Package 'graphscan'

July 8, 2014

Type Package

Title Cluster detection with hypothesis free scan statistic

Version 0.1
Date 2013-10-23
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Description Multiple scan statistic with variable window for one dimension data and scan statistic based on connected components in 2D or 3D.
License GPL-2 GPL-3
Depends R (>= 3.0.2), ape, sp, methods, snowfall, rgl
NeedsCompilation yes
R topics documented:
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graphscan-package	Cluster detection with hypothesis free scan statistic

Description

Multiple scan statistic with variable window for one dimension data and scan statistic based on connected components in 2D or 3D.

Details

Package: graphscan
Type: Package
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Depends: R (>= 3.0.2), ape, sp, methods, snowfall, rgl

NeedsCompilation: yes

Packaged: 2014-05-07 07:44:19 UTC

Built: R 3.0.2; x86_64-pc-linux-gnu; 2014-05-07 07:44:20 UTC; unix

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3D data.

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distributions.

sample3d A 3D cluster example.

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This package implements a statistical method for detecting clusters in dimensions 1, 2 and 3 proposed by Cucala (2008,2009). In 1D, this hypothesis multiple scan statistic with variable window in 1D can detect positive clusters only, negative clusters only, or simultaneously positive or negative clusters. Positive clusters correspond to a particularly high concentration in events, while negative clusters correspond to a particularly low concentration in events. The concentration index of Cucala is based on the properties of the distances between order statistics under the hypothesis of uniform

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distribution of the events. The 1D functions are adapted to study the repartition of mutations along the genome and detect zones with numerous or few mutations when comparing two aligned DNA sequences. When studyind a population, i.e. several DNA sequences, pair comparisons can be created and analysed. In nD (2D or 3D), the flexible spatial scan test for case event data is implemented to detect only the most significant positive cluster. The candidate clusters are build from connected components. There are the number of case minus one candidate clusters. The concentration index of Cucala is based on the properties of the distances between order statistics under the hypothesis of Poisson distribution of the cases. The analysis for nD data returns both the Cucala and the Kulldorff concentration index.

Author(s)

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer Maintainer: David Abrial <graphscan@clermont.inra.fr>

References

Kulldorff, M. 1997. A spatial scan statistic, Commun. Statist. - Theory Meth., 26, 6, p. 1481-1496.

Cucala, L. 2008. A hypothesis-free multiple scan statistic with variable window, Biometrical Journal, 2, p. 299-310.

Cucala, L. 2009. A flexible spatial scan test for case event data, Computational Statistics and Data Analysis, 53, p. 2843-2850.

cluster

Performs cluster analysis on 'graphscan' class object.

Description

This function performs cluster detections for both 'graphscan_1d' and 'graphscan_nd' objects. For 'graphscan_nd' objects both Kulldorff and Cucala indices are computed.

Usage

```
cluster(gr, n_simulation = NULL, cluster_analysis = NULL)
```

Arguments

gr an object of class graphscan.

n_simulation number of simulations (default value 199) to compute the p-value. This value is

set during the generation of the 'graphscan' object or redefined by this argument.

cluster_analysis

type of cluster detection. Possible values are "positive", "negative" or "both".

For nD detection only "positive" is possible.

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Details

This is the main function to run the cluster detection analysis from data and parameters contained in an object of class graphscan. The data come from the slot 'data' and the parameters from the slot 'param' of the 'gr' object. The results of the analysis are saved in the slot 'cluster' of the returned object (see value). The analysis for 1D data returns only the concentration index of Cucala. Analysis for nD data returns both the Cucala and the Kulldorff concentration index.

Value

Returns a 'graphscan' object containing the analysis in the slot 'cluster'. For 1D data, the slot 'cluster' is a list with three matrices 'cluster 1d raw', 'cluster 1d' and 'cluster 1d description'. 'cluster 1d raw' is a matrix with 7 columns containing the raw results of the analysis. 'xleft' and 'xright' columns indicate the boundaries of each detected cluster. 'index' is the concentration index of Cucala and 'pvalue' is the significativity of the cluster. 'positivity' is a boolean indicating if the cluster is positive or negative. 'id_segment' and 'id_serie' are the identifiers respectively of the clusters and the events series. 'cluster_1d' is a matrix with 9 columns containing treated results. Indeed, some clusters are cut into several pieces called segments because they contain one or more included cluster. This matrix indicates for each cluster the start ('xleft') and the end ('xright') of the non-overlapping segments (in 'cluster_1d_raw' matrix clusters are composed by only one segment potentialy overlapping with other detected clusters). Columns 'index', 'pvalue', 'positivity', 'id_segment' and 'id_serie' are the same than in 'cluster_1d_raw' matrix. The column 'n_segment' indicate how many segments compose the cluster and 'length' is size of each segment. 'cluster_1d_description' is a matrix with 4 columns giving general informations on all the clusters. 'n pos' and 'n neg' give respectively the number of positives and negatives clusters for each events series. 'l_pos' and 'l_neg' are respectively the ratio (in percent) of positives and negatives clusters total length. For nD data, the slot 'cluster' is a list with two 'SpatialPoints-DataFrame' named 'cluster_nd_cucala' and 'cluster_nd_kulldorff' and a vector of characters named 'cluster nd description'. The two 'SpatialPointsDataFrame' objects contain the points of the significant cluster, the 'index' of concentration of Cucala or Kulldorff, the 'radius' of circles to draw the cluster area, the 'pvalue', the number of cases and controls present in the cluster. The vector 'cluster_nd_description' gives a brief description.

Author(s)

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

References

Cucala, L. 2008. A hypothesis-free multiple scan statistic with variable window, Biometrical Journal, 2, p. 299-310.

Cucala, L. 2009. A flexible spatial scan test for case event data, Computational Statistics and Data Analysis, 53, p. 2843-2850.

See Also

graphscan_1d, graphscan_nd, plot, hist

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Examples

events_series

A 1D cluster example: events series example.

Description

A set of 9 series of 1000 events generated for the graphscan package.

Usage

```
data(events_series)
```

Format

Data is a list of 9 vectors of reals and a list named 'normalisation_factor' containing the normalisation factor of each events serie.

Source

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

See Also

```
graphscan_1d
```

```
data(events_series)
g1<-graphscan_1d(data=events_series,normalisation_factor=normalisation_factor)</pre>
```

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france_two_clusters A 2D cluster example

A 2D cluster example: France with two clusters.

Description

This data set is a 2D example with 1000 points ditributed on the French territory: 26 cases and 974 controls. The outline of France is included in object "france".

Usage

```
data(france_two_clusters)
```

Format

"france_two_clusters" and "france" are of class "SpatialPointsDataFrame"

Source

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

See Also

graphscan_nd

Examples

```
data(france_two_clusters)
plot(france)
points(france_two_clusters[france_two_clusters$cases==1,],pch=16,col="red")
```

graphscan-class

Class "graphscan"

Description

Class of graphscan objects used to store data, parameters and results of cluster analysis with the method of Cucala.

Objects from the Class

Objects can be created by calls of the form "graphscan_ld(data, ...)" or "graphscan_nd(data, ...)".

Slots

```
data: Object of class "list", containing the data as events series.

param: Object of class "list", containing the parameters used in the analysis.

cluster: Object of class "list", containing the results.
```

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Methods

```
cluster signature(gr = "graphscan"): performed the cluster analysis.
hist signature(x = "graphscan"): draw an histogram of the length of clusters in multiple events series analysis.
```

graphscan_plot signature(x = "graphscan"): plot clusters localisations or the 1D events distributions..

```
print signature(x = "graphscan"): print informations about a graphscan object.
show signature(object = "graphscan"): print informations about a graphscan object.
summary signature(object = "graphscan"): summary for graphscan objects.
```

Author(s)

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer Maintainer: David Abrial <graphscan@clermont.inra.fr>

References

Kulldorff, M. 1997. A spatial scan statistic, Commun. Statist. - Theory Meth., 26, 6, p. 1481-1496. Cucala, L. 2008. A hypothesis-free multiple scan statistic with variable window, Biometrical Journal, 2, p. 299-310.

Cucala, L. 2009. A flexible spatial scan test for case event data, Computational Statistics and Data Analysis, 53, p. 2843-2850.

See Also

```
graphscan_1d, graphscan_nd
```

Examples

```
data(events_series)
g4<-graphscan_1d(events_series,normalisation_factor=normalisation_factor)
g4<-cluster(g4)
graphscan_plot(g4)</pre>
```

graphscan_1d

Creates objects of class 'graphscan' using 1D data.

Description

This function produces objects of class graphscan 1d.

Usage

```
graphscan_1d(data, format = "fasta", events_series = "all", id = NULL,
n_simulation = 199, cluster_analysis = "both",
normalisation_factor = NULL, alpha = 0.05, cluster_user_choice = "positive")
```

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Arguments

data

main argument to defined the format of the data. 'data' can be a vector of character string corresponding to files names of aligned DNA sequences. In this case, the format can be precised with argument 'format'. 'data' can be also a list of class 'DNAbin' produced by 'read.dna' function of 'ape' package. In all cases, DNA sequences must be aligned. Finally 'data' can be a 'list' of numeric vector containing the positions of events. This list is called a series of events. These events are not be necessarily on the same segment and not also necessarily on a [0,1] segment. The argument 'normalisation_factor' allow to fix the upper and lower bounds of each events series.

format

a character string corresponding to the format of the DNA sequences contained in files of argument 'data'. This is used by 'read.dna' function of 'ape' package. Possibles values are "interleaved", "sequential", "clustal" or "fasta" (default).

events_series

used if 'data' is a set of files names of aligned DNA sequences or list of class 'DNAbin'. 'events_series' can be a list of the form 'list(A,B)' where 'A' and 'B' corresponding to 2 vectors of sequences identifiants. The crossing AxB product is made to obtained a list of the series of events corresponding to the comparison between each sequence from 'A' to each sequence from 'B'. 'events_series' can be also a character string containing "all", in this case all possible comparison between sequences is made.

a character string corresponding to the prefix used to create the names of the events series.

 $n_simulation$

number of simulations (default value 199) used to compute the p-values of clusters in a Monte-Carlo process. The value of 'n_simulation' is stored in the slot 'param' and can be modified by the function 'cluster'.

cluster_analysis

a character string corresponding to "positive", "negative" or "both" (default value) to detect respectively only the positives clusters, only the negatives clusters or both positives and negatives clusters. The value of 'cluster_analysis' is stored in the slot 'param' and can be modified by the function 'cluster'.

normalisation_factor

a list of vectors with a size equal to the number of events series. Each vector contain 2 integers: the minimum and the maximum for the events positions of series of events.

The maximum is the length of the DNA sequences if 'data' argument is a vector of character or an object of class "DNAbin". In these cases, the 'normalisation factor' is automatically compute by the function 'graphscan 1d'.

If 'data' is a 'list' of numeric vector containing the positions of events the 'normalisation_factor' must be specified as a 'list' containing the upper and lower bounds of each events series. The values of 'normalisation_factor' are stored in the slot 'param'.

alpha

the threshold of significativity (p-value) for keeping the candidate clusters. The value of 'alpha' is stored in the slot 'param'.

cluster_user_choice

use if 'cluster_analysis="both"'. 'cluster_user_choice' is a string character corresponding to "positive" (default value), "negative" or "random". If two can-

id

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didates clusters one positive and one negative have the same p-value this argument indicates how to choose between these 2 clusters. The value of 'cluster user choice' is stored in the slot 'param'.

Details

This function implements a statistical method for detecting clusters in 1D proposed by Cucala (2008). This hypothesis multiple scan statistic with variable window in 1D can detect positive clusters only, negative clusters only, or simultaneously positive or negative clusters. Positive clusters correspond to a particularly high concentration in events, while negative clusters correspond to a particularly low concentration in events. The concentration index of Cucala is based on the properties of the distances between oder statistics under the hypothesis of uniform distribution of the events. The 1D functions are adapted to study the repartition of mutations along the genome and detect zones with numerous or few mutations when comparing two aligned DNA sequences. When studying a population, i.e. several DNA sequences, pair comparisons can be created and analysed.

Value

'graphscan_1d' returns an object of class 'graphscan' with 3 slots:

param this slot contains the informations about data and the parameters used to perform

the analysis.

data this slot contains data of events series as a list 'x' of numeric vectors.

cluster this slot contains the results of the analysis. For 1d, three matrix 'cluster_1d_raw',

'cluster_1d' and 'cluster_1d_description' (see 'cluster' function for more de-

tails).

Author(s)

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

References

Cucala, L. 2008. A hypothesis-free multiple scan statistic with variable window, Biometrical Journal, 2, p. 299-310.

Cucala, L. 2009. A flexible spatial scan test for case event data, Computational Statistics and Data Analysis, 53, p. 2843-2850.

See Also

cluster,graphscan_2d

```
# example with 2 fasta format files containing each 2 DNA aligned sequences.
# the output object contain 6 events series.
dna_file<-list.files(path=system.file("extdata",package="graphscan"),
    pattern="fna",full.names=TRUE)
g1<-graphscan_1d(data=dna_file)</pre>
```

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```
# to perform only 4 comparisons between DNA sequences
# 1 vs 3, 1 vs 4, 2 vs 3 and 2 vs 4.
g2<-graphscan_1d(data=dna_file,events_series=list(1:2,3:4))
# example with 'DNABin' class object :
data(woodmouse)
g3<-graphscan_1d(data=woodmouse)
# example with a list of 9 events series
data(events_series)
g4<-graphscan_1d(events_series,normalisation_factor=normalisation_factor)</pre>
```

graphscan_nd

Creates objects of class 'graphscan' using 2D or 3D data.

Description

This function produces objects of class 'graphscan' using 'SpatialPointsDataFrame' class objects for 2D and 3D analysis.

Usage

```
graphscan_nd(data,field_cases = NULL,field_controls = NULL,
n_simulation = 199, alpha = 0.05)
```

Arguments

data	a 'SpatialPointsDataFrame	' object containing coordi	nates of cases and controls
	points The 'data frame'	of 'data' must contain tw	a numeric fields: one for

points. The 'data.frame' of 'data' must contain two numeric fields: one for the number of cases for each point (named by default 'cases') and one for the number of controls for each point (named by default 'controls'). The minimal

number of cases to perform the analysis is 2 and 50 for controls.

field_cases a character string to define the name of the field containing the number of cases

per points. If a field named 'cases' exists in the 'data.frame' this argument is

optional.

field_controls a character string to define the name of the field containing the number of con-

trols per points. If a field named 'controls' exists in the 'data.frame' this argu-

ment is optional.

n_simulation number of simulations (default value 199) used to compute the significativity of

the clusters i.e. the p-values computed with a Monte-Carlo process. The value of 'n_simulation' is stored in the slot 'param' and can be modified by the function

'cluster'.

alpha the threshold of significativity (p-value) to keep the candidate clusters. The

value of 'alpha' is stored in the slot 'param'.

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Details

This function implements a statistical method for detecting clusters in nD (2D or 3D) proposed by Cucala (2009). This flexible spatial scan test for case event data is implemented to detect the most significant positive cluster. The candidate clusters are build from connected components. There are the number of case minus one candidate clusters. The concentration index of Cucala is based on the properties of the distances between order statistics under the hypothesis of Poisson distribution of the cases. The analysis for nD data returns both the Cucala and the Kulldorff concentration index.

Value

'graphscan_nd' returns an object of class 'graphscan' with 3 slots:

param this slot contains the informations about the data and the parameters used to

perform the analysis.

data this slot contains a list with one item 'x': the 'SpatialPointsDataFrame' of the

'data' argument.

cluster this slot contains the results of the analysis: a list with three items two 'Spatial-

PointsDataFrame' named 'cluster_nd_cucala' and 'cluster_nd_kulldorff' and a vector of string named 'cluster_nd_description' (see 'cluster' function for more

details).

Author(s)

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

References

Kulldorff, M. 1997. A spatial scan statistic, Commun. Statist. - Theory Meth., 26, 6, p. 1481-1496.

Cucala, L. 2008. A hypothesis-free multiple scan statistic with variable window, Biometrical Journal, 2, p. 299-310.

Cucala, L. 2009. A flexible spatial scan test for case event data, Computational Statistics and Data Analysis, 53, p. 2843-2850.

See Also

```
graphscan_1d, cluster
```

```
data(france_two_clusters)
g3<-graphscan_nd(data=france_two_clusters)</pre>
```

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Description

plot clusters localisations in a 1D, 2D or 3D space. In 1D, the distribution (in frequencies) of number of events on each position can be ploted.

Usage

graphscan_plot(x,events_series=1,map=NULL,indice="cucala",sphere=TRUE,projection=FALSE)

Arguments

X	a graphscan object containing a cluster analysis.
events_series	a numeric or character vector containing 1D cluster identifiants of the events series to draw. If 'events_series="all" the distribution of number of clusters on each position is ploted.
map	a 'SpatialPolygons' or 'SpatialPolygonsDataFrame' object to add to the 2D graph of clusters localisations (generally the outline of the studied region).
indice	a character string used in nD, to define the type of index to draw. Possible values are "cucala" (default) and "kulldorff".
sphere	a boolean ("TRUE" by default) to define if the spheres used to represent the 3D envelope of the cluster are drawn.
projection	a boolean ("FALSE" by default) to draw a projection in 2D of a 3D cluster. Three plots are drawn respectively for 'y vs x', 'z vs x' and 'z vs y'.

Details

To draw the distribution of number of events on each position in 1D, the events series must be of same length and aligned. The 3D representation of cluster use 'OpenGL' (http://www.opengl.org) an environment for interactive 2D and 3D graphics. If the number of cases points is very important old computers will display graphics quite slowly. In this case, use the option "projection=TRUE".

Author(s)

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

References

Cucala, L. 2008. A hypothesis-free multiple scan statistic with variable window, Biometrical Journal, 2, p. 299-310.

Cucala, L. 2009. A flexible spatial scan test for case event data, Computational Statistics and Data Analysis, 53, p. 2843-2850.

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See Also

hist

Examples

```
# 1D example:
data(woodmouse)
g1<-graphscan_1d(data=woodmouse)</pre>
g1<-cluster(g1)
graphscan_plot(g1,events_series=3)
dev.new()
graphscan_plot(g1,events_series="all")
# 2D example:
data(france_two_clusters)
g2<-graphscan_nd(data=france_two_clusters)</pre>
g2<-cluster(g2)
graphscan_plot(g2,map=france)
# 3D example:
data(sample3d)
g3<-graphscan_nd(data=sample3d)
g3<-cluster(g3)
graphscan_plot(g3,projection=TRUE) # 2D plot
graphscan_plot(g3) # 3D plot
```

hist

Histogram of the 1D clusters lengths.

Description

This function draws a histogram of the length of the 1D clusters for several events series of same length and aligned.

Usage

```
hist(x,...)
```

Arguments

```
x a graphscan object containing 1D cluster analysis.
```

... arguments to be passed to methods.

Author(s)

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

sample3d

References

Cucala, L. 2008. A hypothesis-free multiple scan statistic with variable window, Biometrical Journal, 2, p. 299-310.

Cucala, L. 2009. A flexible spatial scan test for case event data, Computational Statistics and Data Analysis, 53, p. 2843-2850.

See Also

```
graphscan_plot
```

Examples

```
dna_file<-list.files(path=system.file("extdata",package="graphscan"),
  pattern="fna",full.names=TRUE)
g1<-graphscan_1d(data=dna_file)
g1<-cluster(g1)
hist(g1)</pre>
```

sample3d

A 3D cluster example.

Description

This data set is a 3D example with 1060 points ditributed into a cube. These points are devided into 60 cases and 1000 controls.

Usage

```
data(sample3d)
```

Format

"sample3d" is of class "SpatialPointsDataFrame"

Source

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

See Also

```
graphscan_nd
```

```
data(sample3d)
g1<-graphscan_nd(data=sample3d)</pre>
```

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summary

Summary for graphscan objects.

Description

gives a summary of a graphscan object including informationd on the data, the parameters and if present the analysis.

Usage

```
summary(object,...)
```

Arguments

object a graphscan object.... arguments to be passed to methods.

Details

If the object contain a cluster analysis, the results will be detailed.

Author(s)

Benoit Giron, David Abrial, Lionel Cucala, Myriam Charras-Garrido, Jocelyn De-Goer.

References

Cucala, L. 2008. A hypothesis-free multiple scan statistic with variable window, Biometrical Journal, 2, p. 299-310. Cucala, L. 2009. A flexible spatial scan test for case event data, Computational Statistics and Data Analysis, 53, p. 2843-2850.

See Also

print

```
# 1D example:
data(woodmouse)
g1<-graphscan_1d(data=woodmouse)
summary(g1)</pre>
```

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