col.gal.nb, style="W", alpha=0.1, verbose=TRUE)
lagcol
lmlag <- lm(CRIME ~ INC + HOVAL + fitted(lagcol), data=columbus)</pre>
anova(lmlag)
anova(lmbase, lmlag)
set.seed(123)
lagcol1 <- ME(CRIME ~ INC + HOVAL, data=columbus, family="gaussian",</pre>
listw=nb2listw(col.gal.nb), alpha=0.1, verbose=TRUE)
lagcol1
lmlag1 <- lm(CRIME ~ INC + HOVAL + fitted(lagcol1), data=columbus)</pre>
anova(lmlag1)
anova(lmbase, lmlag1)
set.seed(123)
lagcol2 <- ME(CRIME ~ INC + HOVAL, data=columbus, family="gaussian",</pre>
listw=nb2listw(col.gal.nb), alpha=0.1, stdev=TRUE, verbose=TRUE)
lmlag2 <- lm(CRIME ~ INC + HOVAL + fitted(lagcol2), data=columbus)</pre>
anova(lmlag2)
anova(lmbase, lmlag2)
example(nc.sids)
glmbase <- glm(SID74 ~ 1, data=nc.sids, offset=log(BIR74),</pre>
 family="poisson")
set.seed(123)
MEpois1 <- ME(SID74 ~ 1, data=nc.sids, offset=log(BIR74),
 family="poisson", listw=nb2listw(ncCR85_nb, style="B"), alpha=0.2, verbose=TRUE)
MEpois1
glmME <- glm(SID74 ~ 1 + fitted(MEpois1), data=nc.sids, offset=log(BIR74),</pre>
 family="poisson")
anova(glmME, test="Chisq")
anova(glmbase, glmME, test="Chisq")
data(hopkins)
hopkins_part <- hopkins[21:36,36:21]</pre>
hopkins_part[which(hopkins_part > 0, arr.ind=TRUE)] <- 1</pre>
hopkins.rook.nb <- cell2nb(16, 16, type="rook")</pre>
glmbase <- glm(c(hopkins_part) ~ 1, family="binomial")</pre>
set.seed(123)
MEbinom1 <- ME(c(hopkins_part) ~ 1, family="binomial",</pre>
listw=nb2listw(hopkins.rook.nb, style="B"), alpha=0.2, verbose=TRUE)
glmME <- glm(c(hopkins_part) ~ 1 + fitted(MEbinom1), family="binomial")</pre>
anova(glmME, test="Chisq")
anova(glmbase, glmME, test="Chisq")
## End(Not run)
```

126 moran

## **Description**

A simple function to compute Moran's I, called by moran. test and moran. mc;

$$I = \frac{n}{\sum_{i=1}^{n} \sum_{j=1}^{n} w_{ij}} \frac{\sum_{i=1}^{n} \sum_{j=1}^{n} w_{ij} (x_i - \bar{x})(x_j - \bar{x})}{\sum_{i=1}^{n} (x_i - \bar{x})^2}$$

#### Usage

```
moran(x, listw, n, S0, zero.policy=NULL, NAOK=FALSE)
```

## **Arguments**

x a numeric vector the same length as the neighbours list in listw

listw a listw object created for example by nb2listw

n number of zones

S0 global sum of weights

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged value

of zones without neighbours, if FALSE assign NA

NAOK if 'TRUE' then any 'NA' or 'NaN' or 'Inf' values in x are passed on to the

foreign function. If 'FALSE', the presence of 'NA' or 'NaN' or 'Inf' values is

regarded as an error.

#### Value

a list of

I Moran's I

K sample kurtosis of x

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 17.

#### See Also

```
moran.test, moran.mc
```

```
data(oldcol)
col.W <- nb2listw(COL.nb, style="W")
crime <- COL.OLD$CRIME
str(moran(crime, col.W, length(COL.nb), Szero(col.W)))
is.na(crime) <- sample(1:length(crime), 10)
str(moran(crime, col.W, length(COL.nb), Szero(col.W), NAOK=TRUE))</pre>
```

moran.mc 127

moran.mc	Permutation test for Moran's I statistic	

## **Description**

A permutation test for Moran's I statistic calculated by using nsim random permutations of x for the given spatial weighting scheme, to establish the rank of the observed statistic in relation to the nsim simulated values.

## Usage

```
moran.mc(x, listw, nsim, zero.policy=NULL, alternative="greater",
    na.action=na.fail, spChk=NULL, return_boot=FALSE)
```

### **Arguments**

x	a numeric vector the same length as the neighbours list in listw
listw	a listw object created for example by nb2listw
nsim	number of permutations
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
alternative	a character string specifying the alternative hypothesis, must be one of "greater" (default), or "less".
na.action	a function (default na.fail), can also be na.omit or na.exclude - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create noneighbour observations. Note that only weights lists created without using the glist argument to nb2listw may be subsetted. na.pass is not permitted because it is meaningless in a permutation test.
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
return_boot	return an object of class boot from the equivalent permutation bootstrap rather

### Value

A list with class htest and mc.sim containing the following components:

than an object of class htest

statistic the value of the observed Moran's I.

parameter the rank of the observed Moran's I.

p.value the pseudo p-value of the test.

alternative a character string describing the alternative hypothesis.

method a character string giving the method used.

data.name a character string giving the name(s) of the data, and the number of simulations.

res nsim simulated values of statistic, final value is observed statistic

128 moran.plot

### Author(s)

Roger Bivand < Roger . Bivand @nhh . no >

#### References

```
Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 63-5.
```

### See Also

```
moran, moran.test
```

## **Examples**

```
data(oldcol)
colw <- nb2listw(COL.nb, style="W")</pre>
nsim <- 99
set.seed(1234)
sim1 <- moran.mc(COL.OLD$CRIME, listw=colw, nsim=nsim)</pre>
mean(sim1$res[1:nsim])
var(sim1$res[1:nsim])
summary(sim1$res[1:nsim])
colold.lags <- nblag(COL.nb, 3)</pre>
set.seed(1234)
sim2 <- moran.mc(COL.OLD$CRIME, nb2listw(colold.lags[[2]],</pre>
 style="W"), nsim=nsim)
summary(sim2$res[1:nsim])
sim3 <- moran.mc(COL.OLD$CRIME, nb2listw(colold.lags[[3]],</pre>
 style="W"), nsim=nsim)
summary(sim3$res[1:nsim])
```

moran.plot

Moran scatterplot

### **Description**

A plot of spatial data against its spatially lagged values, augmented by reporting the summary of influence measures for the linear relationship between the data and the lag. If zero policy is TRUE, such observations are also marked if they occur.

## Usage

```
moran.plot(x, listw, zero.policy=NULL, spChk=NULL, labels=NULL, xlab=NULL, ylab=NULL, quiet=NULL, ...)
```

moran.plot 129

## **Arguments**

X	a numeric vector the same length as the neighbours list in listw
listw	a listw object created for example by nb2listw
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
labels	character labels for points with high influence measures, if set to FALSE, no labels are plotted for points with large influence
xlab	label for x axis
ylab	label for x axis
quiet	default NULL, use !verbose global option value; if TRUE, output of summary of influence object suppressed
	further graphical parameters as in par()

### Value

The function returns an influence object from influence.measures.

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## References

Anselin, L. 1996. The Moran scatterplot as an ESDA tool to assess local instability in spatial association. pp. 111–125 in M. M. Fischer, H. J. Scholten and D. Unwin (eds) Spatial analytical perspectives on GIS, London, Taylor and Francis; Anselin, L. 1995. Local indicators of spatial association, Geographical Analysis, 27, 93–115

## See Also

localmoran, influence.measures

```
data(afcon)
moran.plot(afcon$totcon, nb2listw(paper.nb),
  labels=as.character(afcon$name), pch=19)
moran.plot(as.vector(scale(afcon$totcon)), nb2listw(paper.nb),
  labels=as.character(afcon$name), xlim=c(-2, 4), ylim=c(-2,4), pch=19)
```

moran.test

moran.test	Moran's I test for spatial autocorrelation
------------	--

# Description

Moran's test for spatial autocorrelation using a spatial weights matrix in weights list form. The assumptions underlying the test are sensitive to the form of the graph of neighbour relationships and other factors, and results may be checked against those of moran.mc permutations.

# Usage

```
moran.test(x, listw, randomisation=TRUE, zero.policy=NULL,
   alternative="greater", rank = FALSE, na.action=na.fail, spChk=NULL, adjust.n=TRUE)
```

## **Arguments**

X	a numeric vector the same length as the neighbours list in listw
listw	a listw object created for example by nb2listw
randomisation	variance of I calculated under the assumption of randomisation, if FALSE normality
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
alternative	a character string specifying the alternative hypothesis, must be one of greater (default), less or two.sided.
rank	logical value - default FALSE for continuous variables, if TRUE, uses the adaptation of Moran's I for ranks suggested by Cliff and Ord (1981, p. 46)
na.action	a function (default na.fail), can also be na.omit or na.exclude - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create noneighbour observations. Note that only weights lists created without using the glist argument to nb2listw may be subsetted. If na.pass is used, zero is substituted for NA values in calculating the spatial lag
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
adjust.n	default TRUE, if FALSE the number of observations is not adjusted for no- neighbour observations, if TRUE, the number of observations is adjusted

## Value

A list with class htest containing the following components:

statistic the value of the standard deviate of Moran's I.

p.value the p-value of the test.

estimate the value of the observed Moran's I, its expectation and variance under the method assumption.

moran.test 131

alternative a character string describing the alternative hypothesis.

method a character string giving the assumption used for calculating the standard devi-

ate.

data.name a character string giving the name(s) of the data.

### Note

Var(I) is taken from Cliff and Ord (1969, p. 28), and Goodchild's CATMOG 47 (1986), see also Upton & Fingleton (1985) p. 171; it agrees with SpaceStat, see Tutorial workbook Chapter 22; VI is the second crude moment minus the square of the first crude moment. The derivation of the test (Cliff and Ord, 1981, p. 18) assumes that the weights matrix is symmetric. For inherently non-symmetric matrices, such as k-nearest neighbour matrices, listw2U() can be used to make the matrix symmetric.

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 21.

#### See Also

```
moran, moran.mc, listw2U
```

```
data(oldcol)
coords.OLD <- cbind(COL.OLD$X, COL.OLD$Y)</pre>
moran.test(COL.OLD$CRIME, nb2listw(COL.nb, style="W"))
moran.test(COL.OLD$CRIME, nb2listw(COL.nb, style="B"))
moran.test(COL.OLD$CRIME, nb2listw(COL.nb, style="C"))
moran.test(COL.OLD$CRIME, nb2listw(COL.nb, style="S"))
moran.test(COL.OLD$CRIME, nb2listw(COL.nb, style="W"),
randomisation=FALSE)
colold.lags <- nblag(COL.nb, 3)</pre>
moran.test(COL.OLD$CRIME, nb2listw(colold.lags[[2]],
style="W"))
moran.test(COL.OLD$CRIME, nb2listw(colold.lags[[3]],
style="W"))
print(is.symmetric.nb(COL.nb))
COL.k4.nb <- knn2nb(knearneigh(coords.OLD, 4))</pre>
print(is.symmetric.nb(COL.k4.nb))
moran.test(COL.OLD$CRIME, nb2listw(COL.k4.nb, style="W"))
moran.test(COL.OLD$CRIME, nb2listw(COL.k4.nb, style="W"),
randomisation=FALSE)
cat("Note: non-symmetric weights matrix, use listw2U()")
moran.test(COL.OLD$CRIME, listw2U(nb2listw(COL.k4.nb,
style="W")))
moran.test(COL.OLD$CRIME, listw2U(nb2listw(COL.k4.nb,
```

mstree mstree

```
style="W")), randomisation=FALSE)
ranks <- rank(COL.OLD$CRIME)
names(ranks) <- rownames(COL.OLD)
moran.test(ranks, nb2listw(COL.nb, style="W"), rank=TRUE)
crime <- COL.OLD$CRIME
is.na(crime) <- sample(1:length(crime), 10)
res <- try(moran.test(crime, nb2listw(COL.nb, style="W"),
na.action=na.fail))
res
moran.test(crime, nb2listw(COL.nb, style="W"), zero.policy=TRUE,
na.action=na.omit)
moran.test(crime, nb2listw(COL.nb, style="W"), zero.policy=TRUE,
na.action=na.exclude)
moran.test(crime, nb2listw(COL.nb, style="W"), na.action=na.pass)</pre>
```

mstree

Find the minimal spanning tree

## **Description**

The minimal spanning tree is a connected graph with n nodes and n-1 edges. This is a smaller class of possible partitions of a graph by pruning edges with high dissimilarity. If one edge is removed, the graph is partioned in two unconnected subgraphs. This function implements the algorithm due to Prim (1987).

## Usage

```
mstree(nbw, ini = NULL)
```

## Arguments

nbw An object of listw class returned by nb2listw function. See this help for

details.

ini The initial node in the minimal spanning tree.

## **Details**

The minimum spanning tree algorithm.

Input a connected graph.

Begin a empty set of nodes.

Add an arbitrary note in this set.

While are nodes not in the set, find a minimum cost edge connecting a node in the set and a node out of the set and add this node in the set.

The set of edges is a minimum spanning tree.

mstree 133

### Value

A matrix with n-1 rows and tree columns. Each row is two nodes and the cost, i. e. the edge and it cost.

## Author(s)

Renato M. Assuncao and Elias T. Krainski

#### References

R. C. Prim (1957) Shortest connection networks and some generalisations. In: Bell System Technical Journal, 36, pp. 1389-1401

```
### loading data
require(maptools)
bh <- readShapePoly(system.file("etc/shapes/bhicv.shp",</pre>
      package="spdep")[1])
### data padronized
dpad <- data.frame(scale(bh@data[,5:8]))</pre>
### neighboorhod list
bh.nb <- poly2nb(bh)</pre>
### calculing costs
lcosts <- nbcosts(bh.nb, dpad)</pre>
### making listw
nb.w <- nb2listw(bh.nb, lcosts, style="B")</pre>
### find a minimum spanning tree
system.time(mst.bh <- mstree(nb.w,5))</pre>
dim(mst.bh)
head(mst.bh)
tail(mst.bh)
### the mstree plot
par(mar=c(0,0,0,0))
plot(mst.bh, coordinates(bh), col=2,
     cex.lab=.7, cex.circles=0.035, fg="blue")
plot(bh, border=gray(.5), add=TRUE)
```

nb.set.operations

nb.set.operations Set operations on neighborhood objects

## **Description**

Set operations on neighbors list objects

### Usage

```
intersect.nb(nb.obj1,nb.obj2)
union.nb(nb.obj1,nb.obj2)
setdiff.nb(nb.obj1,nb.obj2)
complement.nb(nb.obj)
```

## **Arguments**

nb.obj a neighbor list created from any of the neighborhood list funtions nb.obj1 a neighbor list created from any of the neighborhood list funtions nb.obj2 a neighbor list created from any of the neighborhood list funtions

#### **Details**

These functions perform set operations on each element of a neighborlist. The arguments must be neighbor lists created from the same coordinates, and the region id attributes must be identical.

## Value

nb.obj A new neighborlist created from the set operations on the input neighbor list(s)

#### Author(s)

Nicholas Lewin-Koh <nikko@hailmail.net>

### See Also

```
intersect.nb, union.nb, setdiff.nb
```

```
example(columbus)
coords <- coordinates(columbus)
col.tri.nb <- tri2nb(coords)
oldpar <- par(mfrow=c(1,2))
col.soi.nb <- graph2nb(soi.graph(col.tri.nb, coords))
plot(columbus, border="grey")
plot(col.soi.nb, coords, add=TRUE)
title(main="Sphere of Influence Graph")
plot(columbus, border="grey")</pre>
```

nb2blocknb 135

```
plot(complement.nb(col.soi.nb), coords, add=TRUE)
title(main="Complement of Sphere of Influence Graph")
par(mfrow=c(2,2))
col2 <- droplinks(col.gal.nb, 21)
plot(intersect.nb(col.gal.nb, col2), coords)
title(main="Intersect")
plot(union.nb(col.gal.nb, col2), coords)
title(main="Union")
plot(setdiff.nb(col.gal.nb, col2), coords)
title(main="Set diff")
par(oldpar)</pre>
```

nb2blocknb

Block up neighbour list for location-less observations

## **Description**

The function blocks up a neighbour list for know spatial locations to create a new neighbour list for multiple location-less observations know to belong to the spatial locations, using the identification tags of the locations as the key.

### Usage

```
nb2blocknb(nb, ID, row.names = NULL)
```

### **Arguments**

nb an object of class nb with a list of integer vectors containing neighbour region

number ids

ID identification tags of the locations for the location-less observations;

sort(unique(as.character(ID))) must be identical to sort(as.character(attr(nb, "region.id")));

same length as row.names if provided.

row.names character vector of observation ids to be added to the neighbours list as attribute

region.id, default seq(1, nrow(x)); same length as ID if provided.

### **Details**

Assume that there is a list of unique locations, then a neighbour list can build for that, to create an input neighbour list. This needs to be "unfolded", so that observations belonging to each unique location are observation neighbours, and observations belonging to the location neighbours of the unique location in question are also observation neighbours, finally removing the observation itself (because it should not be its own neighbour). This scenario also arises when say only post codes are available, and some post codes contain multiple observations, where all that is known is that they belong to a specific post code, not where they are located within it (given that the post code locations are known).

nb2INLA

### Value

The function returns an object of class nb with a list of integer vectors containing neighbour observation number ids.

#### Author(s)

```
Roger Bivand < Roger . Bivand@nhh . no >
```

### See Also

```
knn2nb, dnearneigh, cell2nb, tri2nb, poly2nb
```

## **Examples**

```
## Not run:
data(boston)
summary(as.vector(table(boston.c$TOWN)))
townaggr <- aggregate(boston.utm, list(town=boston.c$TOWN), mean)</pre>
block.rel <- graph2nb(relativeneigh(as.matrix(townaggr[,2:3])),</pre>
 as.character(townaggr[,1]), sym=TRUE)
block.rel
print(is.symmetric.nb(block.rel))
plot(block.rel, as.matrix(townaggr[,2:3]))
points(boston.utm, pch=18, col="lightgreen")
block.nb <- nb2blocknb(block.rel, as.character(boston.c$TOWN))</pre>
block.nb
print(is.symmetric.nb(block.nb))
plot(block.nb, boston.utm)
points(boston.utm, pch=18, col="lightgreen")
moran.test(boston.c$CMEDV, nb2listw(boston.soi))
moran.test(boston.c$CMEDV, nb2listw(block.nb))
## End(Not run)
```

nb2INLA

Output spatial neighbours for INLA

# Description

Output spatial neighbours for INLA

## Usage

```
nb2INLA(file, nb)
```

### **Arguments**

```
file file where adjacency matrix will be stored nb an object of class nb
```

nb2lines 137

## Value

Nothing is returned but a file will be created with the representation of the adjacency matrix as required by INLA for its spatial models.

## Author(s)

Virgilio Gomez-Rubio

### References

http://www.r-inla.org

## **Examples**

```
example(columbus)
td <- tempdir()
x <- nb2INLA(paste(td, "columbus-INLA.adj", sep="/"), col.gal.nb)</pre>
```

nb2lines

Use arc-type shapefiles for import and export of weights

## Description

Use arc-type shapefiles for import and export of weights, storing spatial entity coordinates in the arcs, and the entity indices in the data frame.

# Usage

```
nb2lines(nb, wts, coords, proj4string=CRS(as.character(NA)))
listw2lines(listw, coords, proj4string=CRS(as.character(NA)))
df2sn(df, i="i", i_ID="i_ID", j="j", wt="wt")
```

# Arguments

nb	a neighbour object of class nb
wts	list of general weights corresponding to neighbours
coords	matrix of region point coordinates
proj4string	Object of class CRS; holding a valid proj4 string
listw	a listw object of spatial weights
df	a data frame read from a shapefile, derived from the output of nb2lines
i	character name of column in df with from entity index
i_ID	character name of column in df with from entity region ID
j	character name of column in df with to entity index
wt	character name of column in df with weights

nb2listw

#### **Details**

The maptools package function writeSpatialShape is used to transport out the list of lines made by nb2lines or listw2lines, which is a simple wrapper function. The neighbour and weights objects may be retrieved by converting the specified columns of the data slot of the SpatialLines-DataFrame object into a spatial.neighbour object, which is then converted into a weights list object.

### Value

nb2lines and listw2lines return a SpatialLinesDataFrame object; its data slot contains a data frame with the from and to indices of the neighbour links and their weights. df2sn converts the data retrieved from reading the data from df back into a spatial.neighbour object.

#### Note

Original idea due to Gidske Leknes Andersen, Department of Biology, University of Bergen, Norway

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

```
sn2listw, readShapeLines
```

## **Examples**

```
require(maptools)
example(columbus)
coords <- coordinates(columbus)
res <- listw2lines(nb2listw(col.gal.nb), coords)
summary(res)
fn <- paste(tempdir(), "nbshape", sep="/")
writeLinesShape(res, fn=fn)
inMap <- readShapeLines(fn)
summary(inMap)
diffnb(sn2listw(df2sn(as(inMap, "data.frame")))$neighbours, col.gal.nb)</pre>
```

nb2listw

Spatial weights for neighbours lists

## Description

The function supplements a neighbours list with spatial weights for the chosen coding scheme.

# Usage

```
nb2listw(neighbours, glist=NULL, style="W", zero.policy=NULL)
```

nb2listw 139

### **Arguments**

neighbours an object of class nb

glist list of general weights corresponding to neighbours

style style can take values "W", "B", "C", "U", "minmax" and "S"

zero.policy default NULL, use global option value; if FALSE stop with error for any empty

neighbour sets, if TRUE permit the weights list to be formed with zero-length

weights vectors

#### **Details**

Starting from a binary neighbours list, in which regions are either listed as neighbours or are absent (thus not in the set of neighbours for some definition), the function adds a weights list with values given by the coding scheme style chosen. B is the basic binary coding, W is row standardised (sums over all links to n), C is globally standardised (sums over all links to n), U is equal to C divided by the number of neighbours (sums over all links to unity), while S is the variance-stabilizing coding scheme proposed by Tiefelsdorf et al. 1999, p. 167-168 (sums over all links to n).

If zero policy is set to TRUE, weights vectors of zero length are inserted for regions without neighbour in the neighbours list. These will in turn generate lag values of zero, equivalent to the sum of products of the zero row t(rep(0, length=length(neighbours))) %\*% x, for arbitraty numerical vector x of length length(neighbours). The spatially lagged value of x for the zero-neighbour region will then be zero, which may (or may not) be a sensible choice.

If the sum of the glist vector for one or more observations is zero, a warning message is issued. The consequence for later operations will be the same as if no-neighbour observations were present and the zero.policy argument set to true.

The "minmax" style is based on Kelejian and Prucha (2010), and divides the weights by the minimum of the maximum row sums and maximum column sums of the input weights. It is similar to the C and U styles; it is also available in Stata.

#### Value

A listw object with the following members:

style one of W, B, C, U, S, minmax as above

neighbours the input neighbours list

weights the weights for the neighbours and chosen style, with attributes set to report the

type of relationships (binary or general, if general the form of the glist argu-

ment), and style as above

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

### References

Tiefelsdorf, M., Griffith, D. A., Boots, B. 1999 A variance-stabilizing coding scheme for spatial link matrices, Environment and Planning A, 31, pp. 165–180; Kelejian, H. H., and I. R. Prucha. 2010. Specification and estimation of spatial autoregressive models with autoregressive and heteroskedastic disturbances. Journal of Econometrics, 140: pp. 53–130.

nb2mat

#### See Also

```
summary.nb, read.gal
```

## **Examples**

```
example(columbus)
coords <- coordinates(columbus)</pre>
cards <- card(col.gal.nb)</pre>
col.w <- nb2listw(col.gal.nb)</pre>
plot(cards, unlist(lapply(col.w$weights, sum)),xlim=c(0,10),
ylim=c(0,10), xlab="number of links", ylab="row sums of weights")
col.b <- nb2listw(col.gal.nb, style="B")</pre>
points(cards, unlist(lapply(col.b$weights, sum)), col="red")
col.c <- nb2listw(col.gal.nb, style="C")</pre>
points(cards, unlist(lapply(col.c$weights, sum)), col="green")
col.u <- nb2listw(col.gal.nb, style="U")</pre>
points(cards, unlist(lapply(col.u$weights, sum)), col="orange")
col.s <- nb2listw(col.gal.nb, style="S")</pre>
points(cards, unlist(lapply(col.s$weights, sum)), col="blue")
legend(x=c(0, 1), y=c(7, 9), legend=c("W", "B", "C", "U", "S"),
col=c("black", "red", "green", "orange", "blue"), pch=rep(1,5))
summary(nb2listw(col.gal.nb, style="minmax"))
dlist <- nbdists(col.gal.nb, coords)</pre>
dlist <- lapply(dlist, function(x) 1/x)</pre>
col.w.d <- nb2listw(col.gal.nb, glist=dlist)</pre>
summary(unlist(col.w$weights))
summary(unlist(col.w.d$weights))
# introducing other conditions into weights - only earlier sales count
# see http://sal.uiuc.edu/pipermail/openspace/2005-October/000610.html
data(baltimore)
set.seed(211)
dates <- sample(1:500, nrow(baltimore), replace=TRUE)</pre>
nb_15nn <- knn2nb(knearneigh(cbind(baltimore$X, baltimore$Y), k=15))</pre>
glist <- vector(mode="list", length=length(nb_15nn))</pre>
for (i in seg(along=nb_15nn))
  glist[[i]] <- ifelse(dates[i] > dates[nb_15nn[[i]]], 1, 0)
listw_15nn_dates <- nb2listw(nb_15nn, glist=glist, style="B")</pre>
which(lag(listw_15nn_dates, baltimore$PRICE) == 0.0)
which(sapply(glist, sum) == 0)
ex <- which(sapply(glist, sum) == 0)[1]
dates[ex]
dates[nb_15nn[[ex]]]
```

nb2mat

Spatial weights matrices for neighbours lists

### **Description**

The function generates a weights matrix for a neighbours list with spatial weights for the chosen coding scheme.

nb2mat 141

### Usage

```
nb2mat(neighbours, glist=NULL, style="W", zero.policy=NULL)
listw2mat(listw)
```

## **Arguments**

neighbours an object of class nb

glist list of general weights corresponding to neighbours

style style can take values W, B, C, and S

zero.policy default NULL, use global option value; if FALSE stop with error for any empty

neighbour sets, if TRUE permit the weights list to be formed with zero-length

weights vectors

listw a listw object from for example nb2listw

#### **Details**

Starting from a binary neighbours list, in which regions are either listed as neighbours or are absent (thus not in the set of neighbours for some definition), the function creates an n by n weights matrix with values given by the coding scheme style chosen. B is the basic binary coding, W is row standardised, C is globally standardised, while S is the variance-stabilizing coding scheme proposed by Tiefelsdorf et al. 1999, p. 167-168.

The function leaves matrix rows as zero for any regions with zero neighbours fore zero.policy TRUE. These will in turn generate lag values of zero, equivalent to the sum of products of the zero row t(rep(0, length=length(neighbours))) %\*% x, for arbitraty numerical vector x of length length(neighbours). The spatially lagged value of x for the zero-neighbour region will then be zero, which may (or may not) be a sensible choice.

## Value

An n by n matrix, where n=length(neighbours)

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Tiefelsdorf, M., Griffith, D. A., Boots, B. 1999 A variance-stabilizing coding scheme for spatial link matrices, Environment and Planning A, 31, pp. 165-180.

#### See Also

nb2listw

nb2WB

## **Examples**

```
example(columbus)
coords <- coordinates(columbus)
col005 <- dnearneigh(coords, 0, 0.5, attr(col.gal.nb, "region.id"))
summary(col005)
col005.w.mat <- nb2mat(col005, zero.policy=TRUE)
table(round(apply(col005.w.mat, 1, sum)))</pre>
```

nb2WB

Output spatial weights for WinBUGS

## **Description**

Output spatial weights for WinBUGS

# Usage

```
nb2WB(nb)
listw2WB(listw)
```

## **Arguments**

nb an object of class nb

listw a listw object from for example nb2listw

### Value

A list suitable for convering using dput for WinBUGS

### Author(s)

Virgilio Gomez-Rubio

#### References

http://www.mrc-bsu.cam.ac.uk/bugs/winbugs/geobugs12manual.pdf

# See Also

dput

```
example(columbus)
x <- nb2WB(col.gal.nb)
dput(x, control=NULL)
x <- listw2WB(nb2listw(col.gal.nb))
dput(x, control=NULL)</pre>
```

nbcosts 143

nbcosts

Compute cost of edges

## **Description**

The cost of each edge is the distance between it nodes. This function compute this distance using a data.frame with observations vector in each node.

## Usage

# Arguments

nb	An object of nb class. See poly2nb for details.
data	A matrix with observations in the nodes.
id	Node index to compute the cost
id.neigh	Idex of neighbours nodes of node id
method	Character or function to declare distance method. If method is character, method must be "mahalanobis" or "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowisk". If method is one of "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowisk", see dist for details, because this function as used to compute the distance. If method="mahalanobis", the mahalanobis distance is computed between neighbour areas. If method is a function, this function is used to compute the distance.
p	The power of the Minkowski distance.
cov	The covariance matrix used to compute the mahalanobis distance.
inverted	logical. If 'TRUE', 'cov' is supposed to contain the inverse of the covariance

# Value

A object of nbdist class. See nbdists for details.

matrix.

## Author(s)

Elias T. Krainski and Renato M. Assuncao

### See Also

See Also as nbdists, nb2listw

144 nbdists

nbdists

Spatial link distance measures

## Description

Given a list of spatial neighbour links (a neighbours list of object type nb), the function returns the Euclidean distances along the links in a list of the same form as the neighbours list. If longlat = TRUE, Great Circle distances are used.

## Usage

```
nbdists(nb, coords, longlat = NULL)
```

# **Arguments**

nb an object of class nb

coords matrix of point coordinates or a SpatialPoints object

longlat TRUE if point coordinates are longitude-latitude decimal degrees, in which case

distances are measured in kilometers; if coords is a SpatialPoints object, the

value is taken from the object itself

### Value

A list with class nbdist

# Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## See Also

```
summary.nb, nb2listw
```

```
example(columbus)
coords <- coordinates(columbus)
dlist <- nbdists(col.gal.nb, coords)
dlist <- lapply(dlist, function(x) 1/x)
stem(unlist(dlist))</pre>
```

nblag 145

nblag

Higher order neighbours lists

## **Description**

The function creates higher order neighbour lists, where higher order neighbours are only lags links from each other on the graph described by the input neighbours list. It will refuse to lag neighbours lists with the attribute self.included set to TRUE. nblag\_cumul cumulates neighbour lists to a single neighbour list ("nb" object).

## Usage

```
nblag(neighbours, maxlag)
nblag_cumul(nblags)
```

### Arguments

neighbours input neighbours list of class nb maxlag the maximum lag to be constructed

nblags a list of neighbour lists as output by nblag

#### Value

returns a list of lagged neighbours lists each with class nb

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no> and Giovanni Millo

## See Also

```
summary.nb
```

```
example(columbus)
coords <- coordinates(columbus)
summary(col.gal.nb, coords)
plot(columbus, border="grey")
plot(col.gal.nb, coords, add=TRUE)
title(main="GAL order 1 (black) and 2 (red) links")
col.lags <- nblag(col.gal.nb, 2)
lapply(col.lags, print)
summary(col.lags[[2]], coords)
plot(col.lags[[2]], coords, add=TRUE, col="red", lty=2)
cuml <- nblag_cumul(col.lags)
cuml</pre>
```

nc.sids

nc.sids

North Carolina SIDS data

### **Description**

(Use example(nc.sids) to read the data set from shapefile, together with import of two different list of neighbours).

The nc.sids data frame has 100 rows and 21 columns. It contains data given in Cressie (1991, pp. 386-9), Cressie and Read (1985) and Cressie and Chan (1989) on sudden infant deaths in North Carolina for 1974-78 and 1979-84. The data set also contains the neighbour list given by Cressie and Chan (1989) omitting self-neighbours (ncCC89.nb), and the neighbour list given by Cressie and Read (1985) for contiguities (ncCR85.nb). The data are ordered by county ID number, not alphabetically as in the source tables sidspolys is a "polylist" object of polygon boundaries, and sidscents is a matrix of their centroids.

#### **Usage**

data(nc.sids)

#### **Format**

This data frame contains the following columns:

**SP\\_ID** SpatialPolygons ID

CNTY\\_ID county ID

east eastings, county seat, miles, local projection

north northings, county seat, miles, local projection

L\\_id Cressie and Read (1985) L index

M\\_id Cressie and Read (1985) M index

names County names

AREA County polygon areas in degree units

**PERIMETER** County polygon perimeters in degree units

**CNTY**\\_ Internal county ID

NAME County names

FIPS County ID

FIPSNO County ID

CRESS\\_ID Cressie papers ID

**BIR74** births, 1974-78

**SID74** SID deaths, 1974-78

**NWBIR74** non-white births, 1974-78

BIR79 births, 1979-84

SID79 SID deaths, 1979-84

NWBIR79 non-white births, 1979-84

NY\_data

#### Source

Cressie, N (1991), Statistics for spatial data. New York: Wiley, pp. 386–389; Cressie, N, Chan NH (1989) Spatial modelling of regional variables. *Journal of the American Statistical Association*, 84, 393–401; Cressie, N, Read, TRC (1985) Do sudden infant deaths come in clusters? *Statistics and Decisions* Supplement Issue 2, 333–349; http://sal.agecon.uiuc.edu/datasets/sids.zip.

## **Examples**

```
require(maptools)
nc.sids <- readShapePoly(system.file("etc/shapes/sids.shp", package="spdep")[1],
    ID="FIPSNO", proj4string=CRS("+proj=longlat +ellps=clrk66"))
rn <- sapply(slot(nc.sids, "polygons"), function(x) slot(x, "ID"))
ncCC89_nb <- read.gal(system.file("etc/weights/ncCC89.gal", package="spdep")[1],
    region.id=rn)
ncCR85_nb <- read.gal(system.file("etc/weights/ncCR85.gal", package="spdep")[1],
    region.id=rn)
## Not run:
plot(nc.sids, border="grey")
plot(nc.sids, border="grey")
plot(nc.sids, border="grey")
plot(nc.sids, border="grey")
plot(nc.sids, border="grey")
plot(ncCC89_nb, coordinates(nc.sids), add=TRUE, col="blue")
## End(Not run)</pre>
```

NY data

New York leukemia data

# **Description**

New York leukemia data taken from the data sets supporting Waller and Gotway 2004 (the data should be loaded by running example(NY\_data) to demonstrate spatial data import techniques).

# Usage

```
data(NY_data)
```

### Format

A data frame with 281 observations on the following 12 variables, and the binary coded spatial weights used in the source.

```
AREANAME name of census tract
AREAKEY unique FIPS code for each tract
X x-coordinate of tract centroid (in km)
Y y-coordinate of tract centroid (in km)
POP8 population size (1980 U.S. Census)
TRACTCAS number of cases 1978-1982
```

148 oldcol

PROPCAS proportion of cases per tract

PCTOWNHOME percentage of people in each tract owning their own home

PCTAGE65P percentage of people in each tract aged 65 or more

Z ransformed propoprtions

AVGIDIST average distance between centroid and TCE sites

PEXPOSURE "exposure potential": inverse distance between each census tract centroid and the nearest TCE site, IDIST, transformed via log(100\*IDIST)

#### **Details**

The examples section shows how the DBF files from the book website for Chapter 9 were converted into the nydata data frame and the listw\_NY spatial weights list.

#### Source

```
http://www.sph.emory.edu/~lwaller/ch9index.htm
```

#### References

Waller, L. and C. Gotway (2004) *Applied Spatial Statistics for Public Health Data*. New York: John Wiley and Sons.

### **Examples**

```
## NY leukemia
library(foreign)
nydata <- read.dbf(system.file("etc/misc/nydata.dbf", package="spdep")[1])
coordinates(nydata) <- c("X", "Y")
nyadjmat <- as.matrix(read.dbf(system.file("etc/misc/nyadjwts.dbf",
    package="spdep")[1])[-1])
ID <- as.character(names(read.dbf(system.file("etc/misc/nyadjwts.dbf",
    package="spdep")[1]))[-1])
identical(substring(ID, 2, 10), substring(as.character(nydata$AREAKEY), 2, 10))
nyadjlw <- mat2listw(nyadjmat, as.character(nydata$AREAKEY))
listw_NY <- nb2listw(nyadjlw$neighbours, style="B")</pre>
```

oldcol

Columbus OH spatial analysis data set - old numbering

## **Description**

The COL. OLD data frame has 49 rows and 22 columns. The observations are ordered and numbered as in the original analyses of the data set in the SpaceStat documentation and in Anselin, L. 1988 Spatial econometrics: methods and models, Dordrecht: Kluwer. Unit of analysis: 49 neighbourhoods in Columbus, OH, 1980 data. In addition the data set includes COL.nb, the neighbours list as used in Anselin (1988).

oldcol 149

### Usage

data(oldcol)

#### **Format**

This data frame contains the following columns:

**AREA** computed by ArcView

**PERIMETER** computed by ArcView

**COLUMBUS.** internal polygon ID (ignore)

COLUMBUS.I another internal polygon ID (ignore)

POLYID yet another polygon ID

**NEIG** neighborhood id value (1-49); conforms to id value used in Spatial Econometrics book.

**HOVAL** housing value (in \\$1,000)

**INC** household income (in \\$1,000)

CRIME residential burglaries and vehicle thefts per thousand households in the neighborhood

**OPEN** open space in neighborhood

PLUMB percentage housing units without plumbin

**DISCBD** distance to CBD

**X** x coordinate (in arbitrary digitizing units, not polygon coordinates)

Y y coordinate (in arbitrary digitizing units, not polygon coordinates)

**AREA** neighborhood area (computed by SpaceStat)

**NSA** north-south dummy (North=1)

**NSB** north-south dummy (North=1)

**EW** east-west dummy (East=1)

**CP** core-periphery dummy (Core=1)

THOUS constant=1,000

NEIGNO NEIG+1,000, alternative neighborhood id value

**PERIM** polygon perimeter (computed by SpaceStat)

### **Details**

The row names of COL.OLD and the region.id attribute of COL.nb are set to columbus\$NEIGNO.

## Note

All source data files prepared by Luc Anselin, Spatial Analysis Laboratory, Department of Agricultural and Consumer Economics, University of Illinois, Urbana-Champaign, http://sal.agecon.uiuc.edu/datasets/columbus.zip.

## Source

Anselin, Luc. 1988. Spatial econometrics: methods and models. Dordrecht: Kluwer Academic, Table 12.1 p. 189.

p.adjustSP

	1.0		_
n	ada	justSF	2
υ.	au	iustor	

Adjust local association measures' p-values

#### **Description**

Make an adjustment to local association measures' p-values based on the number of neighbours (+1) of each region, rather than the total number of regions.

## Usage

```
p.adjustSP(p, nb, method = "none")
```

### **Arguments**

p vector of p-values

nb a list of neighbours of class nb

method correction method as defined in p. adjust: "The adjustment methods include the

Bonferroni correction ('"bonferroni"') in which the p-values are multiplied by the number of comparisons. Four less conservative corrections are also included by Holm (1979) ('"holm"'), Hochberg (1988) ('"hochberg"'), Hommel (1988) ('"hommel"') and Benjamini & Hochberg (1995) ('"fdr"'), respectively. A pass-

through option ('"none"') is also included."

#### Value

A vector of corrected p-values using only the number of neighbours + 1.

#### Author(s)

Danlin Yu and Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

```
p.adjust, localG, localmoran
```

```
data(afcon)
oid <- order(afcon$id)
resG <- as.vector(localG(afcon$totcon, nb2listw(include.self(paper.nb))))
non <- format.pval(pnorm(2*(abs(resG)), lower.tail=FALSE), 2)
bon <- format.pval(p.adjustSP(pnorm(2*(abs(resG)), lower.tail=FALSE),
    paper.nb, "bonferroni"), 2)
tot <- format.pval(p.adjust(pnorm(2*(abs(resG)), lower.tail=FALSE),
    "bonferroni", n=length(resG)), 2)
data.frame(resG, non, bon, tot, row.names=afcon$name)[oid,]</pre>
```

plot.mst 151

plot.mst	Plot the Minimum Spanning Tree
----------	--------------------------------

# Description

This function plots a MST, the nodes are circles and the edges are segments.

# Usage

```
## S3 method for class 'mst'
plot(x, coords, label.areas = NULL,
    cex.circles = 1, cex.labels = 1, ...)
```

# Arguments

X	Object of mst class.
coords	A two column matrix with the coordinates of nodes.
label.areas	A vector with the labels of nodes
cex.circles	The length of circles to plot.
cex.labels	The length of nodes labels ploted.
	Further arguments passed to plotting funcitons.

# Author(s)

Elias T. Krainski and Renato M. Assuncao

### See Also

See Also as skater and mstree

## **Examples**

### see example in mstree function documentation

plot.nb	Plot a neighbours list	
---------	------------------------	--

# Description

A function to plot a neighbours list given point coordinates to represent the region in two dimensions; plot.listwis a wrapper that passes its neighbours component to plot.nb.

152 plot.nb

### Usage

```
## S3 method for class 'nb'
plot(x, coords, col="black", points=TRUE, add=FALSE, arrows=FALSE,
  length=0.1, xlim=NULL, ylim=NULL, ...)
## S3 method for class 'listw'
plot(x, coords, col="black", points=TRUE, add=FALSE, arrows=FALSE,
  length=0.1, xlim=NULL, ylim=NULL, ...)
```

## **Arguments**

Х

coords matrix of region point coordinates

col plotting colour

points (logical) add points to plot

add (logical) add to existing plot

arrows (logical) draw arrowheads for asymmetric neighbours

an object of class nb or (for plot.listw) class listw

length in plot inches of arrow heads drawn for asymmetric neighbours lists

xlim, ylim plot window bounds

... further graphical parameters as in par(..)

# Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

# See Also

```
summary.nb
```

```
example(columbus)
coords <- coordinates(columbus)
plot(col.gal.nb, coords)
title(main="GAL order 1 links with first nearest neighbours in red")
col.knn <- knearneigh(coords, k=1)
plot(knn2nb(col.knn), coords, add=TRUE, col="red", length=0.08)</pre>
```

plot.skater 153

plot.skater	Plot the object of skater class	
-------------	---------------------------------	--

# Description

This function displays the results of the skater function. The subgraphs are plotted with different colours.

# Usage

```
## S3 method for class 'skater'
plot(x, coords, label.areas = NULL,
    groups.colors, cex.circles = 1, cex.labels = 1, ...)
```

# Arguments

X	An object of skater class.
coords	A matrix of two colums with coordinates of nodes.
label.areas	A vector of labels of nodes.
groups.colors	A vector with colors of groups ou sub-graphs.
cex.circles	The length of circles with represent the nodes.
cex.labels	The length of labels of nodes.
	Further arguments passed to plotting funcitons.

# Author(s)

Elias T. Krainski and Renato M. Assuncao

### See Also

See Also as skater and mstree

```
### see example in the skater function documentation
```

poly2nb

poly2nb	Construct neighbours list from polygon list

## **Description**

The function builds a neighbours list based on regions with contiguous boundaries, that is sharing one or more boundary point. The current function is in part interpreted and may run slowly for many regions or detailed boundaries, but from 0.2-16 should not fail because of lack of memory when single polygons are built of very many border coordinates.

## Usage

```
poly2nb(pl, row.names = NULL, snap=sqrt(.Machine$double.eps),
   queen=TRUE, useC=TRUE, foundInBox=NULL)
```

## Arguments

•	3	
	pl	list of polygons of class extending SpatialPolygons
	row.names	character vector of region ids to be added to the neighbours list as attribute region.id, default $seq(1, nrow(x))$ ; if polys has a region.id attribute, it is copied to the neighbours list.
	snap	boundary points less than snap distance apart are considered to indicate contiguity
	queen	if TRUE, a single shared boundary point meets the contiguity condition, if FALSE, more than one shared point is required; note that more than one shared boundary point does not necessarily mean a shared boundary line
	useC	default TRUE, doing the work loop in $C$ , may be set to false to revert to $R$ code calling two $C$ functions in an $n * k$ work loop, where $k$ is the average number of candidate neighbours
	foundInBox	default NULL using R code, possibly parallelised if a <b>snow</b> cluster is available, otherwise a list of length (n-1) with integer vectors of candidate neighbours ( $j > i$ ), or NULL if all candidates were ( $j < i$ ) (as created by the poly_findInBoxGEOS function in <b>rgeos</b> for clean polygons)

## Value

A neighbours list with class nb. See card for details of "nb" objects.

## Note

From 0.5-8, the function includes faster bounding box indexing and other improvements contributed by Micah Altman. If a cluster is provided using set.ClusterOption, it will be used for finding candidate bounding box overlaps for exact testing for contiguity.

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no > with contributions from Micah Altman

predict.sarIm 155

## See Also

```
summary.nb, card
```

## **Examples**

```
example(columbus)
coords <- coordinates(columbus)</pre>
xx <- poly2nb(columbus)</pre>
dxx <- diffnb(xx, col.gal.nb)</pre>
plot(columbus, border="grey")
plot(col.gal.nb, coords, add=TRUE)
plot(dxx, coords, add=TRUE, col="red")
title(main=paste("Differences (red) in Columbus GAL weights (black)",
"and polygon generated queen weights", sep="\n"))
xxx <- poly2nb(columbus, queen=FALSE)</pre>
dxxx <- diffnb(xxx, col.gal.nb)</pre>
plot(columbus, border = "grey")
plot(col.gal.nb, coords, add = TRUE)
plot(dxxx, coords, add = TRUE, col = "red")
title(main=paste("Differences (red) in Columbus GAL weights (black)",
 "and polygon generated rook weights", sep="\n"))
cards <- card(xx)</pre>
maxconts <- which(cards == max(cards))</pre>
if(length(maxconts) > 1) maxconts <- maxconts[1]</pre>
fg <- rep("grey", length(cards))</pre>
fg[maxconts] <- "red"
fg[xx[[maxconts]]] <- "green"</pre>
plot(columbus, col=fg)
title(main="Region with largest number of contiguities")
example(nc.sids)
system.time(xxnb <- poly2nb(nc.sids))</pre>
plot(nc.sids)
plot(xxnb, coordinates(nc.sids), add=TRUE, col="blue")
```

predict.sarlm

Prediction for spatial simultaneous autoregressive linear model objects

## Description

predict.sarlm() calculates predictions as far as is at present possible for for spatial simultaneous autoregressive linear model objects, using Haining's terminology for decomposition into trend, signal, and noise — see reference.

# Usage

```
## S3 method for class 'sarlm'
predict(object, newdata = NULL, listw = NULL,
  zero.policy = NULL, legacy=TRUE, power=NULL, order=250,
```

156 predict.sarlm

```
tol=.Machine$double.eps^(3/5), #pred.se=FALSE, lagImpact=NULL,
...)
## S3 method for class 'sarlm.pred'
print(x, ...)
## S3 method for class 'sarlm.pred'
as.data.frame(x, ...)
```

#### **Arguments**

object sarlm object returned by lagsarlm or errorsarlm

newdata Data frame in which to predict — if NULL, predictions are for the data on which

the model was fitted

listw a listw object created for example by nb2listw

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged value

of zones without neighbours, if FALSE (default) assign NA - causing the func-

tion to terminate with an error

legacy (Only applies to lag and Durbin (mixed) models) default TRUE: use ad-hoc

predictor, if FALSE use DGP-based predictor

power (Only applies to lag and Durbin (mixed) models) use powerWeights, if default

NULL, set FALSE if object\$method is "eigen", otherwise TRUE

order Power series maximum limit if power is TRUE

tol Tolerance for convergence of power series if power is TRUE

x the object to be printed

... further arguments passed through

#### **Details**

In the following, the trend is the non-spatial smooth, the signal is the spatial smooth, and the noise is the residual. The fit returned is the sum of the trend and the signal.

The function approaches prediction first by dividing invocations between those with or without newdata. When no newdata is present, the response variable may be reconstructed as the sum of the trend, the signal, and the noise (residuals). Since the values of the response variable are known, their spatial lags are used to calculate signal components (Cressie 1993, p. 564). For the error model, trend =  $X\beta$ , and signal =  $\lambda Wy - \lambda WX\beta$ . For the lag and mixed models, trend =  $X\beta$ , and signal =  $\rho Wy$ .

This approach differs from the design choices made in other software, for example GeoDa, which does not use observations of the response variable, and corresponds to the newdata situation described below.

When however newdata is used for prediction, no observations of the response variable being predicted are available. Consequently, while the trend components are the same, the signal cannot take full account of the spatial smooth. In the error model and Durbin error model, the signal is set to zero, since the spatial smooth is expressed in terms of the error:  $(I - \lambda W)^{-1}\varepsilon$ .

In the lag model, the signal can be expressed in the following way (for legacy=TRUE):

$$(I - \rho W)y = X\beta + \varepsilon$$

predict.sarlm 157

,

$$y = (I - \rho W)^{-1} X \beta + (I - \rho W)^{-1} \varepsilon$$

giving a feasible signal component of:

$$\rho Wy = \rho W(I - \rho W)^{-1} X\beta$$

For legacy=FALSE, the trend is computed first as:

$$X\beta$$

next the prediction using the DGP:

$$(I - \rho W)^{-1} X \beta$$

and the signal is found as the difference between prediction and trend. The numerical results for the legacy and DGP methods are identical.

setting the error term to zero. This also means that predictions of the signal component for lag and mixed models require the inversion of an n-by-n matrix.

Because the outcomes of the spatial smooth on the error term are unobservable, this means that the signal values for newdata are incomplete. In the mixed model, the spatially lagged RHS variables influence both the trend and the signal, so that the root mean square prediction error in the examples below for this case with newdata is smallest, although the model was not the best fit

### Value

predict.sarlm() returns a vector of predictions with two attribute vectors of trend and signal values with class sarlm.pred. print.sarlm.pred is a print function for this class, printing and returning a data frame with columns: "fit", "trend" and "signal".

## Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

#### References

Haining, R. 1990 Spatial data analysis in the social and environmental sciences, Cambridge: Cambridge University Press, p. 258; Cressie, N. A. C. 1993 Statistics for spatial data, Wiley, New York.

## See Also

errorsarlm, lagsarlm

158 probmap

### **Examples**

```
data(oldcol)
lw <- nb2listw(COL.nb)</pre>
COL.lag.eig <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD, lw)
COL.mix.eig <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD, lw,
  type="mixed")
COL.err.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD, lw)
COL.SDerr.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD, lw,
 etype="emixed")
print(p1 <- predict(COL.mix.eig))</pre>
print(p2 <- predict(COL.mix.eig, newdata=COL.OLD, listw=lw))</pre>
AIC(COL.mix.eig)
sqrt(deviance(COL.mix.eig)/length(COL.nb))
sqrt(sum((COL.OLD$CRIME - as.vector(p1))^2)/length(COL.nb))
sqrt(sum((COL.OLD$CRIME - as.vector(p2))^2)/length(COL.nb))
AIC(COL.err.eig)
sqrt(deviance(COL.err.eig)/length(COL.nb))
sqrt(sum((COL.OLD$CRIME - as.vector(predict(COL.err.eig)))^2)/length(COL.nb))
sqrt(sum((COL.OLD$CRIME - as.vector(predict(COL.err.eig, newdata=COL.OLD,
  listw=lw)))^2)/length(COL.nb))
AIC(COL.SDerr.eig)
sqrt(deviance(COL.SDerr.eig)/length(COL.nb))
sqrt(sum((COL.OLD$CRIME - as.vector(predict(COL.SDerr.eig)))^2)/length(COL.nb))
sqrt(sum((COL.OLD$CRIME - as.vector(predict(COL.SDerr.eig, newdata=COL.OLD,
  listw=lw)))^2)/length(COL.nb))
AIC(COL.lag.eig)
sqrt(deviance(COL.lag.eig)/length(COL.nb))
sqrt(sum((COL.OLD$CRIME - as.vector(predict(COL.lag.eig)))^2)/length(COL.nb))
sqrt(sum((COL.OLD$CRIME - as.vector(predict(COL.lag.eig, newdata=COL.OLD,
  listw=lw)))^2)/length(COL.nb))
p3 <- predict(COL.mix.eig, newdata=COL.OLD, listw=lw, legacy=FALSE)
all.equal(p2, p3)
p4 <- predict(COL.mix.eig, newdata=COL.OLD, listw=lw, legacy=FALSE, power=TRUE)</pre>
all.equal(p2, p4)
p5 <- predict(COL.mix.eig, newdata=COL.OLD, listw=lw, legacy=TRUE, power=TRUE)
all.equal(p2, p5)
```

probmap

Probability mapping for rates

# Description

The function returns a data frame of rates for counts in populations at risk with crude rates, expected counts of cases, relative risks, and Poisson probabilities.

#### **Usage**

```
probmap(n, x, row.names=NULL, alternative="less")
```

probmap 159

### **Arguments**

n a numeric vector of counts of cases
x a numeric vector of populations at risk

row.names row names passed through to output data frame

alternative default "less", may be set to "greater"

### **Details**

The function returns a data frame, from which rates may be mapped after class intervals have been chosen. The class intervals used in the examples are mostly taken from the referenced source.

#### Value

raw (crude) rates

expCount expected counts of cases assuming global rate

relRisk relative risks: ratio of observed and expected counts of cases multiplied by 100 pmap Poisson probability map values: probability of getting a more "extreme" count

than actually observed - one-tailed, default alternative observed "less" than ex-

pected

## Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

#### References

Bailey T, Gatrell A (1995) Interactive Spatial Data Analysis, Harlow: Longman, pp. 300-303.

## See Also

```
EBest, EBlocal, ppois
```

prunecost

```
cols <- cm.colors(6)
plot(auckland, col=cols[findInterval(res$relRisk, brks, all.inside=TRUE)])
legend("bottomleft", fill=cols, legend=leglabs(brks), bty="n")
title(main="Standardised mortality ratios for Auckland child deaths")
brks <- c(0,0.05,0.1,0.2,0.8,0.9,0.95,1)
cols <- cm.colors(7)
plot(auckland, col=cols[findInterval(res$pmap, brks, all.inside=TRUE)])
legend("bottomleft", fill=cols, legend=leglabs(brks), bty="n")
title(main="Poisson probabilities for Auckland child mortality")</pre>
```

prunecost

Compute cost of prune each edge

### **Description**

If any edge are dropped, the MST are pruned. This generate a two subgraphs. So, it makes a tree graphs and tree dissimilarity values are computed, one for each graph. The dissimilarity is the sum over sqared differences between the observactions in the nodes and mean vector of observations in the graph. The dissimilarity of original graph and the sum of dissimilarity of subgraphs are returned.

## Usage

# Arguments

edges A matrix with 2 colums with each row is one edge
data A data frame with observations in the nodes.

method Character or function to declare distance method. If method is character, method

must be "mahalanobis" or "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowisk". If method is one of "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowisk", see dist for details, because this function as used to compute the distance. If method="mahalanobis", the mahalanobis distance is computed between neighbour areas. If method is a function,

this function is used to compute the distance.

p The power of the Minkowski distance.

cov The covariance matrix used to compute the mahalanobis distance.

inverted logical. If 'TRUE', 'cov' is supposed to contain the inverse of the covariance

matrix.

#### Value

A vector with the differences between the dissimilarity of all nodes and the dissimilarity sum of all subgraphs obtained by pruning one edge each time.

prunemst 161

### Author(s)

Elias T. Krainski and Renato M. Assuncao

### See Also

See Also as prunemst

## **Examples**

```
d <- data.frame(a=-2:2, b=runif(5))
e <- matrix(c(1,2, 2,3, 3,4, 4,5), ncol=2, byrow=TRUE)
sum(sweep(d, 2, colMeans(d))^2)
prunecost(e, d)</pre>
```

prunemst

Prune a Minimun Spanning Tree

## **Description**

This function deletes a first edge and makes two subsets of edges. Each subset is a Minimun Spanning Treee.

## Usage

```
prunemst(edges, only.nodes = TRUE)
```

# Arguments

edges A matrix with two colums with each row is one edge

only.nodes If only.nodes=FALSE, return a edges and nodes of each MST resulted. If

only.nodes=TRUE, return a two sets of nodes. Defalt is TRUE

### Value

A list of length two. If only.nodes=TRUE each element is a vector of nodes. If only.nodes=FALSE each element is a list with nodes and edges.

## Author(s)

Elias T. Krainski and Renato M. Assuncao

## See Also

See Also as mstree

read.gal

### **Examples**

```
e <- matrix(c(2,3, 1,2, 3,4, 4,5), ncol=2, byrow=TRUE)
e
prunemst(e)
prunemst(e, only.nodes=FALSE)</pre>
```

read.gal

Read a GAL lattice file into a neighbours list

## **Description**

The function read.gal() reads a GAL lattice file into a neighbours list for spatial analysis. It will read old and new style (GeoDa) GAL files. The function read.geoda is a helper file for reading comma separated value data files, calling read.csv().

## Usage

```
read.gal(file, region.id=NULL, override.id=FALSE)
read.geoda(file, row.names=NULL, skip=0)
```

## **Arguments**

file	name of file with GAL lattice data
region.id	region IDs in specified order to coerse neighbours list order and numbering to that of the region.id
override.id	override any given (or NULL) region.id, collecting region.id numbering and order from the GAL file.
row.names	as in row.names in $read.csv()$ , typically a character string naming the column of the file to be used
skip	skip number of lines, as in read.csv()

## Details

Luc Anselin (2003): Spatial Analysis Laboratory, Department of Agricultural and Consumer Economics, University of Illinois, Urbana-Champaign, http://sal.agecon.uiuc.edu/weights/howto.html; Luc Anselin (2003) *GeoDa 0.9 User's Guide*, pp. 80–81, Spatial Analysis Laboratory, Department of Agricultural and Consumer Economics, University of Illinois, Urbana-Champaign, http://agec221.agecon.uiuc.edu/csiss/pdf/geoda093.pdf; GAL-Geographical Algorithms Library, University of Newcastle

#### Value

The function read.gal() returns an object of class nb with a list of integer vectors containing neighbour region number ids. The function read.geoda returns a data frame, and issues a warning if the returned object has only one column.

read.gwt2nb 163

#### Note

Example data downloaded from: http://sal.agecon.uiuc.edu/weights/zips/us48.zip

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

```
summary.nb
```

## **Examples**

```
us48.fipsno <- read.geoda(system.file("etc/weights/us48.txt",</pre>
package="spdep")[1])
us48.q <- read.gal(system.file("etc/weights/us48_q.GAL", package="spdep")[1],
us48.fipsno$Fipsno)
us48.r <- read.gal(system.file("etc/weights/us48_rk.GAL", package="spdep")[1],
us48.fipsno$Fipsno)
data(state)
if (as.numeric(paste(version$major, version$minor, sep="")) < 19) {</pre>
m50.48 <- match(us48.fipsno$"State.name", state.name)</pre>
} else {
m50.48 <- match(us48.fipsno$"State_name", state.name)</pre>
}
plot(us48.q, as.matrix(as.data.frame(state.center))[m50.48,])
plot(diffnb(us48.r, us48.q),
as.matrix(as.data.frame(state.center))[m50.48,], add=TRUE, col="red")
title(main="Differences between rook and queen criteria imported neighbours lists")
```

read.gwt2nb

Read and write spatial neighbour files

# Description

The "gwt" functions read and write GeoDa GWT files (the example file baltk4.GWT was downloaded from the site given in the reference), and the "dat" functions read and write Matlab sparse matrix files as used by James LeSage's Spatial Econometrics Toolbox (the example file wmat.dat was downloaded from the site given in the reference). The body of the files after any headers should have three columns separated by white space, and the third column must be numeric in the locale of the reading platform (correct decimal separator).

# Usage

```
read.gwt2nb(file, region.id=NULL)
write.sn2gwt(sn, file, shpfile=NULL, ind=NULL, useInd=FALSE, legacy=FALSE)
read.dat2listw(file)
write.sn2dat(sn, file)
```

read.gwt2nb

### **Arguments**

file name of file with weights data

region.id region IDs

sn a spatial.neighbour object

shpfile character string: if not given Shapefile name taken from GWT file for this dataset

ind character string: region id indicator field name

useInd default FALSE, if TRUE, write region.id attribute ID key tags to output file

(use in OpenGeoDa will depend on the shapefile having the field named in the

ind argument matching the exported tags)

legacy default FALSE; if TRUE, header has single field with number of observations

only

### **Details**

Now attempts to honour the region.id argument given when reading GWT files.

#### Value

read.gwt2nb returns a neighbour "nb" object with the generalised weights stored as a list element called "dlist" of the "GeoDa" attribute.

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## References

Luc Anselin (2003) *GeoDa 0.9 User's Guide*, pp. 80–81, Spatial Analysis Laboratory, Department of Agricultural and Consumer Economics, University of Illinois, Urbana-Champaign, http://agec221.agecon.uiuc.edu/csiss/pdf/geoda093.pdf; also http://spatial-econometrics.com/data/contents.html

#### See Also

```
read.gal
```

```
data(baltimore)
STATION <- baltimore$STATION
gwt1 <- read.gwt2nb(system.file("etc/weights/baltk4.GWT", package="spdep")[1],
    STATION)
cat(paste("Neighbours list symmetry;", is.symmetric.nb(gwt1, FALSE, TRUE),
    "\n"))
listw1 <- nb2listw(gwt1, style="B", glist=attr(gwt1, "GeoDa")$dist)
tmpGWT <- tempfile()
write.sn2gwt(listw2sn(listw1), tmpGWT)
gwt2 <- read.gwt2nb(tmpGWT, STATION)</pre>
```

residuals.sarlm 165

```
cat(paste("Neighbours list symmetry;", is.symmetric.nb(gwt2, FALSE, TRUE),
   "\n"))
diffnb(gwt1, gwt2)
data(oldcol)
tmpMAT <- tempfile()
COL.W <- nb2listw(COL.nb)
write.sn2dat(listw2sn(COL.W), tmpMAT)
listwmat1 <- read.dat2listw(tmpMAT)
diffnb(listwmat1$neighbours, COL.nb, verbose=TRUE)
listwmat2 <- read.dat2listw(system.file("etc/weights/wmat.dat",
   package="spdep")[1])
diffnb(listwmat1$neighbours, listwmat2$neighbours, verbose=TRUE)</pre>
```

residuals.sarlm

Access functions for spatial simultaneous autoregressive linear model objects

### **Description**

Access functions for residuals, deviance, coefficients and fitted values from spatial simultaneous autoregressive linear model objects

## Usage

```
## S3 method for class 'sarlm'
residuals(object, ...)
## S3 method for class 'sarlm'
deviance(object, ...)
## S3 method for class 'sarlm'
coef(object, ...)
## S3 method for class 'sarlm'
vcov(object, ...)
## S3 method for class 'sarlm'
fitted(object, ...)
```

# Arguments

object sarlm object returned by lagsarlm or errorsarlm ... further arguments passed through

## Value

Revelant vectors of numerical values.

#### Note

fitted.sarlm() returns the difference between residuals() for the same object and the response variable; predict.sarlm() returns a decomposition into trend and signal for the fit.

166 Rotation

### Author(s)

```
Roger Bivand, <Roger.Bivand@nhh.no>
```

#### See Also

```
errorsarlm, lagsarlm, predict.sarlm
```

Rotation

Rotate a set of point by a certain angle

## **Description**

Rotate a set of XY coordinates by an angle (in radians)

## Usage

```
Rotation(xy, angle)
```

# Arguments

xy A 2-columns matrix or data frame containing a set of X and Y coordinates.

angle Numeric. A scalar giving the angle at which the points should be rotated. The

angle is in radians.

### Value

A 2-columns matrix of the same size as xy giving the rotated coordinates.

### Author(s)

F. Guillaume Blanchet

```
set.seed(1)
### Create a set of coordinates
coords<-cbind(runif(20),runif(20))
### Create a series of angles
rad<-seq(0,pi,l=20)

opar <- par(mfrow=c(5,4))
for(i in rad){
coords.rot<-Rotation(coords,i)
plot(coords.rot)
}
par(opar)</pre>
```

```
### Rotate the coordinates by an angle of 90 degrees
coords.90<-Rotation(coords,90*pi/180)
coords.90

plot(coords,xlim=range(rbind(coords.90,coords)[,1]),ylim=range(rbind(coords.90,coords)[,2]),asp=1)
points(coords.90,pch=19)</pre>
```

sacsarlm

Spatial simultaneous autoregressive SAC model estimation

## **Description**

Maximum likelihood estimation of spatial simultaneous autoregressive "SAC/SARAR" models of the form:

$$y = \rho W 1 y + X \beta + u, u = \lambda W 2 u + \varepsilon$$

where  $\rho$  and  $\lambda$  are found by nlminb or optim() first, and  $\beta$  and other parameters by generalized least squares subsequently

### Usage

```
sacsarlm(formula, data = list(), listw, listw2 = NULL, na.action, type="sac",
method = "eigen", quiet = NULL, zero.policy = NULL, tol.solve = 1e-10,
llprof=NULL, interval1=NULL, interval2=NULL, trs1=NULL, trs2=NULL,
control = list())
```

## **Arguments**

formula	a symbolic description of the model to be fit. The details of model specification are given for lm()
data	an optional data frame containing the variables in the model. By default the variables are taken from the environment which the function is called
listw	a listw object created for example by nb2listw
listw2	a listw object created for example by nb2listw, if not given, set to the same spatial weights as the listw argument
na.action	a function (default options ("na.action")), can also be na.omit or na.exclude with consequences for residuals and fitted values - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create no-neighbour observations. Note that only weights lists created without using the glist argument to nb2listw may be subsetted.
type	default "sac", may be set to "sacmixed" for the Manski model to include the spatially lagged independent variables added to X using listw; when "sacmixed", the lagged intercept is dropped for spatial weights style "W", that is row-standardised weights, but otherwise included

method "eigen" (default) - the Jacobian is computed as the product of (1 - rho\*eigenvalue)

using eigenw, and "spam" or "Matrix" for strictly symmetric weights lists of styles "B" and "C", or made symmetric by similarity (Ord, 1975, Appendix C) if possible for styles "W" and "S", using code from the spam or Matrix packages to calculate the determinant; "LU" provides an alternative sparse matrix decomposition approach. In addition, there are "Chebyshev" and Monte Carlo "MC"

approximate log-determinant methods.

quiet default NULL, use !verbose global option value; if FALSE, reports function

values during optimization.

zero.policy default NULL, use global option value; if TRUE assign zero to the lagged

value of zones without neighbours, if FALSE (default) assign NA - causing

sacsarlm() to terminate with an error

tol. solve the tolerance for detecting linear dependencies in the columns of matrices to be

inverted - passed to solve() (default=1.0e-10). This may be used if necessary to extract coefficient standard errors (for instance lowering to 1e-12), but errors in solve() may constitute indications of poorly scaled variables: if the variables have scales differing much from the autoregressive coefficient, the values in this matrix may be very different in scale, and inverting such a matrix is analytically possible by definition, but numerically unstable; rescaling the RHS variables

alleviates this better than setting tol.solve to a very small value

11prof default NULL, can either be an integer, to divide the feasible ranges into a grid of

points, or a two-column matrix of spatial coefficient values, at which to evaluate

the likelihood function

trs1, trs2 default NULL, if given, vectors for each weights object of powered spatial

weights matrix traces output by trW; when given, used in some Jacobian meth-

ods

interval1, interval2

default is NULL, search intervals for each weights object for autoregressive pa-

ameters

control list of extra control arguments - see section below

#### **Details**

Because numerical optimisation is used to find the values of lambda and rho, care needs to be shown. It has been found that the surface of the 2D likelihood function often forms a "banana trench" from (low rho, high lambda) through (high rho, high lambda) to (high rho, low lambda) values. In addition, sometimes the banana has optima towards both ends, one local, the other global, and consequently the choice of the starting point for the final optimization becomes crucial. The default approach is not to use just (0, 0) as a starting point, nor the (rho, lambda) values from gstsls, which lie in a central part of the "trench", but either four values at (low rho, high lambda), (0, 0), (high rho, high lambda), and (high rho, low lambda), and to use the best of these start points for the final optimization. Optionally, nine points can be used spanning the whole (lower, upper) space.

## Value

A list object of class sarlm

type "sac"

rho lag simultaneous autoregressive lag coefficient error simultaneous autoregressive error coefficient

coefficients GLS coefficient estimates

rest.se asymptotic standard errors if ase=TRUE, otherwise approximate numeriacal Hessian-

based values

ase TRUE if method=eigen

LL log likelihood value at computed optimum

SSE GLS residual variance
SSE sum of squared GLS errors
number of parameters estimated

logLik\_lm.model

Log likelihood of the non-spatial linear model

AIC\_lm.model AIC of the non-spatial linear model

method the method used to calculate the Jacobian

call the call used to create this object

residuals GLS residuals

tarX model matrix of the GLS model tary response of the GLS model

y response of the linear model for  $\rho=0$ X model matrix of the linear model for  $\rho=0$ opt object returned from numerical optimisation

pars starting parameter values for final optimization, either given or found by trial

point evaluation

mxs if default input pars, optimal objective function values at trial points

se.fit Not used yet

rho. se if ase=TRUE, the asymptotic standard error of  $\rho$ , otherwise approximate nume-

riacal Hessian-based value

lambda.se if ase=TRUE, the asymptotic standard error of  $\lambda$ 

resvar the asymptotic coefficient covariance matrix for (s2, rho, lambda, B)

zero.policy zero.policy for this model

aliased the aliased explanatory variables (if any)
LLNullLlm Log-likelihood of the null linear model

fdHess the numerical Hessian-based coefficient covariance matrix for (rho, lambda, B)

if computed

resvar asymptotic coefficient covariance matrix

optimHess FALSE

timings processing timings

na.action (possibly) named vector of excluded or omitted observations if non-default na.action

argument used

#### Control arguments

**fdHess:** default NULL, then set to (method != "eigen") internally; use fdHess to compute an approximate Hessian using finite differences when using sparse matrix methods with fdHess from **nlme**; used to make a coefficient covariance matrix when the number of observations is large; may be turned off to save resources if need be

**LAPACK:** default FALSE; logical value passed to qr in the SSE log likelihood function

**Imult:** default 2; used for preparing the Cholesky decompositions for updating in the Jacobian function

cheb\_q: default 5; highest power of the approximating polynomial for the Chebyshev approximation

MC\_p: default 16; number of random variates

MC\_m: default 30; number of products of random variates matrix and spatial weights matrix

**super:** default FALSE using a simplicial decomposition for the sparse Cholesky decomposition, if TRUE, use a supernodal decomposition

opt\_method: default "nlminb", may be set to "L-BFGS-B" to use box-constrained optimisation in optim

opt\_control: default list(), a control list to pass to nlminb or optim

**pars:** default NULL, for which five trial starting values spanning the lower/upper range are tried and the best selected, starting values of  $\rho$  and  $\lambda$ 

**npars** default integer 4L, four trial points; if not default value, nine trial points

pre\_eig1, pre\_eig2 default NULL; may be used to pass pre-computed vectors of eigenvalues

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Anselin, L. 1988 Spatial econometrics: methods and models. (Dordrecht: Kluwer); LeSage J and RK Pace (2009) Introduction to Spatial Econometrics. CRC Press, Boca Raton

### See Also

help,

```
data(oldcol)
COL.sacW.eig <- sacsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    nb2listw(COL.nb, style="W"))
summary(COL.sacW.eig, correlation=TRUE)
W <- as(as_dgRMatrix_listw(nb2listw(COL.nb, style="W")), "CsparseMatrix")
trMatc <- trW(W, type="mult")
summary(impacts(COL.sacW.eig, tr=trMatc, R=2000), zstats=TRUE, short=TRUE)
COL.msacW.eig <- sacsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    nb2listw(COL.nb, style="W"), type="sacmixed")
summary(COL.msacW.eig, correlation=TRUE)
summary(impacts(COL.msacW.eig, tr=trMatc, R=2000), zstats=TRUE, short=TRUE)</pre>
```

set.spChkOption 171

## **Description**

Provides support for checking the mutual integrity of spatial neighbour weights and spatial data; similar mechanisms are used for passing global verbose and zero.policy options, and for providing access to a running cluster for embarrassingly parallel tasks.

## Usage

```
set.spChkOption(check)
get.spChkOption()
chkIDs(x, listw)
spNamedVec(var, data)
set.VerboseOption(check)
get.VerboseOption()
set.ZeroPolicyOption(check)
get.ZeroPolicyOption()
set.ClusterOption(cl)
get.ClusterOption()
set.rlecuyerSeedOption(seed)
get.rlecuyerSeedOption()
```

## Arguments

check	a logical value, TRUE or FALSE
cl	a cluster object created by makeCluster in <b>parallel</b>
seed	an integer vector of length 6 to seed the Lecuyer streams for cluster RNG
X	a vector the same length, or a two-dimensional array, or data frame with the same number of rows as the neighbours list in listw
listw	a listw object or nb object inheriting from "nb"
var	a character string or integer value for the column to be selected
data	a two-dimensional array or data frame containing var

#### **Details**

Analysis functions will have an spChk argument by default set to NULL, and will call get.spChkOption() to get the global spatial option for whether to check or not — this is initialised to FALSE, and consequently should not break anything. It can be changed to TRUE using set.spChkOption(TRUE), or the spChk argument can be assigned in analysis functions. spNamedVec() is provided to ensure that rownames are passed on to single columns taken from two-dimensional arrays and data frames.

172 similar.listw

### Value

set.spChkOption() returns the old logical value, get.spChkOption() returns the current logical value, and chkIDs() returns a logical value for the test lack of difference. spNamedVec() returns the selected column with the names set to the row names of the object from which it has been extracted.

#### Note

The motivation for this mechanism is provided by the observation that spatial objects on a map and their attribute data values need to be linked uniquely, to avoid spurious results. The reordering between the legacy Columbus data set used the earlier publications and that available for download from the Spacestat website is just one example of a common problem.

### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

```
data(oldcol)
rownames(COL.OLD)
data(columbus)
rownames(columbus)
get.spChkOption()
oldChk <- set.spChkOption(TRUE)</pre>
get.spChkOption()
chkIDs(COL.OLD, nb2listw(COL.nb))
chkIDs(columbus, nb2listw(col.gal.nb))
chkIDs(columbus, nb2listw(COL.nb))
tmp <- try(moran.test(spNamedVec("CRIME", COL.OLD), nb2listw(COL.nb)))</pre>
print(tmp)
tmp <- try(moran.test(spNamedVec("CRIME", columbus), nb2listw(col.gal.nb)))</pre>
tmp <- try(moran.test(spNamedVec("CRIME", columbus), nb2listw(COL.nb)))</pre>
print(tmp)
set.spChkOption(FALSE)
get.spChkOption()
moran.test(spNamedVec("CRIME", columbus), nb2listw(COL.nb))
tmp <- try(moran.test(spNamedVec("CRIME", columbus), nb2listw(COL.nb),</pre>
spChk=TRUE))
print(tmp)
set.spChkOption(oldChk)
get.spChkOption()
```

similar.listw 173

## **Description**

From Ord's 1975 paper, it is known that the Jacobian for SAR models may be found by "symmetrizing" by similarity (the eigenvalues of similar matrices are identical, so the Jacobian is too). This applies only to styles "W" and "S" with underlying symmetric binary neighbour relations or symmetric general neighbour relations (so no k-nearest neighbour relations). The function is invoked automatically within the SAR fitting functions, to call eigen on a symmetric matrix for the default eigen method, or to make it possible to use the Matrix method on weights that can be "symmetrized" in this way.

## Usage

```
similar.listw(listw)
```

## **Arguments**

listw

a listw object created for example by nb2listw

#### Value

a listw object

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Ord, J. K. 1975 Estimation methods for models of spatial interaction, *Journal of the American Statistical Association*, 70, 120-126

### See Also

lagsarlm, errorsarlm

```
data(oldcol)
COL.W <- nb2listw(COL.nb, style="W")
COL.S <- nb2listw(COL.nb, style="S")
sum(log(1 - 0.5 * eigenw(COL.W)))
sum(log(1 - 0.5 * eigenw(similar.listw(COL.W))))
W_J <- as(as_dsTMatrix_listw(similar.listw(COL.W)), "CsparseMatrix")
I <- as_dsCMatrix_I(dim(W_J)[1])
c(determinant(I - 0.5 * W_J, logarithm=TRUE)$modulus)
sum(log(1 - 0.5 * eigenw(COL.S)))
sum(log(1 - 0.5 * eigenw(similar.listw(COL.S))))
W_J <- as(as_dsTMatrix_listw(similar.listw(COL.S)), "CsparseMatrix")
c(determinant(I - 0.5 * W_J, logarithm=TRUE)$modulus)</pre>
```

174 skater

oval
(

## **Description**

This function implements a SKATER procedure for spatial clustering analysis. This procedure essentially begins with an edges set, a data set and a number of cuts. The output is an object of 'skater' class and is valid for input again.

## Usage

```
skater(edges, data, ncuts, crit, vec.crit, method = c("euclidean",
    "maximum", "manhattan", "canberra", "binary", "minkowski",
    "mahalanobis"), p = 2, cov, inverted = FALSE)
```

## **Arguments**

edges A matrix with 2 colums with each row is an edge data A data.frame with data observed over nodes.

ncuts The number of cuts

crit A scalar ow 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.

vec.crit A vector for evaluating criteria.

method Character or function to declare distance method. If method is character, method

must be "mahalanobis" or "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowisk". If method is one of "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowski", see dist for details, because this function as used to compute the distance. If method="mahalanobis", the mahalanobis distance is computed between neighbour areas. If method is a function,

this function is used to compute the distance.

p The power of the Minkowski distance.

The covariance matrix used to compute the mahalanobis distance.

inverted logical. If 'TRUE', 'cov' is supposed to contain the inverse of the covariance

matrix.

### Value

A object of skater 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.

A vector identifying the groups with are candidates to partition.

The total dissimilarity in each step of edge removal.

skater 175

#### Author(s)

Renato M. Assuncao and Elias T. Krainski

#### References

Assuncao, R.M., Lage J.P., and Reis, E.A. (2002). Analise de conglomerados espaciais via arvore geradora minima. Revista Brasileira de Estatistica, 62, 1-23.

Assuncao, R. M, Neves, M. C., Camara, G. and Freitas, C. da C. (2006). Efficient regionalization techniques for socio-economic geographical units using minimum spanning trees. International Journal of Geographical Information Science Vol. 20, No. 7, August 2006, 797-811

#### See Also

See Also as mstree

```
### loading data
require(maptools)
bh <- readShapePoly(system.file("etc/shapes/bhicv.shp",</pre>
      package="spdep")[1])
### data standardized
dpad <- data.frame(scale(bh@data[,5:8]))</pre>
### neighboorhod list
bh.nb <- poly2nb(bh)
### calculating costs
lcosts <- nbcosts(bh.nb, dpad)</pre>
### making listw
nb.w <- nb2listw(bh.nb, lcosts, style="B")</pre>
### find a minimum spanning tree
mst.bh <- mstree(nb.w,5)</pre>
### the mstree plot
par(mar=c(0,0,0,0))
plot(mst.bh, coordinates(bh), col=2,
     cex.lab=.7, cex.circles=0.035, fg="blue")
plot(bh, border=gray(.5), add=TRUE)
### three groups with no restriction
res1 <- skater(mst.bh[,1:2], dpad, 2)
### groups size
table(res1$groups)
### the skater plot
par(mar=c(0,0,0,0))
plot(res1, coordinates(bh), cex.circles=0.035, cex.lab=.7)
```

176 skater

```
### the skater plot, using other colors
plot(res1, coordinates(bh), cex.circles=0.035, cex.lab=.7,
     groups.colors=rainbow(length(res1$ed)))
### the Spatial Polygons plot
plot(bh, col=heat.colors(length(res1$edg))[res1$groups])
### EXPERT OPTIONS
### more one partition
res1b <- skater(res1, dpad, 1)
### length groups frequency
table(res1$groups)
table(res1b$groups)
### thee groups with minimum population
res2 <- skater(mst.bh[,1:2], dpad, 2, 200000, bh@data$Pop)
### thee groups with minimun number of areas
res3 <- skater(mst.bh[,1:2], dpad, 2, 3, rep(1,nrow(bh@data)))</pre>
### thee groups with minimun and maximun number of areas
res4 <- skater(mst.bh[,1:2], dpad, 2, c(20,50), rep(1,nrow(bh@data)))
table(res2$groups)
table(res3$groups)
table(res4$groups)
### if I want to get groups with 20 to 40 elements
res5 <- skater(mst.bh[,1:2], dpad, 2,</pre>
  c(20,40), rep(1,nrow(bh@data))) ## DON'T MAKE DIVISIONS
table(res5$groups)
### In this MST don't have groups with this restrictions
### In this case, first I do one division
### with the minimun criteria
res5a <- skater(mst.bh[,1:2], dpad, 1, 20, rep(1,nrow(bh@data)))
table(res5a$groups)
### and do more one division with the full criteria
res5b <- skater(res5a, dpad, 1, c(20, 40), rep(1,nrow(bh@data)))
table(res5b$groups)
### and do more one division with the full criteria
res5c <- skater(res5b, dpad, 1, c(20, 40), rep(1,nrow(bh@data)))
table(res5c$groups)
### It don't have another divison with this criteria
res5d <- skater(res5c, dpad, 1, c(20, 40), rep(1,nrow(bh@data)))</pre>
table(res5d$groups)
```

sp.correlogram 177

Spatial correlogram
---------------------

# Description

Spatial correlograms for Moran's I and the autocorrelation coefficient, with print and plot helper functions.

## Usage

```
sp.correlogram(neighbours, var, order = 1, method = "corr",
  style = "W", randomisation = TRUE, zero.policy = NULL, spChk=NULL)
## S3 method for class 'spcor'
plot(x, main, ylab, ylim, ...)
## S3 method for class 'spcor'
print(x, p.adj.method="none", ...)
```

### **Arguments**

neighbours	an object of class nb
var	a numeric vector
order	maximum lag order
method	"corr" for correlation, "I" for Moran's I, "C" for Geary's C
style	style can take values W, B, C, and S
randomisation	variance of I or C calculated under the assumption of randomisation, if FALSE normality
zero.policy	default NULL, use global option value; if FALSE stop with error for any empty neighbour sets, if TRUE permit the weights list to be formed with zero-length weights vectors
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
x	an object from sp.correlogram() of class spcor
p.adj.method	correction method as in p.adjust
main	an overall title for the plot
ylab	a title for the y axis
ylim	the y limits of the plot
	further arguments passed through

178 sp.correlogram

#### **Details**

The print function also calculates the standard deviates of Moran's I or Geary's C and a two-sided probability value, optionally using p.adjust to correct by the nymber of lags. The plot function plots a bar from the estimated Moran's I, or Geary's C value to +/- twice the square root of its variance (in previous releases only once, not twice). The table includes the count of included observations in brackets after the lag order. Care needs to be shown when interpreting results for few remaining included observations as lag order increases.

#### Value

returns a list of class spcor:

res for "corr" a vector of values; for "I", a matrix of estimates of "I", expectations,

and variances

method "I" or "corr"

cardnos list of tables of neighbour cardinalities for the lag orders used

var variable name

## Author(s)

Roger Bivand, <Roger.Bivand@nhh.no>

#### References

Cliff, A. D., Ord, J. K. 1981 *Spatial processes*, Pion, pp. 118–122, Martin, R. L., Oeppen, J. E. 1975 The identification of regional forecasting models using space-time correlation functions, *Transactions of the Institute of British Geographers*, 66, 95–118.

#### See Also

```
nblag, moran, p.adjust
```

```
example(nc.sids)
ft.SID74 <- sqrt(1000)*(sqrt(nc.sids$SID74/nc.sids$BIR74) +
 sqrt((nc.sids$SID74+1)/nc.sids$BIR74))
tr.SIDS74 <- ft.SID74*sqrt(nc.sids$BIR74)</pre>
cspc <- sp.correlogram(ncCC89_nb, tr.SIDS74, order=8, method="corr",</pre>
zero.policy=TRUE)
print(cspc)
plot(cspc)
Ispc <- sp.correlogram(ncCC89_nb, tr.SIDS74, order=8, method="I",</pre>
zero.policy=TRUE)
print(Ispc)
print(Ispc, "bonferroni")
plot(Ispc)
Cspc <- sp.correlogram(ncCC89_nb, tr.SIDS74, order=8, method="C",</pre>
zero.policy=TRUE)
print(Cspc)
```

sp.mantel.mc 179

```
print(Cspc, "bonferroni")
plot(Cspc)
drop.no.neighs <- !(1:length(ncCC89_nb) %in% which(card(ncCC89_nb) == 0))
sub.ncCC89.nb <- subset(ncCC89_nb, drop.no.neighs)
plot(sp.correlogram(sub.ncCC89.nb, subset(tr.SIDS74, drop.no.neighs),
    order=8, method="corr"))</pre>
```

sp.mantel.mc

Mantel-Hubert spatial general cross product statistic

## **Description**

A permutation test for the spatial general cross product statistic with Moran  $(C_{ij}=z_iz_j)$ , Geary  $(C_{ij}=(z_i-z_j)^2)$ , and Sokal  $(C_{ij}=|z_i-z_j|)$  criteria, for  $z_i=(x_i-\bar{x})/\sigma_x$ . plot.mc.sim is a helper function to plot the outcomes of the permutation test.

## Usage

```
sp.mantel.mc(var, listw, nsim, type = "moran", zero.policy = NULL,
  alternative = "greater", spChk=NULL, return_boot=FALSE)
## S3 method for class 'mc.sim'
plot(x, xlim, xlab, main, sub, ..., ptype="density")
```

### **Arguments**

var	a numeric vector the same length as the neighbours list in listw
listw	a listw object created for example by nb2listw
nsim	number of permutations
type	"moran", "geary" or "sokal" criteria for similarity
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE assign NA
alternative	a character string specifying the alternative hypothesis, must be one of "greater" (default), or "less".
spChk	should the data vector names be checked against the spatial objects for identity integrity, TRUE, or FALSE, default NULL to use get.spChkOption()
return_boot	return an object of class boot from the equivalent permutation bootstrap rather than an object of class htest
X	the object to be plotted
xlim	the range of the x axis
xlab	a title for the x axis
main	an overall title for the plot
sub	a sub title for the plot
ptype	either "density" or "hist"
	further arguments passed through

180 sp.mantel.mc

#### Value

A list with class htest and mc.sim containing the following components:

statistic the value of the observed Geary's C.

parameter the rank of the observed Geary's C.

alternative a character string describing the alternative hypothesis.

method a character string giving the method used.

data.name a character string giving the name(s) of the data, and the number of simulations.

p. value the pseudo p-value of the test.

res nsim simulated values of statistic, final value is observed statistic

estimate the mean and variance of the simulated distribution.

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 22-24, Haining, R. 1990 *Spatial data analysis in the social and environmental sciences*, Cambridge: Cambridge University Press, p. 230–1. The function has been checked against general matrix code posted to the r-help list by Ben Bolker on 1 May 2001; another mantel() function is in the vegan package.

## See Also

```
moran.mc, joincount.mc, geary.mc
```

```
data(oldcol)
sim1 <- sp.mantel.mc(COL.OLD$CRIME, nb2listw(COL.nb),
    nsim=99, type="geary", alternative="less")
sim1
plot(sim1)
sp.mantel.mc(COL.OLD$CRIME, nb2listw(COL.nb), nsim=99,
    type="sokal", alternative="less")
sp.mantel.mc(COL.OLD$CRIME, nb2listw(COL.nb), nsim=99,
    type="moran")</pre>
```

SpatialFiltering 181

SpatialFiltering	Semi-parametric spatial filtering	
------------------	-----------------------------------	--

## Description

The function selects eigenvectors in a semi-parametric spatial filtering approach to removing spatial dependence from linear models. Selection is by brute force by finding the single eigenvector reducing the standard variate of Moran's I for regression residuals most, and continuing until no candidate eigenvector reduces the value by more than tol. It returns a summary table from the selection process and a matrix of selected eigenvectors for the specified model.

## Usage

```
SpatialFiltering(formula, lagformula, data = list(), nb, glist = NULL, style = "C", zero.policy = NULL, tol = 0.1, zerovalue = 1e-04, ExactEV = FALSE, symmetric = TRUE, alpha=NULL, alternative="two.sided", verbose=NULL)
```

## **Arguments**

formula	a symbolic description of the model to be fit, assuming a spatial error represen- tation; when lagformula is given, it should include only the response and the intercept term
lagformula	An extra one-sided formula to be used when a spatial lag representation is desired; the intercept is excluded within the function if present because it is part of the formula argument, but excluding it explicitly in the lagformula argument in the presence of factors generates a collinear model matrix
data	an optional data frame containing the variables in the model
nb	an object of class nb
glist	list of general weights corresponding to neighbours
style	style can take values W, B, C, U, and S
zero.policy	default NULL, use global option value; if FALSE stop with error for any empty neighbour sets, if TRUE permit the weights list to be formed with zero-length weights vectors
tol	tolerance value for convergence of spatial filtering
zerovalue	eigenvectors with eigenvalues of an absolute value smaller than zerovalue will be excluded in eigenvector search
ExactEV	Set ExactEV=TRUE to use exact expectations and variances rather than the expectation and variance of Moran's I from the previous iteration, default FALSE
symmetric	Should the spatial weights matrix be forced to symmetry, default TRUE
alpha	if not NULL, used instead of the tol= argument as a stopping rule to choose all eigenvectors up to and including the one with a probability value exceeding alpha.
alternative	a character string specifying the alternative hypothesis, must be one of greater, less or two.sided (default).
verbose	default NULL, use global option value; if TRUE report eigenvectors selected

SpatialFiltering

#### Value

An SFResult object, with:

selection a matrix summarising the selection of eigenvectors for inclusion, with columns:

**Step** Step counter of the selection procedure

**SelEvec** number of selected eigenvector (sorted descending)

Eval its associated eigenvalue

MinMi value Moran's I for residual autocorrelation

**ZMinMi** standardized value of Moran's I assuming a normal approximation **pr(ZI)** probability value of the permutation-based standardized deviate for the given value of the alternative argument

**R2** R\^2 of the model including exogenous variables and eigenvectors

gamma regression coefficient of selected eigenvector in fit

The first row is the value at the start of the search

dataset a matrix of the selected eigenvectors in order of selection

## Author(s)

Yongwan Chun, Michael Tiefelsdorf, Roger Bivand

#### References

Tiefelsdorf M, Griffith DA. (2007) Semiparametric Filtering of Spatial Autocorrelation: The Eigenvector Approach. Environment and Planning A, 39 (5) 1193 - 1221. http://www.spatialfiltering.com

## See Also

```
lm, eigen, nb2listw, listw2U
```

```
example(columbus)
lmbase <- lm(CRIME ~ INC + HOVAL, data=columbus)
sarcol <- SpatialFiltering(CRIME ~ INC + HOVAL, data=columbus,
    nb=col.gal.nb, style="W", ExactEV=TRUE)
sarcol
lmsar <- lm(CRIME ~ INC + HOVAL + fitted(sarcol), data=columbus)
lmsar
anova(lmbase, lmsar)
lm.morantest(lmsar, nb2listw(col.gal.nb))
lagcol <- SpatialFiltering(CRIME ~ 1, ~ INC + HOVAL - 1, data=columbus,
    nb=col.gal.nb, style="W")
lagcol
lmlag <- lm(CRIME ~ INC + HOVAL + fitted(lagcol), data=columbus)
lmlag
anova(lmbase, lmlag)
lm.morantest(lmlag, nb2listw(col.gal.nb))</pre>
```

spautolm	Spatial conditional and simultaneous autoregression model estimation

## **Description**

Function taking family and weights arguments for spatial autoregression model estimation by Maximum Likelihood, using dense matrix methods, not suited to large data sets with thousands of observations. With one of the sparse matrix methods, larger numbers of observations can be handled, but the interval= argument should be set. The implementation is GLS using the single spatial coefficient value, here termed lambda, found by line search using optimize to maximise the log likelihood.

## Usage

```
spautolm(formula, data = list(), listw, weights,
na.action, family = "SAR", method="eigen", verbose = NULL, trs=NULL,
interval=NULL, zero.policy = NULL, tol.solve=.Machine$double.eps,
llprof=NULL, control=list())
## S3 method for class 'spautolm'
summary(object, correlation = FALSE, adj.se=FALSE,
Nagelkerke=FALSE, ...)
```

## **Arguments**

, amends	
formula	a symbolic description of the model to be fit. The details of model specification are given for $lm()$
data	an optional data frame containing the variables in the model. By default the variables are taken from the environment which the function is called.
listw	a listw object created for example by nb2listw
weights	an optional vector of weights to be used in the fitting process
na.action	a function (default options ("na.action")), can also be na.omit or na.exclude with consequences for residuals and fitted values - in these cases the weights list will be subsetted to remove NAs in the data. Note that only weights lists created without using the glist argument to nb2listw may be subsetted.
family	character string: either "SAR" or "CAR" for simultaneous or conditional autoregressions; "SMA" for spatial moving average added thanks to Jielai Ma - "SMA" is only implemented for method="eigen" because it necessarily involves dense matrices
method	character string: default "eigen" for use of dense matrices, "Matrix_J" or "spam" for sparse matrices (restricted to spatial weights symmetric or similar to symmetric) using methods in either the Matrix or spam packages; "Matrix" and "spam_update" provide updating Cholesky decomposition methods. Note that users should change from spam to Matrix. Values of method may also include "LU", which provides an alternative sparse matrix decomposition approach, and
	data listw weights na.action

the "Chebyshev" and Monte Carlo "MC" approximate log-determinant methods.

verbose default NULL, use global option value; if TRUE, reports function values during optimization. default NULL, if given, a vector of powered spatial weights matrix traces output trs, by trW; when given, used in some Jacobian methods interval search interval for autoregressive parameter when not using method="eigen"; default is c(-1,0.999), optimize will reset NA/NaN to a bound and gives a warning when the interval is poorly set; method="Matrix" will attempt to search for an appropriate interval, if find\\_interval=TRUE (fails on some platforms) zero.policy default NULL, use global option value; Include list of no-neighbour observations in output if TRUE — otherwise zero.policy is handled within the listw argument tol.solve the tolerance for detecting linear dependencies in the columns of matrices to be inverted - passed to solve() (default=double precision machine tolerance). Errors in solve() may constitute indications of poorly scaled variables: if the variables have scales differing much from the autoregressive coefficient, the values in this matrix may be very different in scale, and inverting such a matrix is analytically possible by definition, but numerically unstable; rescaling the RHS variables alleviates this better than setting tol.solve to a very small value llprof default NULL, can either be an integer, to divide the feasible range into llprof points, or a sequence of spatial coefficient values, at which to evaluate the likelihood function control list of extra control arguments - see section below object spautolm object from spautolm logical; if 'TRUE', the correlation matrix of the estimated parameters is returned correlation and printed (default=FALSE) adj.se if TRUE, adjust the coefficient standard errors for the number of fitted coefficients Nagelkerke if TRUE, the Nagelkerke pseudo R-squared is reported further arguments passed to or from other methods

## **Details**

This implementation is based on lm.gls and errorsarlm. In particular, the function does not (yet) prevent asymmetric spatial weights being used with "CAR" family models. It appears that both numerical issues (convergence in particular) and uncertainties about the exact spatial weights matrix used make it difficult to reproduce Cressie and Chan's 1989 results, also given in Cressie 1993.

Note that the fitted() function for the output object assumes that the response variable may be reconstructed as the sum of the trend, the signal, and the noise (residuals). Since the values of the response variable are known, their spatial lags are used to calculate signal components (Cressie 1993, p. 564). This differs from other software, including GeoDa, which does not use knowledge of the response variable in making predictions for the fitting data.

#### Value

A list object of class spautolm:

fit a list, with items:

**coefficients** ML coefficient estimates **SSE** ML sum of squared errors

s2 ML residual variance

imat ML coefficient covariance matrix

signal\\_trend non-spatial component of fitted.values
signal\\_stochastic spatial component of fitted.values

**fitted.values** sum of non-spatial and spatial components of fitted.values

residuals difference between observed and fitted values

1ambda ML autoregressive coefficientLL log likelihood for fitted model

LL0 log likelihood for model with lambda=0

call the call used to create this object number of parameters estimated

aliased if not NULL, details of aliased variables

method Jacobian method chosen

family family chosen
zero.policy zero.policy used
weights case weights used

interval the line search interval used

timings processing timings

na.action (possibly) named vector of excluded or omitted observations if non-default na.action

argument used

11prof if not NULL, a list with components lambda and ll of equal length

lambda.se Numerical Hessian-based standard error of lambda fdHess Numerical Hessian-based variance-covariance matrix

X covariates used in model fitting
Y response used in model fitting
weights weights used in model fitting

## **Control arguments**

tol.opt: the desired accuracy of the optimization - passed to optimize() (default=.Machine\$double.eps^(2/3))

**fdHess:** default NULL, then set to (method != "eigen") internally; use fdHess to compute an approximate Hessian using finite differences when using sparse matrix methods; used to make a coefficient covariance matrix when the number of observations is large; may be turned off to save resources if need be

**optimHess:** default FALSE, use fdHess from **nlme**, if TRUE, use optim to calculate Hessian at optimum

optimHessMethod: default "optimHess", may be "nlm" or one of the optim methods

**Imult:** default 2; used for preparing the Cholesky decompositions for updating in the Jacobian function

**super:** if NULL (default), set to FALSE to use a simplicial decomposition for the sparse Cholesky decomposition and method "Matrix\_J", set to as.logical(NA) for method "Matrix", if TRUE, use a supernodal decomposition

cheb\_q: default 5; highest power of the approximating polynomial for the Chebyshev approximation

**MC\_p:** default 16; number of random variates

MC\_m: default 30; number of products of random variates matrix and spatial weights matrix

spamPivot: default "MMD", alternative "RCM"

in\_coef default 0.1, coefficient value for initial Cholesky decomposition in "spam\_update"

**type** default "MC", used with method "moments"; alternatives "mult" and "moments", for use if trs is missing, trW

correct default TRUE, used with method "moments" to compute the Smirnov/Anselin correction term

**trunc** default TRUE, used with method "moments" to truncate the Smirnov/Anselin correction term

SE\_method default "LU", may be "MC"

**nrho** default 200, as in SE toolbox; the size of the first stage Indet grid; it may be reduced to for example 40

interpn default 2000, as in SE toolbox; the size of the second stage Indet grid

small\_asy default TRUE; if the method is not "eigen", use asymmetric covariances rather than numerical Hessian ones if n <= small</pre>

small default 1500; threshold number of observations for asymmetric covariances when the method is not "eigen"

**SEIndet** default NULL, may be used to pass a pre-computed SE toolbox style matrix of coefficients and their lndet values to the "SE\_classic" and "SE\_whichMin" methods

**LU\_order** default FALSE; used in "LU\_prepermutate", note warnings given for 1u method **pre\_eig** default NULL; may be used to pass a pre-computed vector of eigenvalues

#### Note

The standard errors given in Waller and Gotway (2004) are adjusted for the numbers of parameters estimated, and may be reproduced by using the additional argument adj.se=TRUE in the summary method. In addition, the function returns fitted values and residuals as given by Cressie (1993) p. 564.

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion; Ord, J. K. 1975 Estimation methods for models of spatial interaction, Journal of the American Statistical Association, 70, 120-126; Waller, L. A., Gotway, C. A. 2004 Applied spatial statistics for public health, Wiley, Hoboken, NJ, 325-380; Cressie, N. A. C. 1993 Statistics for spatial data, Wiley, New York, 548-568; Ripley, B. D. 1981 Spatial statistics, Wiley, New York, 88-95; LeSage J and RK Pace (2009) Introduction to Spatial Econometrics. CRC Press, Boca Raton.

#### See Also

```
optimize, errorsarlm, do_ldet
```

```
example(NY data)
lm0 <- lm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata)</pre>
summary(1m0)
lmOw <- lm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata, weights=POP8)</pre>
summary(lm0w)
esar0 <- errorsarlm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY)
summary(esar0)
system.time(esar1f <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,</pre>
data=nydata, listw=listw_NY, family="SAR", method="eigen", verbose=TRUE))
summary(esar1f)
system.time(esar1M <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,
data=nydata, listw=listw_NY, family="SAR", method="Matrix", verbose=TRUE))
summary(esar1M)
## Not run:
system.time(esar1M <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,
data=nydata, listw=listw_NY, family="SAR", method="Matrix", verbose=TRUE,
control=list(super=TRUE)))
summary(esar1M)
system.time(esar1s <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,</pre>
data=nydata, listw=listw_NY, family="SAR", method="spam", verbose=TRUE))
summary(esar1s)
esar1wf <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, weights=POP8, family="SAR", method="eigen")
summary(esar1wf)
system.time(esar1wM <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,</pre>
data=nydata, listw=listw_NY, weights=POP8, family="SAR", method="Matrix"))
summary(esar1wM)
esar1ws <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, weights=POP8, family="SAR", method="spam")
summary(esar1ws)
esar1wlu <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, weights=POP8, family="SAR", method="LU")
summary(esar1wlu)
esar1wch <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, weights=POP8, family="SAR", method="Chebyshev")
summary(esar1wch)
```

```
ecar1f <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, family="CAR", method="eigen")
summary(ecar1f)
system.time(ecar1M <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,</pre>
data=nydata, listw=listw_NY, family="CAR", method="Matrix"))
summary(ecar1M)
ecar1s <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, family="CAR", method="spam")
summary(ecar1s)
ecar1wf <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, weights=nydata$POP8, family="CAR", method="eigen")
summary(ecar1wf)
system.time(ecar1wM <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME,</pre>
data=nydata, listw=listw_NY, weights=POP8, family="CAR", method="Matrix"))
summary(ecar1wM)
ecar1ws <- spautolm(Z ~ PEXPOSURE + PCTAGE65P + PCTOWNHOME, data=nydata,
listw=listw_NY, weights=POP8, family="CAR", method="spam")
summary(ecar1ws)
example(nc.sids)
ft.SID74 <- sqrt(1000)*(sqrt(nc.sids$SID74/nc.sids$BIR74) +
sqrt((nc.sids$SID74+1)/nc.sids$BIR74))
lm_nc <- lm(ft.SID74 ~ 1)
sids.nhbr30 <- dnearneigh(cbind(nc.sids$east, nc.sids$north), 0, 30, row.names=row.names(nc.sids))
sids.nhbr30.dist <- nbdists(sids.nhbr30, cbind(nc.sids$east, nc.sids$north))</pre>
sids.nhbr <- listw2sn(nb2listw(sids.nhbr30, glist=sids.nhbr30.dist, style="B", zero.policy=TRUE))
dij <- sids.nhbr[,3]</pre>
n <- nc.sids$BIR74
el1 <- min(dij)/dij
el2 <- sqrt(n[sids.nhbr$to]/n[sids.nhbr$from])</pre>
sids.nhbr$weights <- el1*el2</pre>
sids.nhbr.listw <- sn2listw(sids.nhbr)</pre>
both <- factor(paste(nc.sids$L_id, nc.sids$M_id, sep=":"))</pre>
ft.NWBIR74 <- sqrt(1000)*(sqrt(nc.sids$NWBIR74/nc.sids$BIR74) +
sqrt((nc.sids$NWBIR74+1)/nc.sids$BIR74))
mdata <- data.frame(both, ft.NWBIR74, ft.SID74, BIR74=nc.sids$BIR74)</pre>
outl <- which.max(rstandard(lm_nc))</pre>
as.character(nc.sids$names[outl])
mdata.4 <- mdata[-outl,]</pre>
W <- listw2mat(sids.nhbr.listw)</pre>
W.4 <- W[-outl, -outl]</pre>
sids.nhbr.listw.4 <- mat2listw(W.4)</pre>
esarI <- errorsarlm(ft.SID74 ~ 1, data=mdata, listw=sids.nhbr.listw,</pre>
zero.policy=TRUE)
summary(esarI)
esarIa <- spautolm(ft.SID74 ~ 1, data=mdata, listw=sids.nhbr.listw,
family="SAR")
summary(esarIa)
esarIV <- errorsarlm(ft.SID74 ~ ft.NWBIR74, data=mdata, listw=sids.nhbr.listw,
zero.policy=TRUE)
summary(esarIV)
esarIVa <- spautolm(ft.SID74 ~ ft.NWBIR74, data=mdata, listw=sids.nhbr.listw,
family="SAR")
summary(esarIVa)
```

spdep 189

```
esarIaw <- spautolm(ft.SID74 ~ 1, data=mdata, listw=sids.nhbr.listw,
weights=BIR74, family="SAR")
summary(esarIaw)
esarIIaw <- spautolm(ft.SID74 ~ both - 1, data=mdata, listw=sids.nhbr.listw,
weights=BIR74, family="SAR")
summary(esarIIaw)
esarIVaw <- spautolm(ft.SID74 ~ ft.NWBIR74, data=mdata,
listw=sids.nhbr.listw, weights=BIR74, family="SAR")
summary(esarIVaw)
ecarIaw <- spautolm(ft.SID74 ~ 1, data=mdata.4, listw=sids.nhbr.listw.4,
weights=BIR74, family="CAR")
summary(ecarIaw)
ecarIIaw <- spautolm(ft.SID74 ~ both - 1, data=mdata.4,
listw=sids.nhbr.listw.4, weights=BIR74, family="CAR")
summary(ecarIIaw)
ecarIVaw <- spautolm(ft.SID74 ~ ft.NWBIR74, data=mdata.4,
listw=sids.nhbr.listw.4, weights=BIR74, family="CAR")
summary(ecarIVaw)
nc.sids$fitIV <- append(fitted.values(ecarIVaw), NA, outl-1)</pre>
spplot(nc.sids, c("fitIV"), cuts=12) # Cressie 1993, p. 565
data(oldcol)
COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
nb2listw(COL.nb, style="W"))
summary(COL.errW.eig)
COL.errW.sar <- spautolm(CRIME ~ INC + HOVAL, data=COL.OLD,
nb2listw(COL.nb, style="W"))
summary(COL.errW.sar)
data(boston)
gp1 <- spautolm(log(CMEDV) \sim CRIM + ZN + INDUS + CHAS + I(NOX^2)
+ I(RM^2) + AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
data=boston.c, nb2listw(boston.soi), family="SMA")
summary(gp1)
## End(Not run)
```

spdep

Return package version number

## **Description**

The function retreives package version and build information

#### Usage

```
spdep(build = FALSE)
```

#### **Arguments**

build

if TRUE, also returns build information

spweights.constants

## Value

a character vector with one or two elements

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no >

spweights.constants

Provides constants for spatial weights matrices

## **Description**

The function calculates the constants needed for tests of spatial autocorrelation for general weights matrices represented as listw objects. Note: from spdep 0.3-32, the values of S1 and S2 are returned correctly for both underlying symmetric and asymmetric neighbour lists, before 0.3-32, S1 and S2 were wrong for listw objects based on asymmetric neighbour lists, such as k-nearest neighbours (thanks to Luc Anselin for finding the bug).

## Usage

```
spweights.constants(listw, zero.policy=NULL, adjust.n=TRUE)
Szero(listw)
```

## **Arguments**

listw	a listw object from for example nb2listw
zero.policy	default NULL, use global option value; if TRUE ignore zones without neighbours, if FALSE fail when encountered
adjust.n	default TRUE, if FALSE the number of observations is not adjusted for no- neighbour observations, if TRUE, the number of observations is adjusted

## Value

n	number of zones
n1	n - 1
n2	n - 2
n3	n - 3
nn	n * n
S0	global sum of weights
S1	S1 sum of weights
S2	S2 sum of weights

## Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

ssw 191

#### References

Haining, R. 1990 Spatial data analysis in the social and environmental sciences, Cambridge University Press, p. 233; Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion, p. 19, 21.

#### See Also

```
nb2listw
```

#### **Examples**

```
data(oldcol)
B <- spweights.constants(nb2listw(COL.nb, style="B"))
W <- spweights.constants(nb2listw(COL.nb, style="W"))
C <- spweights.constants(nb2listw(COL.nb, style="C"))
S <- spweights.constants(nb2listw(COL.nb, style="S"))
U <- spweights.constants(nb2listw(COL.nb, style="U"))
print(data.frame(rbind(unlist(B), unlist(W), unlist(C), unlist(S), unlist(U)),
    row.names=c("B", "W", "C", "S", "U")))</pre>
```

SSW

Compute the sum of dissimilarity

## Description

This function computes the sum of dissimilarity between each observation and the mean (scalar of vector) of the observations.

#### Usage

#### **Arguments**

data A matrix with observations in the nodes.

id Node index to compute the cost

method Character or function to declare distance method. If method is character, method

must be "mahalanobis" or "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowisk". If method is one of "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowisk", see dist for details, because this function as used to compute the distance. If method="mahalanobis", the mahalanobis distance is computed between neighbour areas. If method is a function,

this function is used to compute the distance.

p The power of the Minkowski distance.

cov The covariance matrix used to compute the mahalanobis distance.

inverted logical. If 'TRUE', 'cov' is supposed to contain the inverse of the covariance

matrix.

192 stsls

## Value

A numeric, the sum of dissimilarity between the observations id of data and the mean (scalar of vector) of this observations.

## Author(s)

Elias T. Krainski and Renato M. Assuncao

## See Also

See Also as nbcost

## **Examples**

```
data(USArrests)
n <- nrow(USArrests)
ssw(USArrests, 1:n)
ssw(USArrests, 1:(n/2))
ssw(USArrests, (n/2+1):n)
ssw(USArrests, 1:(n/2)) + ssw(USArrests, (n/2+1):n)</pre>
```

stsls

Generalized spatial two stage least squares

## **Description**

The function fits a spatial lag model by two stage least squares, with the option of adjusting the results for heteroskedasticity.

## Usage

```
stsls(formula, data = list(), listw, zero.policy = NULL,
na.action = na.fail, robust = FALSE, HC=NULL, legacy=FALSE, W2X = TRUE)
```

## **Arguments**

formula	a symbolic description of the model to be fit. The details of model specification are given for $lm()$
data	an optional data frame containing the variables in the model. By default the variables are taken from the environment which the function is called.
listw	a listw object created for example by nb2listw
zero.policy	default NULL, use global option value; if TRUE assign zero to the lagged value of zones without neighbours, if FALSE (default) assign NA - causing lagsarlm() to terminate with an error

stsls 193

na.action a function (default na.fail), can also be na.omit or na.exclude with conse-

quences for residuals and fitted values - in these cases the weights list will be subsetted to remove NAs in the data. It may be necessary to set zero.policy to TRUE because this subsetting may create no-neighbour observations. Note that only weights lists created without using the glist argument to nb2listw may be

subsetted.

robust default FALSE, if TRUE, apply a heteroskedasticity correction to the coeffi-

cients covariances

HC default NULL, if robust is TRUE, assigned "HC0", may take values "HC0" or

"HC1" for White estimates or MacKinnon-White estimates respectively

legacy the argument chooses between two implementations of the robustness correc-

tion: default FALSE - use the estimate of Omega only in the White consistent estimator of the variance-covariance matrix, if TRUE, use the original implementation which runs a GLS using the estimate of Omega, and yields different

coefficient estimates as well - see example below

W2X default TRUE, if FALSE only WX are used as instruments in the spatial two

stage least squares; until release 0.4-60, only WX were used - see example below

#### **Details**

The fitting implementation fits a spatial lag model:

$$y = \rho W y + X \beta + \varepsilon$$

by using spatially lagged X variables as instruments for the spatially lagged dependent variable.

#### Value

an object of class "stsls" containing:

coefficients coefficient estimates

var coefficient covariance matrix

sse sum of squared errors

residuals model residuals df degrees of freedom

## Author(s)

Luc Anselin, Gianfranco Piras and Roger Bivand

## References

Kelejian, H.H. and I.R. Prucha (1998). A generalized spatial two stage least squares procedure for estimating a spatial autoregressive model with autoregressive disturbances. *Journal of Real Estate Finance and Economics* 17, 99-121.

194 subset.listw

## See Also

lagsarlm

#### **Examples**

```
data(oldcol)
COL.lag.eig <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD, nb2listw(COL.nb))
summary(COL.lag.eig, correlation=TRUE)
COL.lag.stsls <- stsls(CRIME ~ INC + HOVAL, data=COL.OLD, nb2listw(COL.nb))
summary(COL.lag.stsls, correlation=TRUE)
COL.lag.stslsW <- stsls(CRIME ~ INC + HOVAL, data=COL.OLD, nb2listw(COL.nb), W2X=FALSE)
summary(COL.lag.stslsW, correlation=TRUE)
COL.lag.stslsR <- stsls(CRIME ~ INC + HOVAL, data=COL.OLD, nb2listw(COL.nb),
robust=TRUE, W2X=FALSE)
summary(COL.lag.stslsR, correlation=TRUE)
COL.lag.stslsRl <- stsls(CRIME ~ INC + HOVAL, data=COL.OLD, nb2listw(COL.nb),
robust=TRUE, legacy=TRUE, W2X=FALSE)
summary(COL.lag.stslsRl, correlation=TRUE)
data(boston)
gp2a \leftarrow stsls(log(CMEDV) \sim CRIM + ZN + INDUS + CHAS + I(NOX^2) + I(RM^2) +
 AGE + log(DIS) + log(RAD) + TAX + PTRATIO + B + log(LSTAT),
data=boston.c, nb2listw(boston.soi))
summary(gp2a)
```

subset.listw

Subset a spatial weights list

## **Description**

The function subsets a spatial weights list, retaining objects for which the subset argument vector is TRUE. At present it will only subset non-general weights lists (that is those created by nb2listw with glist=NULL).

## Usage

```
## S3 method for class 'listw'
subset(x, subset, zero.policy = NULL, ...)
```

## **Arguments**

x	an object of class listw
subset	logical expression
zero.policy	default NULL, use global option value; if FALSE stop with error for any empty neighbour sets, if TRUE permit the weights list to be formed with zero-length weights vectors - passed through to nb21istw
	generic function pass-through

subset.nb 195

## Value

The function returns an object of class listw with component style the same as the input object, component neighbours a list of integer vectors containing neighbour region number ids (compacted to run from 1:number of regions in subset), and component weights as the weights computed for neighbours using style.

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

```
nb2listw, subset.nb
```

## **Examples**

```
example(columbus)
to.be.dropped <- c(31, 34, 36, 39, 42, 46)
pre <- nb2listw(col.gal.nb)
print(pre)
post <- subset(pre, !(1:length(col.gal.nb) %in% to.be.dropped))
print(post)</pre>
```

subset.nb

Subset a neighbours list

## Description

The function subsets a neighbors list, retaining objects for which the subset argument vector is TRUE.

## Usage

```
## S3 method for class 'nb'
subset(x, subset, ...)
```

## **Arguments**

```
x an object of class nbsubset logical expression... generic function pass-through
```

#### Value

The function returns an object of class nb with a list of integer vectors containing neighbour region number ids (compacted to run from 1:number of regions in subset).

196 summary.nb

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

#### See Also

```
nb2listw
```

## **Examples**

```
example(columbus)
coords <- coordinates(columbus)
plot(col.gal.nb, coords)
to.be.dropped <- c(31, 34, 36, 39, 42, 46)
text(coords[to.be.dropped,1], coords[to.be.dropped,2], labels=to.be.dropped,
    pos=2, offset=0.3)
sub.col.gal.nb <- subset(col.gal.nb,
    !(1:length(col.gal.nb) %in% to.be.dropped))
plot(sub.col.gal.nb, coords[-to.be.dropped,], col="red", add=TRUE)
which(!(attr(col.gal.nb, "region.id"))%in%
    attr(sub.col.gal.nb, "region.id"))))</pre>
```

summary.nb

Print and summary function for neighbours and weights lists

## **Description**

The function prints summary measures for links in a neighbours list. If a matrix of coordinates is given as well, summary descriptive measures for the link lengths are also printed. Print and summary functions are also available for "listw" weights list objects.

#### Usage

## **Arguments**

```
object an object of class nb
coords matrix of region point coordinates or a SpatialPoints object
```

summary.sarlm 197

longlat	TRUE if point coordinates are longitude-latitude decimal degrees, in which case distances are measured in kilometers; if coords is a SpatialPoints object, the value is taken from the object itself
	additional arguments affecting the output produced
x	an object of class nb
zero.policy	default NULL, use global option value; if FALSE stop with error for any empty neighbour sets
scale	passed through to stem() for control of plot length

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no >

## See Also

```
plot.nb
```

## **Examples**

```
example(columbus)
coords <- coordinates(columbus)
col.gal.nb
summary(col.gal.nb, coords)
col.listw <- nb2listw(col.gal.nb, style="W")
col.listw
summary(col.listw)</pre>
```

summary.sarlm

summary method for class sarlm

## **Description**

Methods used for presenting the results of estimating spatial SAR models.

## Usage

```
## S3 method for class 'sarlm'
summary(object, correlation = FALSE, Nagelkerke = FALSE, Hausman=FALSE, adj.se=FALSE, ...)
## S3 method for class 'sarlm'
print(x, ...)
## S3 method for class 'summary.sarlm'
print(x, digits = max(5, .Options$digits - 3),
signif.stars = FALSE, ...)
```

198 summary.sarlm

## Arguments

object sarlm object from lagsarlm or errorsarlm correlation logical; if 'TRUE', the correlation matrix of the estimated parameters including sigma is returned and printed (default=FALSE) Nagelkerke if TRUE, the Nagelkerke pseudo R-squared is reported Hausman if TRUE, the results of the Hausman test for error models are reported adj.se if TRUE, adjust the coefficient standard errors for the number of fitted coefficients sarlm object from lagsarlm or errorsarlm in print.sarlm, summary object Х from summary.sarlm for print.summary.sarlm the number of significant digits to use when printing digits logical. If TRUE, "significance stars" are printed for each coefficient. signif.stars

further arguments passed to or from other methods

#### Value

. . .

The summary function summary.sarlm returns the sarlm object augmented with a coefficient matrix with probability values for coefficient asymptotic standard errors for type="error" and for type="lag" or "mixed" when object\\$ase=TRUE, or a coefficient matrix with probability values for likelihood ratio tests between the model as reported and models with independent variables dropped in turn.

#### Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## References

Cliff, A. D., Ord, J. K. 1981 Spatial processes, Pion; Ord, J. K. 1975 Estimation methods for models of spatial interaction, Journal of the American Statistical Association, 70, 120-126; Anselin, L. 1988 Spatial econometrics: methods and models. (Dordrecht: Kluwer); Anselin, L. 1995 SpaceStat, a software program for the analysis of spatial data, version 1.80. Regional Research Institute, West Virginia University, Morgantown, WV (www.spacestat.com); Anselin L, Bera AK (1998) Spatial dependence in linear regression models with an introduction to spatial econometrics. In: Ullah A, Giles DEA (eds) Handbook of applied economic statistics. Marcel Dekker, New York, pp. 237-289; Nagelkerke NJD (1991) A note on a general definition of the coefficient of determination. Biometrika 78: 691-692.

#### See Also

errorsarlm, lagsarlm, summary.lm

tolerance.nb 199

## **Examples**

```
data(oldcol)
COL.mix.eig <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    nb2listw(COL.nb), type="mixed", method="eigen")
summary(COL.mix.eig, correlation=TRUE, Nagelkerke=TRUE)
COL.mix.M <- lagsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    nb2listw(COL.nb), type="mixed", method="Matrix")
summary(COL.mix.M, correlation=TRUE, Nagelkerke=TRUE)
COL.errW.eig <- errorsarlm(CRIME ~ INC + HOVAL, data=COL.OLD,
    nb2listw(COL.nb, style="W"), method="eigen")
summary(COL.errW.eig, correlation=TRUE, Nagelkerke=TRUE, Hausman=TRUE)</pre>
```

tolerance.nb

Function to construct edges based on a tolerance angle and a maximum distance

## **Description**

This function creates an object of class nb (defined in the library spdep) containing a connexion diagram. The edges between sites are based on a tolerance angle and a maximum distance. The angle is directional; its direction is always from the bottow to the top of the screen.

## Usage

```
tolerance.nb(coords, unit.angle = "degrees", max.dist, tolerance, rot.angle,
plot.sites=FALSE)
```

## Arguments

coords	A matrix or a data frame containing the X and Y coordinates of the study sites.
unit.angle	Character. The measurement units in which angles are defined: either "degrees" (default) or "radians".
max.dist	Numeric. The maximum distance of an edge linking two sites together.
tolerance	Numeric. The tolerance angle in which a site can influence another site. The angle is measured vertically and from bottom to top of the pictures after rotation of the points.
rot.angle	Numeric, optional. An angle at which a set of coordinates should be rotated before creating the connexion diagram. The set of coordinates is rotated counterclockwise. Negative values will produce a clockwise rotation.
plot.sites	Logical (TRUE, FALSE) determining if the site should be plotted in a graphic window. This graph allows one to make sure the points are rotated in a correct direction.

## **Details**

Even though this function creates a connexion diagram based on a tolerance angle going from the bottom to the top of the screen, the resulting object is symmetric, meaning that a site influences another and vice versa. The final object does not represent a directional connexion network.

200 tolerance.nb

#### Value

The function returns an object of class nb with a list of integer vectors corresponding to neighbour region numbers.

## Warning

This function was not design to handle a large number of rows in coords. To use this function for a set of coordinates with more than 1500 entries is memory intensive.

## Author(s)

F. Guillaume Blanchet

## See Also

```
dnearneigh, cell2nb, graph2nb, tri2nb, knn2nb
```

```
set.seed(1)
ex.data<-cbind(runif(50),rexp(50))</pre>
### Construct object of class nb with a tolerance angle of 30 degrees
### and a maximum distance of 2 m.
nb.ex<-tolerance.nb(ex.data, unit.angle = "degrees", max.dist=1,</pre>
tolerance = 30)
### Construct object of class nb with a tolerance angle of 30 degrees
\#\#\# and a maximum distance of 2 m. The coordinates are rotated at an angle
### of 45 degrees counterclockwise.
nb.ex2<-tolerance.nb(ex.data, unit.angle = "degrees", max.dist=1,</pre>
tolerance = 30, rot.angle = 45)
### Construct object of class nb with a tolerance angle of pi/8 radians
### and a maximum distance of 1.5 m. The coordinates are rotated at
### an angle of pi/4 radians clockwise.
nb.ex3<-tolerance.nb(ex.data, unit.angle = "radians", max.dist=1.5,</pre>
tolerance = pi/8, rot.angle = -pi*2/3)
par(mfrow=c(1,3))
plot(nb.ex,ex.data,asp=1)
plot(nb.ex2,ex.data,asp=1)
plot(nb.ex3,ex.data,asp=1)
```

*tri2nb* 201

tri2nb

Neighbours list from tri object

## **Description**

The function uses the deldir package to convert a matrix of two-dimensional coordinates into a neighbours list of class nb with a list of integer vectors containing neighbour region number ids.

## Usage

```
tri2nb(coords, row.names = NULL)
```

## **Arguments**

coords matrix of point coordinates with two columns

row.names character vector of region ids to be added to the neighbours list as attribute

region.id, default seq(1, nrow(x))

#### **Details**

If coordinates are duplicated, this function cannot be used. If the coordinates are from a grid, then they need to be ordered such that the first three are not collinear, so that the first triangle can be constructed. This can be achieved by randomising the order of the coordinates (possibly several times), and then re-ordering the order of the data to match the new order of the neighbour list - if this fix is used, remember to re-order the row.names argument as well as the coordinates! Please also note that triangulation of grid points will give arbitrary diagonal neighbours, which may not be a sensible outcome, and dnearneigh() may serve better where tri2nb() cannot be used.

## Value

The function returns an object of class nb with a list of integer vectors containing neighbour region number ids.

#### Author(s)

```
Roger Bivand < Roger . Bivand @nhh . no >
```

#### See Also

```
knn2nb, dnearneigh, cell2nb
```

```
example(columbus)
coords <- coordinates(columbus)
ind <- sapply(slot(columbus, "polygons"), function(x) slot(x, "ID"))
col.tri.nb <- tri2nb(coords, row.names=ind)
plot(columbus, border="grey")</pre>
```

202 trW

```
plot(col.tri.nb, coords, add=TRUE)
title(main="Raw triangulation links")
x <- seq(0,1,0.1)
y <- seq(0,2,0.2)
xy <- expand.grid(x, y)
try(xy.nb <- tri2nb(xy))
seed <- 1234
xid <- sample(1:nrow(xy))
xy.nb <- tri2nb(xy[xid,])
plot(xy.nb, xy[xid,])</pre>
```

trW

Spatial weights matrix powers traces

## **Description**

The function is used to prepare a vector of traces of powers of a spatial weights matrix

## Usage

```
trW(W=NULL, m = 30, p = 16, type = "mult", listw=NULL, momentsSymmetry=TRUE)
mom_calc(lw, m)
mom_calc_int2(is, m, nb, weights, Card)
```

## Arguments

W	A spatial weights matrix in CsparseMatrix form	
m	The number of powers; must be an even number for 'type'="moments" (default changed from 100 to 30 (2010-11-17))	
р	The number of samples used in Monte Carlo simulation of the traces if type is MC (default changed from 50 to 16 (2010-11-17))	
type	Either "mult" (default) for powering a sparse matrix (with moderate or larger N, the matrix becomes dense, and may lead to swapping), or "MC" for Monte Carlo simulation of the traces (the first two simulated traces are replaced by their analytical equivalents), or "moments" to use the looping space saving algorithm proposed by Smirnov and Anselin (2009) - for "moments", W must be symmetric, for row-standardised weights through a similarity transformation	
listw, lw	a listw object, which should either be fully symmetric, or be constructed as similar to symmetric from intrinsically symmetric neighbours using similar.listw, used with 'type'="moments"	
momentsSymmetry		
	default TRUE; assert Smirnov/Anselin symmetry assumption	
is	(used internally only in mom_calc_int2 for 'type'="moments" on a cluster)	
nb	(used internally only in mom_calc_int2 for 'type'="moments" on a cluster)	
weights	(used internally only in mom_calc_int2 for 'type'="moments" on a cluster)	
Card	(used internally only in mom_calc_int2 for 'type'="moments" on a cluster)	

trW 203

#### Value

A numeric vector of m traces, with "timings" and "type" attributes; the 'type'="MC" also returns the standard deviation of the p-vector V divided by the square root of p as a measure of spread for the trace estimates.

#### Note

mom\_calc and mom\_calc\_int2 are for internal use only

## Author(s)

Roger Bivand < Roger . Bivand@nhh . no>

## References

LeSage J and RK Pace (2009) *Introduction to Spatial Econometrics*. CRC Press, Boca Raton, pp. 96–105; Smirnov O and L Anselin (2009) An O(N) parallel method of computing the Log-Jacobian of the variable transformation for models with spatial interaction on a lattice. *Computational Statistics and Data Analysis* 53 (2009) 2983–2984.

#### See Also

```
as_dgRMatrix_listw, nb2listw
```

```
example(columbus)
listw <- nb2listw(col.gal.nb)</pre>
W <- as(as_dgRMatrix_listw(listw), "CsparseMatrix")</pre>
system.time(trMat <- trW(W, type="mult"))</pre>
str(trMat)
set.seed(1100)
system.time(trMC <- trW(W, type="MC"))</pre>
str(trMC)
plot(trMat, trMC)
abline(a=0, b=1)
for(i in 3:length(trMC)) {
 segments(trMat[i], trMC[i]-2*attr(trMC, "sd")[i], trMat[i],
  trMC[i]+2*attr(trMC, "sd")[i])
listwS <- similar.listw(listw)</pre>
W <- as(as_dgRMatrix_listw(listwS), "CsparseMatrix"), "symmetricMatrix")</pre>
system.time(trmom <- trW(W, m=24, type="moments"))</pre>
str(trmom)
all.equal(trMat[1:24], trmom, check.attributes=FALSE)
system.time(trMat <- trW(W, m=24, type="mult"))</pre>
str(trMat)
all.equal(trMat, trmom, check.attributes=FALSE)
set.seed(1)
system.time(trMC <- trW(W, m=24, type="MC"))</pre>
str(trMC)
```

204 used.cars

used.cars

US 1960 used car prices

## **Description**

The used.cars data frame has 48 rows and 2 columns. The data set includes a neighbours list for the 48 states excluding DC from poly2nb().

## Usage

```
data(used.cars)
```

#### **Format**

This data frame contains the following columns:

```
tax.charges taxes and delivery charges for 1955-9 new cars price.1960 1960 used car prices by state
```

#### Source

Hanna, F. A. 1966 Effects of regional differences in taxes and transport charges on automobile consumption, in Ostry, S., Rhymes, J. K. (eds) Papers on regional statistical studies, Toronto: Toronto University Press, pp. 199-223.

## References

Hepple, L. W. 1976 A maximum likelihood model for econometric estimation with spatial series, in Masser, I (ed) Theory and practice in regional science, London: Pion, pp. 90-104.

```
data(used.cars)
moran.test(used.cars$price.1960, nb2listw(usa48.nb))
moran.plot(used.cars$price.1960, nb2listw(usa48.nb),
  labels=rownames(used.cars))
uc.lm <- lm(price.1960 ~ tax.charges, data=used.cars)
summary(uc.lm)
lm.morantest(uc.lm, nb2listw(usa48.nb))
lm.morantest.sad(uc.lm, nb2listw(usa48.nb))
lm.LMtests(uc.lm, nb2listw(usa48.nb))
uc.err <- errorsarlm(price.1960 ~ tax.charges, data=used.cars,</pre>
  nb2listw(usa48.nb), tol.solve=1.0e-13, control=list(tol.opt=.Machine$double.eps^0.3))
summary(uc.err)
uc.lag <- lagsarlm(price.1960 ~ tax.charges, data=used.cars,</pre>
  nb2listw(usa48.nb), tol.solve=1.0e-13, control=list(tol.opt=.Machine$double.eps^0.3))
summary(uc.lag)
uc.lag1 <- lagsarlm(price.1960 ~ 1, data=used.cars,
  nb2listw(usa48.nb), tol.solve=1.0e-13, control=list(tol.opt=.Machine$double.eps^0.3))
```

wheat 205

```
summary(uc.lag1)
uc.err1 <- errorsarlm(price.1960 ~ 1, data=used.cars,
   nb2listw(usa48.nb), tol.solve=1.0e-13, control=list(tol.opt=.Machine$double.eps^0.3))
summary(uc.err1)</pre>
```

wheat

Mercer and Hall wheat yield data

## Description

Mercer and Hall wheat yield data, based on version in Cressie (1993), p. 455.

## Usage

data(wheat)

#### **Format**

The format of the object generated by running data(wheat) is a three column data frame made available by Hongfei Li. The example section shows how to convert this to the object used in demonstrating the aple function, and is a: Formal class 'SpatialPolygonsDataFrame' [package "sp"] with 5 slots; the data slot is a data frame with 500 observations on the following 6 variables.

lat local coordinates northings ordered north to south

yield Mercer and Hall wheat yield data

r rows south to north; levels in distance units of plot centres

c columns west to east; levels in distance units of plot centres

lon local coordinates eastings

1at1 local coordinates northings ordered south to north

## Source

Cressie, N. A. C. (1993) Statistics for Spatial Data. Wiley, New York, p. 455.

#### References

Mercer, W. B. and Hall, A. D. (1911) The experimental error of field trials. Journal of Agricultural Science 4, 107-132.

206 write.nb.gal

## **Examples**

```
## Not run:
data(wheat)
wheat$lat1 <- 69 - wheat$lat</pre>
wheat$r <- factor(wheat$lat1)</pre>
wheat$c <- factor(wheat$lon)</pre>
wheat_sp <- wheat</pre>
coordinates(wheat_sp) <- c("lon", "lat1")</pre>
wheat_spg <- wheat_sp</pre>
gridded(wheat_spg) <- TRUE</pre>
wheat_spl <- as(wheat_spg, "SpatialPolygons")</pre>
df <- as(wheat_spg, "data.frame")</pre>
row.names(df) <- sapply(slot(wheat_spl, "polygons"),</pre>
function(x) slot(x, "ID"))
wheat <- SpatialPolygonsDataFrame(wheat_spl, data=df)</pre>
## End(Not run)
require(maptools)
wheat <- readShapeSpatial(system.file("etc/shapes/wheat.shp",</pre>
 package="spdep")[1])
```

write.nb.gal

Write a neighbours list as a GAL lattice file

## Description

Write a neighbours list as a GAL lattice file, may also use newer GeoDa header format

## Usage

```
write.nb.gal(nb, file, oldstyle=TRUE, shpfile=NULL, ind=NULL)
```

## **Arguments**

nb	an object of class nb with a list of integer vectors containing neighbour region number ids.
file	name of file with GAL lattice data
oldstyle	if TRUE, first line of file contains only number of spatial units, if FALSE, uses newer GeoDa style
shpfile	Shapefile name taken from GAL file for this dataset
ind	region id indicator variable name

## Author(s)

Roger Bivand < Roger . Bivand@nhh.no>

write.nb.gal 207

## See Also

```
read.gal
```

```
example(columbus)
GALfile <- tempfile("GAL")
write.nb.gal(col.gal.nb, GALfile)
col.queen <- read.gal(GALfile)
summary(diffnb(col.queen, col.gal.nb))</pre>
```

# **Index**

*Topic <b>cluster</b>	anava carlm 7
nbcosts, 143	anova.sarlm, $7$ aple, $8$
plot.skater, 153	aple.mc, 9
prunecost, 160	aple.plot, 11
prunemst, 161	as_dgRMatrix_listw, 12
•	
skater, 174	autocov_dist, 14
ssw, 191	bptest.sarlm, 20
*Topic datasets	card, 22
afcon, 4	cell2nb, 23
auckland, 13	choynowski, 24
baltimore, 16	diffnb, 27
boston, 18	dnearneigh, 28
columbus, 25	do_ldet, 29
eire, 46	droplinks, 36
elect80,48	EBest, 37
getisord, 59	EBImoran.mc, 39
hopkins, 71	EBlocal, 41
house, 71	edit.nb,42
huddersfield, 73	eigenw,43
nc.sids, 146	errorsarlm, 49
NY_data, 147	geary, 54
oldcol, 148	geary.mc, 56
used.cars, 204	geary.test,57
wheat, 205	globalG.test, $60$
*Topic <b>data</b>	GMerrorsar, 61
bhicv, 18	Graph Components, 65
*Topic <b>graphs</b>	graphneigh,66
mstree, 132	gstsls, 68
prunecost, 160	impacts, 74
*Topic <b>hplot</b>	include.self,78
plot.mst, 151	invIrM, 79
plot.skater, 153	is.symmetric.nb, 81
*Topic <b>manip</b>	joincount.mc, 82
Rotation, 166	joincount.multi,84
*Topic multivariate	joincount.test,86
ssw, 191	knearneigh, 87
*Topic <b>spatial</b>	knn2nb, 89
aggregate.nb, 5	lag.listw, 90
airdist, 6	lagmess, 91
	-

lagsarlm, 94	subset.nb, 195
listw2sn, 99	summary.nb, 196
lm.LMtests, 101	summary.sarlm, 197
1m.morantest, 103	tolerance.nb, 199
<pre>lm.morantest.exact, 105</pre>	tri2nb, 201
lm.morantest.sad, 106	trW, 202
localG, 109	write.nb.gal,206
localmoran, 110	*Topic <b>tree</b>
localmoran.exact, 113	plot.mst, 151
localmoran.sad, 115	prunemst, 161
LR.sarlm, 118	skater, 174
mat2listw, 120	
MCMCsamp, 121	afcon, 4
ME, 123	africa.rook.nb(afcon),4
moran, 125	afxy (afcon), 4
moran.mc, 127	aggregate.nb, 5
moran.plot, 128	AIC, 8
moran.test, 130	airdist, 6
mstree, 132	anova.sarlm, 7, <i>119</i>
nb.set.operations, 134	aple, 8, <i>10</i> , <i>12</i>
nb2blocknb, 135	aple.mc, 9, 9
nb2INLA, 136	aple.plot, $9$ , $11$
nb2lines, 137	as.data.frame.localmoranex
nb2listw, 138	(localmoran.exact), 113
nb2mat, 140	as.data.frame.localmoransad
nb2WB, 142	(localmoran.sad), 115
nbcosts, 143	as.data.frame.sarlm.pred
nbdists, 144	(predict.sarlm), 155
nblag, 145	as.spam.listw(listw2sn),99
p.adjustSP, 150	as_dgRMatrix_listw, 12,203
plot.nb, 151	<pre>as_dsCMatrix_I (as_dgRMatrix_listw), 12</pre>
poly2nb, 154	<pre>as_dsCMatrix_IrW(as_dgRMatrix_listw),</pre>
predict.sarlm, 155	12
probmap, 158	as_dsTMatrix_listw
read.gal, 162	(as_dgRMatrix_listw), 12
read.gwt2nb, 163	auckland, 13
residuals.sarlm, 165	auckpolys (auckland), 13
sacsarlm, 167	autocov_dist, 14
set.spChkOption, 171	1.1.1
similar.listw, 172	baltimore, 16
sp.correlogram, 177	bbs (columbus), 25
sp.mantel.mc, 179	bhicv, 18
SpatialFiltering, 181	boot, 10
·	boston, 18
spautolm, 183	bptest.sarlm, 20
spdep, 189	and 22 22 20 67 90 154 155
spweights.constants, 190	card, 22, 23, 28, 67, 89, 154, 155
stsls, 192	cell2nb, 23, 136, 200, 201
subset.listw, 194	cheb_setup (do_ldet), 29

chkIDs (set.spChkOption), 171	fitted.SFResult(SpatialFiltering), 181
Cholesky, 30, 31, 34	fitted.spautolm(spautolm), 183
choynowski, 24	
coef.gmsar(GMerrorsar), 61	gabrielneigh (graphneigh), 66
coef.lagmess(lagmess), 91	geary, 54, <i>57</i> , <i>59</i>
coef.sarlm(residuals.sarlm), 165	geary.mc, 55, 56, 59, 180
coef.spautolm(spautolm), 183	geary.test, <i>55–57</i> , <i>57</i>
coef.stsls(stsls), 192	<pre>get.ClusterOption(set.spChkOption), 171</pre>
col.gal.nb (columbus), 25	<pre>get.rlecuyerSeedOption</pre>
COL.nb (oldcol), 148	(set.spChkOption), 171
COL.OLD (oldcol), 148	<pre>get.spChkOption(set.spChkOption), 171</pre>
columbus, 25	<pre>get.VerboseOption(set.spChkOption), 171</pre>
complement.nb (nb.set.operations), 134	<pre>get.ZeroPolicyOption(set.spChkOption),</pre>
coords (columbus), 25	171
	getisord, 59
deviance.gmsar(GMerrorsar), 61	glm, 124
deviance.lagmess (lagmess), 91	globalG.test, 60
deviance.sarlm(residuals.sarlm), 165	GMargminImage, 70
deviance.spautolm(spautolm), 183	GMargminImage (GMerrorsar), 61
deviance.stsls(stsls), 192	GMerrorsar, $61,70$
df2sn (nb2lines), 137	Graph Components, 65
diffnb, 27	graph2nb, 200
dist, 143, 160, 174, 191	graph2nb (graphneigh), 66
dll (elect80), 48	graphneigh, 66
dnearneigh, 28, 67, 88, 136, 200, 201	griffith_sone (eigenw), 43
do_ldet, 29, 53, 98, 187	gstsls, 68
dput, <i>142</i>	
droplinks, 36	Hausman.test (LR.sarlm), 118
	help, <i>170</i>
e80_queen (elect80), 48	hopkins, 71
EBest, 37, 40, 42, 159	house, 71
EBImoran (EBImoran.mc), 39	HPDinterval, 76
EBImoran.mc, $38$ , $39$	<pre>HPDinterval.lagImpact(impacts), 74</pre>
EBlocal, 38, 41, 159	huddersfield, 73
edit.nb, 42	
eigen, 44, 182	impacts, 74
eigen_pre_setup(do_ldet),29	impacts.sarlm,98
eigen_setup(do_ldet), 29	include.self, 78
eigenw, 43, 98	influence.measures, 129
eire, 46	intersect.nb, <i>134</i>
elect80, 48	<pre>intersect.nb (nb.set.operations), 134</pre>
elect80_lw(elect80),48	invIrM, 79
errorsarlm, 21, 34, 49, 64, 98, 117, 122, 157,	invIrW(invIrM), 79
166, 173, 184, 187, 198	<pre>is.symmetric.glist(is.symmetric.nb), 81</pre>
	is.symmetric.nb, 37, 81
fitted.gmsar (GMerrorsar), 61	
fitted.lagmess (lagmess), 91	Jacobian_W (as_dgRMatrix_listw), 12
fitted.ME_res (ME), 123	joincount.mc, 82, 87, 180
fitted.sarlm(residuals.sarlm), 165	joincount.multi, 84, 87

joincount.test, <i>83</i> , <i>85</i> , <i>86</i>	<pre>mom_calc(trW), 202</pre>
1.1.(7	<pre>mom_calc_int2(trW), 202</pre>
k4 (elect80), 48	<pre>moments_setup (do_ldet), 29</pre>
knearneigh, 28, 67, 87, 89	moran, 40, 125, 128, 131, 178
knn, 88	moran.mc, 40, 126, 127, 131, 180
knn2nb, 67, 88, 89, 136, 200, 201	moran.plot, 128
1 1:	moran.test, <i>126</i> , <i>128</i> , 130
lag.listw, 90	mrc2vi (cell2nb), 23
lagmess, 91	mstree, 132, <i>151</i> , <i>153</i> , <i>161</i> , <i>175</i>
lagsarlm, 21, 34, 53, 76, 93, 94, 122, 157, 166, 173, 194, 198	mvrnorm, 75, 76
listw2lines (nb2lines), 137	n.comp.nb(Graph Components), 65
listw2mat (nb2mat), 140	nb.set.operations, 134
listw2sn, 99	nb2blocknb, 135
listw2star (localmoran.sad), 115	
listw2U, 59, 87, 131, 182	nb2INLA, 136
	nb2lines, 137
listw2U (lm. morantest), 103	nb2listw, 9, 15, 76, 80, 90, 100, 120, 132,
listw2WB (nb2WB), 142	138, 141, 143, 144, 182, 191, 195,
listw_NY (NY_data), 147	196, 203
lm, 53, 98, 102, 104, 182	nb2mat, 120, 140
lm.gls, 184	nb2WB, 142
lm.LMtests, 101, 104	nbcost, <i>192</i>
lm.morantest, 103, 108, 117	nbcost (nbcosts), 143
lm.morantest.exact, 105, 114	nbcosts, 143
lm.morantest.sad, 106, 106, 117	nbdists, <i>143</i> , 144
L0_nb (house), 71	nblag, 145, <i>178</i>
localAple (aple.plot), 11	nblag_cumul (nblag), 145
localG, 60, 61, 109, 112, 150	nc.sids, 146
localmoran, 110, 117, 129, 150	ncCC89.nb (nc.sids), 146
localmoran.exact, 113	ncCR85.nb (nc.sids), 146
localmoran.sad, <i>114</i> , 115	nlminb, 62, 64, 69, 70
locator, 7	nn2, <i>88</i>
logLik.lagmess(lagmess), 91	NY_data, 147
logLik.lm, <i>119</i>	nydata (NY_data), 147
logLik.sarlm(LR.sarlm), 118	
logLik.spautolm(spautolm), 183	old.make.sym.nb(is.symmetric.nb),81
LR.sarlm, 8, 118	oldcol, 148
LR1.sarlm(LR.sarlm), 118	optim, 62, 64, 69, 70, 93
LR1.spautolm(spautolm), 183	optimize, 187
LU_prepermutate_setup (do_ldet), 29	
LU_setup (do_ldet), 29	p.adjust, <i>102</i> , <i>150</i> , <i>178</i>
	p.adjustSP, <i>111</i> , 150
<pre>make.sym.nb(is.symmetric.nb), 81</pre>	paper.nb(afcon),4
mat2listw, 120	plot.Gabriel (graphneigh), 66
Matrix_J_setup (do_ldet), 29	<pre>plot.lagImpact(impacts), 74</pre>
Matrix_setup (do_ldet), 29	plot.listw(plot.nb), 151
<pre>mcdet_setup (do_ldet), 29</pre>	plot.mc.sim(sp.mantel.mc), 179
MCMCsamp, 121	plot.mcmc, 76
ME, 123	plot.mst, 151

plot.nb, 43, 65, 151, 197	read.gwt2nb, 163
plot.relative(graphneigh),66	readShapeLines, 138
plot.skater, 153	relativeneigh (graphneigh), 66
plot.spcor(sp.correlogram), 177	residuals.gmsar(GMerrorsar),61
poly2nb, <i>136</i> , <i>143</i> , 154	residuals.lagmess(lagmess), 91
polys (columbus), 25	residuals.sarlm, 53, 98, 165
powerWeights (invIrM), 79	residuals.spautolm(spautolm), 183
ppois, <i>159</i>	residuals.stsls(stsls), 192
predict.sarlm, 53, 98, 155, 166	rookcell (cell2nb), 23
print.gmsar(GMerrorsar), 61	Rotation, 166
print.jclist(joincount.test), 86	rwmetrop, 121, 122
print.jcmulti(joincount.multi), 84	
print.lagImpact (impacts), 74	sacsarlm, <i>122</i> , 167
print.lagmess (lagmess), 91	SE_classic_setup (do_ldet), 29
print.listw(summary.nb), 196	SE_interp_setup(do_ldet), 29
print.LMtestlist (lm.LMtests), 101	<pre>SE_whichMin_setup(do_ldet), 29</pre>
print.localmoranex (localmoran.exact),	<pre>set.ClusterOption(set.spChkOption), 171</pre>
113	set.rlecuyerSeedOption
<pre>print.localmoransad(localmoran.sad),</pre>	(set.spChkOption), 171
115	set.spChkOption, 171
print.ME_res (ME), 123	<pre>set.VerboseOption(set.spChkOption), 171</pre>
print.moranex (1m.morantest.exact), 105	<pre>set.ZeroPolicyOption(set.spChkOption),</pre>
print.moransad (lm.morantest.sad), 106	171
print.nb (summary.nb), 196	setdiff.nb, <i>134</i>
print.sarlm(summary.sarlm), 197	setdiff.nb(nb.set.operations), 134
print.sarlm.pred(predict.sarlm), 155	sidscents (nc.sids), 146
print.SFResult (SpatialFiltering), 181	sidspolys (nc.sids), 146
print.spautolm(spautolm), 183	similar.listw, <i>53</i> , 172, <i>202</i>
print.spcor(sp.correlogram), 177	skater, <i>151</i> , <i>153</i> , 174
print.stsls(stsls), 192	sn2listw, <i>138</i>
print.summary.gmsar(GMerrorsar), 61	sn2listw(listw2sn), 99
print.summary.lagImpact(impacts), 74	soi.graph(graphneigh),66
print.summary.lagmess (lagmess), 91	sp.correlogram, 177
print.summary.localmoransad	sp.mantel.mc, <i>55</i> , 179
(localmoran.sad), 115	spam_setup (do_ldet), 29
print.summary.moransad	<pre>spam_update_setup (do_ldet), 29</pre>
(lm.morantest.sad), 106	SpatialFiltering, <i>124</i> , 181
<pre>print.summary.sarlm(summary.sarlm), 197</pre>	spautolm, <i>34</i> , <i>121</i> , <i>122</i> , 183
<pre>print.summary.spautolm(spautolm), 183</pre>	spdep, 189
print.summary.stsls(stsls), 192	<pre>spNamedVec (set.spChkOption), 171</pre>
probmap, 25, 38, 42, 158	spweights.constants, 190
prunecost, 160	ssw, 191
prunemst, <i>161</i> , 161	stsls, 192
	subgraph_eigenw(eigenw),43
queencell (cell2nb), 23	subset.listw, 194
	subset.nb, <i>195</i> , 195
read.dat2listw(read.gwt2nb), 163	summary.gmsar(GMerrorsar),61
read.gal, 82, 140, 162, 164, 207	<pre>summary.lagImpact(impacts), 74</pre>
read.geoda (read.gal), 162	summary.lagmess(lagmess), 91

```
summary.listw(summary.nb), 196
summary.lm, 198
summary.LMtestlist(lm.LMtests), 101
summary.localmoransad(localmoran.sad),
        115
summary.mcmc, 76
summary.moransad(lm.morantest.sad), 106
summary.nb, 22, 23, 43, 78, 140, 144, 145,
        152, 155, 163, 196
summary.sarlm, 197
summary.spautolm(spautolm), 183
summary.stsls(stsls), 192
sym.attr.nb(is.symmetric.nb), 81
Szero (spweights.constants), 190
tolerance.nb, 199
tri2nb, 136, 200, 201
trMat (house), 71
trW, 30, 31, 52, 76, 98, 186, 202
union.nb, 134
union.nb (nb.set.operations), 134
usa48.nb (used.cars), 204
used.cars, 204
vcov.sarlm(residuals.sarlm), 165
vi2mrc (cell2nb), 23
Wald1.sarlm(LR.sarlm), 118
wheat, 205
write.nb.gal, 206
write.sn2dat(read.gwt2nb), 163
write.sn2gwt(read.gwt2nb), 163
x (getisord), 59
xyz (getisord), 59
y (getisord), 59
```