

# Package ‘fastLISA’

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**Type** Package

**Title** Fast Local Indicators of Spatial Association (LISA)

**Version** 1.0.0

**Description** Computes various Local Indicators of Spatial Association (LISA) statistics, including univariate and bivariate local Moran's I, Empirical Bayes local Moran's I, univariate and multivariate local Geary's C, and Getis-Ord G and G\* statistics. The methods follow Anselin (1995), Getis and Ord (1992), and Anselin (2019). Leverages a high-performance, plain-C backend with optional OpenMP multi-core support for fast permutation-based pseudo-p-value calculation. Accepts any 'spdep' listw spatial weight matrix, including custom and non-contiguity weights. Uses sample standardisation (n-1) and 'rgeoda'-style permutation p-values. Output cluster codes match 'rgeoda' conventions, including the Isolated category for observations without neighbours.

**License** GPL-3

**Encoding** UTF-8

**Imports** stats

**Suggests** spdep

**NeedsCompilation** yes

**SystemRequirements** C99, optional OpenMP

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local_g	<i>Local Getis-Ord G</i>
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### Description

local\_g computes the Getis-Ord local  $G_i$  statistic, a Local Indicator of Spatial Association that detects local clustering of high values (“hot spots”) and low values (“cold spots”). For observation  $i$ , with row-standardised spatial weights  $w_{ij}^*$  and the focal value *excluded* from both the lag and the denominator,

$$G_i = \frac{\sum_{j \neq i} w_{ij}^* x_j}{\sum_k x_k - x_i}.$$

A large  $G_i$  indicates that  $i$  is surrounded by high values; a small  $G_i$  indicates a low-value neighbourhood.  $G_i$  contains no self term; see [local\\_gstar](#) for the self-inclusive  $G_i^*$ .

### Usage

```
local_g(
  x,
  listw,
  nsim = 999L,
  iseed = NULL,
  p.value = 0.05,
  n.cores = 1L,
  moments = FALSE,
  p.method = c("count", "rank")
)
```

### Arguments

x	Numeric vector of length n.
listw	A listw object from <b>spdep</b> .
nsim	Integer; number of permutations. Default 999L. Must be at least 1.
iseed	Integer seed for RNG, or NULL.
p.value	Numeric significance cutoff. Default 0.05.
n.cores	Integer; number of OpenMP threads. Default 1L; set higher to use multiple cores. Ignored on platforms without OpenMP.
moments	Logical; if TRUE, append permutation-distribution moments E.Gi, Var.Gi, Skew.Gi, and Kurt.Gi. Default FALSE.

**p.method** Character; permutation pseudo p-value method, "count" (default) or "rank". "count" counts permutations at least as extreme as the observed value; "rank" uses **spdep**'s ties-averaged rank of the observed value. Both return a folded (smaller-tail) value.

## Details

Inference uses a conditional permutation test: the focal value  $x_i$  is held fixed while the neighbouring values are randomly permuted `nsim` times. The pseudo p-value is folded (two-tailed),

$$p_i = \frac{\min(g, \text{nsim} - g) + 1}{\text{nsim} + 1},$$

where  $g$  is the number of permutations with  $G_i^{\text{perm}} \geq G_i^{\text{obs}}$ . The standardised score is  $Z.G_i = (G_i - E_{\text{perm}}) / \sqrt{\text{Var}_{\text{perm}}}$ , computed from the permutation mean and variance; `Skew.Gi` and `Kurt.Gi` (when `moments = TRUE`) follow the **e1071** type-3 convention.

Observations with a missing  $x$  value are labelled `Undefined` and observations with no neighbours are labelled `Isolated`; both receive NA for the p-value, Z-score and moments. The C backend re-seeds its random number generator per observation, so results are identical for any `n.cores`; `n.cores` is ignored when the package is built without OpenMP.

## Value

A numeric matrix of class `c("localG", "matrix", "array")` with columns `Gi`, `Z.Gi`, and `Pr` (folded) `Sim`. When `moments = TRUE`, the permutation-moment columns are appended. It has the following attributes:

**cluster** A significance-filtered factor with levels `Not significant`, `High-High`, `Low-Low`, `Undefined` and `Isolated`.

**gstari** Logical flag set to `FALSE` indicating local G (not G\*).

**call** The matched call.

## References

Getis, A. and Ord, J. K. (1992) The Analysis of Spatial Association by Use of Distance Statistics. *Geographical Analysis* **24**(3), 189–206. doi:10.1111/j.15384632.1992.tb00261.x

Ord, J. K. and Getis, A. (1995) Local Spatial Autocorrelation Statistics: Distributional Issues and an Application. *Geographical Analysis* **27**(4), 286–306. doi:10.1111/j.15384632.1995.tb00912.x

## Examples

```
lw <- spdep::nb2listw(spdep::cell2nb(7, 7))
x <- as.numeric(seq_len(49))
res <- local_g(x, lw, nsim = 99L, n.cores = 1L)
head(res)
```

local\_geary

*Univariate Local Geary's C***Description**

local\_geary computes the univariate local Geary's  $C_i$ , a squared-difference Local Indicator of Spatial Association that measures how much a unit differs from its neighbours. On the sample  $(n - 1)$  standardised variable  $z$  (when scale = TRUE) with row-standardised weights  $w_{ij}^*$ ,

$$C_i = \sum_j w_{ij}^* (z_i - z_j)^2 = z_i^2 - 2z_i \text{lag}(z)_i + \text{lag}(z^2)_i.$$

A small  $C_i$  means  $i$  resembles its neighbours (positive spatial association); a large  $C_i$  means it differs from them (negative association, a spatial outlier).

**Usage**

```
local_geary(
  x,
  listw,
  nsim = 999L,
  scale = TRUE,
  iseed = NULL,
  p.value = 0.05,
  n.cores = 1L,
  moments = FALSE,
  p.method = c("count", "rank")
)
```

**Arguments**

x	Numeric vector of length n.
listw	A listw object from <b>spdep</b> .
nsim	Integer; number of permutations. Default 999L. Must be at least 1.
scale	Logical; if TRUE (default), standardise data in R.
iseed	Integer seed for RNG, or NULL.
p.value	Numeric significance cutoff. Default 0.05.
n.cores	Integer; number of OpenMP threads. Default 1L; set higher to use multiple cores. Ignored on platforms without OpenMP.
moments	Logical; if TRUE, append permutation-distribution moments E.Ci, Var.Ci, Skew.Ci, and Kurt.Ci. Default FALSE.
p.method	Character; permutation pseudo p-value method, "count" (default) or "rank". "count" counts permutations at least as extreme as the observed value; "rank" uses <b>spdep</b> 's ties-averaged rank of the observed value. Both return a folded (smaller-tail) value.

## Details

Inference uses a one-tailed conditional permutation test (nsim reps): the observed  $C_i$  is compared with the permutation mean to select the tail, and

$$p_i = \frac{t + 1}{\text{nsim} + 1},$$

where  $t$  counts permuted statistics in that tail. The standardised score is  $Z.C_i = (C_i - E_{\text{perm}}) / \sqrt{\text{Var}_{\text{perm}}}$ ; Skew.Ci/Kurt.Ci (when moments = TRUE) follow the **e1071** type-3 convention. The cluster factor splits significant positive association into High-High/Low-Low/Other Positive and labels significant dissimilarity Negative.

Observations with a missing  $x$  value are labelled Undefined and observations with no neighbours are labelled Isolated; both receive NA for the p-value, Z-score and moments. The C backend re-seeds its random number generator per observation, so results are identical for any n.cores; n.cores is ignored when the package is built without OpenMP.

## Value

A numeric matrix of class c("localC", "matrix", "array") with columns Ci, Z.Ci, and Pr Sim. When moments = TRUE, the permutation-moment columns are appended. It has the following attributes:

**cluster** A significance-filtered factor with levels Not significant, High-High, Low-Low, Other Positive, Negative, Undefined, and Isolated.

**call** The matched call.

## References

Anselin, L. (1995) Local Indicators of Spatial Association—LISA. *Geographical Analysis* **27**(2), 93–115. doi:10.1111/j.15384632.1995.tb00338.x

## Examples

```
lw <- spdep::nb2listw(spdep::cell2nb(7, 7))
x <- as.numeric(seq_len(49))
res <- local_geary(x, lw, nsim = 99L, n.cores = 1L)
head(res)
```

## Description

local\_gstar computes the Getis-Ord local  $G_i^*$  statistic, the self-inclusive companion of `local_g`: observation  $i$  is treated as its own neighbour (weight 1). With  $m_i$  valid neighbours, row-standardised neighbour weights  $w_{ij}$ , and global total  $S = \sum_k x_k$ ,

$$G_i^* = \frac{\left( \frac{\sum_{j \in N_i} w_{ij} x_j}{\sum_j w_{ij}} \right) m_i + x_i}{(m_i + 1) S},$$

i.e. the average value over the focal unit and its neighbours divided by the global sum. Large  $G_i^*$  flags a hot spot and small  $G_i^*$  a cold spot, with the focal unit included.

## Usage

```
local_gstar(
  x,
  listw,
  nsim = 999L,
  iseed = NULL,
  p.value = 0.05,
  n.cores = 1L,
  moments = FALSE,
  p.method = c("count", "rank")
)
```

## Arguments

x	Numeric vector of length n.
listw	A listw object from <b>spdep</b> .
nsim	Integer; number of permutations. Default 999L. Must be at least 1.
iseed	Integer seed for RNG, or NULL.
p.value	Numeric significance cutoff. Default 0.05.
n.cores	Integer; number of OpenMP threads. Default 1L; set higher to use multiple cores. Ignored on platforms without OpenMP.
moments	Logical; if TRUE, append permutation-distribution moments E.G*i, Var.G*i, Skew.G*i, and Kurt.G*i. Default FALSE.
p.method	Character; permutation pseudo p-value method, "count" (default) or "rank". "count" counts permutations at least as extreme as the observed value; "rank" uses <b>spdep</b> 's ties-averaged rank of the observed value. Both return a folded (smaller-tail) value.

## Details

Inference uses a conditional permutation test (nsim reps), with the focal  $x_i$  held fixed while neighbour values are permuted. The pseudo p-value is folded (two-tailed),

$$p_i = \frac{\min(g, nsim - g) + 1}{nsim + 1},$$

where  $g$  counts permutations with  $G_i^{*\text{perm}} \geq G_i^{*\text{obs}}$ . The standardised score  $Z.G_i^* = (G_i^* - E_{\text{perm}}) / \sqrt{\text{Var}_{\text{perm}}}$  uses the permutation moments; `Skew.G*i/Kurt.G*i` (when `moments = TRUE`) follow the **e1071** type-3 convention.

Observations with a missing  $x$  value are labelled `Undefined` and observations with no neighbours are labelled `Isolated`; both receive `NA` for the  $p$ -value,  $Z$ -score and moments. The `C` backend re-seeds its random number generator per observation, so results are identical for any `n.cores`; `n.cores` is ignored when the package is built without `OpenMP`.

## Value

A numeric matrix of class `c("localG", "matrix", "array")` with columns `G*i`, `Z.G*i`, and `Pr(folded)` `Sim`. When `moments = TRUE`, the permutation-moment columns are appended. It has the following attributes:

**cluster** A significance-filtered factor with levels `Not significant`, `High-High`, `Low-Low`, `Undefined` and `Isolated`.

**gstari** Logical flag set to `TRUE` indicating local  $G^*$ .

**call** The matched call.

## References

Getis, A. and Ord, J. K. (1992) The Analysis of Spatial Association by Use of Distance Statistics. *Geographical Analysis* **24**(3), 189–206. [doi:10.1111/j.15384632.1992.tb00261.x](https://doi.org/10.1111/j.15384632.1992.tb00261.x)

Ord, J. K. and Getis, A. (1995) Local Spatial Autocorrelation Statistics: Distributional Issues and an Application. *Geographical Analysis* **27**(4), 286–306. [doi:10.1111/j.15384632.1995.tb00912.x](https://doi.org/10.1111/j.15384632.1995.tb00912.x)

## Examples

```
lw <- spdep::nb2listw(spdep::cell2nb(7, 7))
x <- as.numeric(seq_len(49))
res <- local_gstar(x, lw, nsim = 99L, n.cores = 1L)
head(res)
```

---

local\_moran

*Univariate Local Moran's I*

---

## Description

`local_moran` computes the univariate local Moran's  $I_i$ , the classic Local Indicator of Spatial Association measuring whether a value coincides with its neighbours' average. On the sample  $(n - 1)$  standardised variable  $z$  with row-standardised weights  $w_{ij}^*$ ,

$$I_i = z_i \sum_j w_{ij}^* z_j = z_i \text{lag}(z)_i.$$

A positive  $I_i$  indicates similarity to neighbours (a High-High or Low-Low cluster); a negative  $I_i$  indicates a spatial outlier (High-Low or Low-High). Standardisation uses the sample standard deviation ( $n - 1$  denominator).

**Usage**

```
local_moran(
  x,
  listw,
  nsim = 999L,
  iseed = NULL,
  p.value = 0.05,
  n.cores = 1L,
  moments = FALSE,
  p.method = c("count", "rank")
)
```

**Arguments**

<code>x</code>	Numeric vector of length <code>n</code> .
<code>listw</code>	A <code>listw</code> object from <b>spdep</b> .
<code>nsim</code>	Integer; number of permutations. Default 999L. Must be at least 1.
<code>iseed</code>	Integer seed for RNG, or NULL.
<code>p.value</code>	Numeric significance cutoff. Default 0.05.
<code>n.cores</code>	Integer; number of OpenMP threads. Default 1L; set higher to use multiple cores. Ignored on platforms without OpenMP.
<code>moments</code>	Logical; if TRUE, append the permutation-distribution moments <code>E.Ii</code> , <code>Var.Ii</code> , <code>Skew.Ii</code> , and <code>Kurt.Ii</code> to the result matrix. Default FALSE.
<code>p.method</code>	Character; permutation pseudo p-value method, "count" (default) or "rank". "count" counts permutations at least as extreme as the observed value; "rank" uses <b>spdep</b> 's ties-averaged rank of the observed value. Both return a folded (smaller-tail) value.

**Details**

This is the  $x = y$  special case of bivariate local Moran's I, so the computation delegates to [local\\_moran\\_bv](#) (with `scale = TRUE`) and relabels the columns. Inference uses a conditional permutation test (`nsim` reps) with the folded two-tailed pseudo p-value  $p_i = (\min(g, nsim - g) + 1) / (nsim + 1)$ , where  $g$  is the number of permutations with  $I^{\text{perm}} \geq I^{\text{obs}}$ . The standardised score is  $Z.I_i = (I_i - E_{\text{perm}}) / \sqrt{\text{Var}_{\text{perm}}}$  and `Skew.Ii/Kurt.Ii` (when `moments = TRUE`) use the **e1071** type-3 convention. NA observations are Undefined and neighbourless observations are Isolated (both NA in p-value, Z-score and moments). The C backend re-seeds its random number generator per observation, so results are identical for any `n.cores`.

**Value**

A numeric matrix of class `c("localmoran", "matrix", "array")` with `n` rows and 3 columns by default:

**Ii** Observed univariate Moran statistic.

**Z.Ii** Standardised Z-score computed from permutation moments.

**Pr(folded) Sim** rgeoda-style folded empirical permutation p-value.

When `moments = TRUE`, the permutation-distribution columns `E.Ii`, `Var.Ii`, `Skew.Ii`, and `Kurt.Ii` are appended. The matrix has the following attributes:

**quadr** Moran scatter-plot quadrant classification.

**cluster** A significance-filtered factor with levels `Not significant`, `High-High`, `Low-Low`, `Low-High`, `High-Low`, `Undefined` and `Isolated`.

## References

Anselin, L. (1995) Local Indicators of Spatial Association—LISA. *Geographical Analysis* **27**(2), 93–115. doi:10.1111/j.15384632.1995.tb00338.x

## Examples

```
lw <- spdep::nb2listw(spdep::cell2nb(7, 7))
x <- as.numeric(seq_len(49))
res <- local_moran(x, lw, nsim = 99L, n.cores = 1L, moments = TRUE)
head(res)
```

---

local\_moran\_bv

*Bivariate Local Moran's I*

---

## Description

`local_moran_bv` computes the bivariate local Moran's  $I_{bv,i}$ , which correlates a variable  $x$  at  $i$  with the spatial lag of a second variable  $y$  over  $i$ 's neighbours. On the sample  $(n - 1)$  standardised variables  $z_x, z_y$  (when `scale = TRUE`) with row-standardised weights  $w_{ij}^*$ ,

$$I_{bv,i} = z_{x,i} \sum_j w_{ij}^* z_{y,j} = z_{x,i} \text{lag}(z_y)_i.$$

A positive  $I_{bv,i}$  means  $i$ 's  $x$  value coincides with high lagged  $y$  nearby; a negative value indicates spatial mismatch. The univariate local Moran's I is the special case  $x = y$  (see `local_moran`). The backend is plain C with optional OpenMP parallelism.

## Usage

```
local_moran_bv(
  x,
  y,
  listw,
  nsim = 999L,
  scale = TRUE,
  iseed = NULL,
  p.value = 0.05,
  n.cores = 1L,
```

```

moments = FALSE,
p.method = c("count", "rank")
)

```

### Arguments

x	Numeric vector of length n; the first variable.
y	Numeric vector of length n; the second variable (lagged).
listw	A listw object from <b>spdep</b> (any style: "W", "B", "C", etc., including custom distance-decay weights). Observations with no neighbours receive cluster code 6 (Isolated).
nsim	Integer; number of permutations for the pseudo p-value. Default 999L.
scale	Logical; if TRUE (default), x and y are standardised in R (sample std dev). Set to FALSE only if you have pre-standardised the data.
iseed	Integer seed for the RNG, or NULL (default) to use the package default (123456789). Passed as the seed argument to the C backend.
p.value	Numeric; observations with $p > p.value$ are recoded to cluster 0 (Not significant). Default 0.05.
n.cores	Integer; number of OpenMP threads. Default 1L; set higher to use multiple cores. Ignored on platforms without OpenMP.
moments	Logical; if TRUE, append the permutation-distribution moments $E.I_{bv,i}$ , $Var.I_{bv,i}$ , $Skew.I_{bv,i}$ , and $Kurt.I_{bv,i}$ to the result matrix. Default FALSE.
p.method	Character; permutation pseudo p-value method, "count" (default) or "rank". "count" counts permutations at least as extreme as the observed value; "rank" uses <b>spdep</b> 's ties-averaged rank of the observed value. Both return a folded (smaller-tail) value.

### Details

Inference uses a conditional permutation test (`nsim` reps): the neighbour  $y$ -values are permuted with the focal observation held fixed. See the *P-values* and *Cluster codes* sections below for the folded p-value and the cluster coding. The standardised score is  $Z.I_{bv,i} = (I_{bv,i} - E_{perm}) / \sqrt{Var_{perm}}$ ;  $Skew.I_{bv,i}/Kurt.I_{bv,i}$  (when `moments = TRUE`) follow the **e1071** type-3 convention. NA observations are Undefined and neighbourless observations are Isolated (both NA in p-value, Z-score and moments). The C backend re-seeds its random number generator per observation, so results are identical for any `n.cores`.

### Value

A numeric matrix of class `c("localmoran", "matrix", "array")` with `n` rows and 3 columns by default:

**Ibvi** Observed bivariate Moran statistic.

**Z.Ibvi** Standardised Z-score computed from permutation moments.

**Pr(folded) Sim** rgeoda-style folded empirical permutation p-value.

When `moments = TRUE`, the permutation-distribution columns `E.Ibvi`, `Var.Ibvi`, `Skew.Ibvi`, and `Kurt.Ibvi` are appended. The matrix has the following attributes:

**quadr** A data.frame with three factor columns (`mean`, `median`, `pysal`) giving the Moran scatter-plot quadrant for each observation, computed on the original (unscaled) data scale.

**cluster** A factor representing cluster classification (Not significant, High-High, Low-Low, Low-High, High-Low, Undefined, Isolated).

**call** The matched call.

### Standardisation

Both `x` and `y` are standardised using the **sample** standard deviation ( $n - 1$  denominator) in R before computing the statistic, consistent with the `blisa` backend.

### P-values

Permutation p-values use the folded two-tailed formula matching `rgeoda`:

$$p = (\min(\#\{perm \geq obs\}, \#\{perm < obs\}) + 1) / (nsim + 1)$$

No normal approximation is computed.

### Cluster codes

The returned `cluster` factor attribute is based on integer codes 0–6:

- 0 Not significant
- 1 High-High
- 2 Low-Low
- 3 Low-High
- 4 High-Low
- 5 Undefined (NA input)
- 6 Isolated (no neighbours)

Codes 5 and 6 are preserved regardless of the significance cutoff.

### References

Anselin, L. (1995) Local Indicators of Spatial Association—LISA. *Geographical Analysis* **27**(2), 93–115. doi:10.1111/j.15384632.1995.tb00338.x

### Examples

```
lw <- spdep::nb2listw(spdep::cell2nb(7, 7))
x <- as.numeric(seq_len(49))
y <- rev(x)
res <- local_moran_bv(x, y, lw, nsim = 99L, n.cores = 1L, moments = TRUE)
head(res)
attr(res, "quadr")
```

---

local_moran_eb	<i>Local Moran's I with Empirical Bayes (EB) Rate</i>
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---

### Description

local\_moran\_eb computes local Moran's I on Empirical Bayes (EB) variance-stabilised rates, for event-count data observed over a population at risk. Raw rates  $p_i = \text{event}_i / \text{base}_i$  from small populations are noisy; EB standardisation shrinks them toward the global rate  $b = \sum \text{event} / \sum \text{base}$  using a variance component  $\hat{a}$ ,

$$z_i = \frac{p_i - b}{\sqrt{\hat{a} + b / \text{base}_i}},$$

and local Moran's I is then computed on  $z$ . This follows the GeoDa/libgeoda EBLocalMoran formulation.

### Usage

```
local_moran_eb(
  event,
  base,
  listw,
  nsim = 999L,
  iseed = NULL,
  p.value = 0.05,
  n.cores = 1L,
  moments = FALSE,
  p.method = c("count", "rank")
)
```

### Arguments

event	Numeric vector of events (e.g. case counts).
base	Numeric vector of populations at risk.
listw	A listw object from <b>spdep</b> .
nsim	Integer; number of permutations. Default 999L. Must be at least 1.
iseed	Integer seed for RNG, or NULL.
p.value	Numeric significance cutoff. Default 0.05.
n.cores	Integer; number of OpenMP threads. Default 1L; set higher to use multiple cores. Ignored on platforms without OpenMP.
moments	Logical; if TRUE, append the permutation-distribution moments E.Ii, Var.Ii, Skew.Ii, and Kurt.Ii to the result matrix. Default FALSE.
p.method	Character; permutation pseudo p-value method, "count" (default) or "rank". "count" counts permutations at least as extreme as the observed value; "rank" uses <b>spdep</b> 's ties-averaged rank of the observed value. Both return a folded (smaller-tail) value.

## Details

Two standardisations are applied and are not redundant: the EB rate standardisation above stabilises the rate variance, and the usual sample  $(n - 1)$  z-score standardisation that local Moran's I requires is then applied to the EB rates internally. Because univariate Moran's I on a standardised variable equals the bivariate statistic with  $x = y$ , the permutation engine is reused via `local_moran_bv`. Inference is a conditional permutation test (`nsim` reps); the folded two-tailed pseudo p-value and the score  $Z.I_i = (I_i - E_{perm}) / \sqrt{Var_{perm}}$  are as in `local_moran`, and `Skew.Ii/Kurt.Ii` (when `moments = TRUE`) use the **e1071** type-3 convention.

Observations with NA event/base or base  $\leq 0$  are labelled Undefined; observations with no neighbours are Isolated; both receive NA for the p-value, Z-score and moments. The C backend re-seeds its random number generator per observation, so results are identical for any `n.cores`.

## Value

A numeric matrix of class `c("local_moran_eb", "matrix")` with `n` rows and 3 columns by default:

**Ii** Observed Local Moran statistic computed on EB-standardised rates.

**Z.Ii** Standardised Z-score computed from permutation moments.

**Pr(folded) Sim** Folded empirical permutation p-value.

When `moments = TRUE`, the permutation-distribution columns `E.Ii`, `Var.Ii`, `Skew.Ii`, and `Kurt.Ii` are appended. The matrix has the following attributes:

**quadr** Moran scatter-plot quadrant classification.

**cluster** A significance-filtered factor with levels Not significant, High-High, Low-Low, Low-High, High-Low, Undefined and Isolated.

**call** The matched call.

**nsim** Number of simulations used.

## References

Assunção, R. M. and Reis, E. A. (1999) A new proposal to adjust Moran's I for population density. *Statistics in Medicine* **18**(16), 2147–2162. doi:10.1002/(SICI)10970258(19990830)18:16<2147::AID-SIM179>3.0.CO;2I

Anselin, L. (1995) Local Indicators of Spatial Association—LISA. *Geographical Analysis* **27**(2), 93–115. doi:10.1111/j.15384632.1995.tb00338.x

## Examples

```
lw <- spdep::nb2listw(spdep::cell2nb(7, 7))
event <- as.numeric(seq_len(49))
base <- rep(100, 49)
res <- local_moran_eb(event, base, lw, nsim = 99L, n.cores = 1L)
head(res)
```

---

local\_multigeary      *Multivariate Local Geary's C*

---

### Description

local\_multigeary computes the multivariate local Geary's  $C_i$  (Anselin 2019), the average across  $K$  variables of the univariate squared-difference statistic. On the sample  $(n - 1)$  standardised variables  $z^1, \dots, z^K$  (when `scale = TRUE`) with row-standardised weights  $w_{ij}^*$ ,

$$C_i = \frac{1}{K} \sum_{v=1}^K \sum_j w_{ij}^* (z_i^v - z_j^v)^2.$$

A small  $C_i$  indicates that  $i$  is similar to its neighbours across all variables (positive association); a large  $C_i$  indicates multivariate dissimilarity (a spatial outlier).

### Usage

```
local_multigeary(
  df,
  listw,
  nsim = 999L,
  scale = TRUE,
  iseed = NULL,
  p.value = 0.05,
  n.cores = 1L,
  moments = FALSE,
  p.method = c("count", "rank")
)
```

### Arguments

<code>df</code>	A data.frame or matrix with selected variables.
<code>listw</code>	A listw object from <b>spdep</b> .
<code>nsim</code>	Integer; number of permutations. Default 999L. Must be at least 1.
<code>scale</code>	Logical; if TRUE (default), standardise each variable in R.
<code>iseed</code>	Integer seed for RNG, or NULL.
<code>p.value</code>	Numeric significance cutoff. Default 0.05.
<code>n.cores</code>	Integer; number of OpenMP threads. Default 1L; set higher to use multiple cores. Ignored on platforms without OpenMP.
<code>moments</code>	Logical; if TRUE, append permutation-distribution moments <code>E.Ci</code> , <code>Var.Ci</code> , <code>Skew.Ci</code> , and <code>Kurt.Ci</code> . Default FALSE.
<code>p.method</code>	Character; permutation pseudo p-value method, "count" (default) or "rank". "count" counts permutations at least as extreme as the observed value; "rank" uses <b>spdep</b> 's ties-averaged rank of the observed value. Both return a folded (smaller-tail) value.

## Details

Inference uses a one-tailed conditional permutation test (`nsim` reps); each replicate applies the same permuted neighbour configuration to every variable. The observed  $C_i$  is compared with the permutation mean to choose the tail, and  $p_i = (t + 1)/(nsim + 1)$ , where  $t$  counts permuted statistics in that tail. The standardised score is  $Z.C_i = (C_i - E_{perm})/\sqrt{Var_{perm}}$ ; `Skew.Ci/Kurt.Ci` (when `moments = TRUE`) follow the **e1071** type-3 convention. Significant units are labelled `Positive` (similar) or `Negative` (dissimilar).

Rows with any missing value are labelled `Undefined` and observations with no neighbours are labelled `Isolated`; both receive NA for the p-value, Z-score and moments. The C backend re-seeds its random number generator per observation, so results are identical for any `n.cores`; `n.cores` is ignored when the package is built without `OpenMP`.

## Value

A numeric matrix of class `c("localC", "matrix", "array")` with columns `Ci`, `Z.Ci`, and `Pr Sim`. When `moments = TRUE`, the permutation-moment columns are appended. It has the following attributes:

**cluster** A significance-filtered factor with levels `Not significant`, `Positive`, `Negative`, `Undefined`, and `Isolated`.

**call** The matched call.

## References

Anselin, L. (2019) A Local Indicator of Multivariate Spatial Association: Extending Geary's *c*. *Geographical Analysis* **51**(2), 133–150. doi:10.1111/gean.12164

## Examples

```
lw <- spdep::nb2listw(spdep::cell2nb(7, 7))
x <- as.numeric(seq_len(49))
df <- cbind(x, rev(x))
res <- local_multigeary(df, lw, nsim = 99L, n.cores = 1L)
head(res)
```

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