Package 'dagR'

October 13, 2022

Type Package
Title Directed Acyclic Graphs: Analysis and Data Simulation
Version 1.2.1
Date 2022-10-08
Author Lutz P Breitling
Maintainer Lutz P Breitling <1.breitling@posteo.de>
Description Draw, manipulate, and evaluate directed
acyclic graphs and simulate corresponding data, as described in International Journal of Epidemiology 50(6):1772-1777.
Imports graphics, stats, utils
Suggests dagitty(>= 0.3-1)
License GPL-2
NeedsCompilation no
Repository CRAN

Date/Publication 2022-10-09 09:20:02 UTC

R topics documented:

lagR-package	2
ldd.arc	4
ldd.node	5
lddAngle	6
llCombs	6
ngle	7
nglePoint	8
ssoc.exists	8
prute.search	9
lag.adjust	10
lag.adjustment	11
lag.ancestors	12
lag.draw	13
lag.init	14

dealeand	16
dag.legend	
dag.letter	16
dag.letter2	17
dag.move	18
dag.search	18
dag.sim	19
dag.sim2	20
dagR2dagitty	23
demo.dag0	24
demo.dag1	24
demo.dag2	25
demo.dag3	26
demo.dag4	26
demo.dag5	27
demo.dag6	27
demo.dag7	28
distPoints	29
eval.paths	29
find.paths	30
garrows	31
inAngle	32
is.acvclic	32
is.in	33
is.unknown	33
msas	34
plot.dagRdag	35
print.dagRdag	36
rm.arc	37
rm.node	37
smoothArc	38
summary.dagRdag	39
summary_dagRdag	40
viv	41
write.paths	41
	43

Index

dagR-package

Directed Acyclic Graphs: Analysis and Data Simulation

Description

The package dagR contains a couple of functions to draw, manipulate and evaluate directed acyclic graphs (DAG), with a focus on epidemiologic applications, namely the assessment of adjustment sets and potentially biasing paths. The functions for finding and evaluating paths essentially implement the graphical algorithms outlined in Greenland (1999).

When using this package for your work, please cite Breitling (2010) and/or Breitling et al. (2022).

For motivations to use this package in epidemiology teaching and methodological research, please refer to Duan et al. (2022).

Note: As spelled out in the license, this suite of functions comes without any warranty, and cautious use is strongly advised. Although testing was carried out as meticulously as possible, it must be expected that bugs or errors remain, in particular in the early versions of the package. Please report any problems, concerns, but also suggestions for improvements or extensions to the author.

Important additions in future versions could be e.g. improved drawing routines with better formatting of alternative node symbols in the DAG (taking into account the string length) and algorithms with intelligent/efficient search for minimal adjustment sets.

Details

Package:	dagR
Type:	Package
Version:	1.2.1
Date:	2022-10-09
License:	GPL-2
LazyLoad:	yes

dag.init is used for setting up DAGs. See the code of the functions demo.dag0 to demo.dag0 for example code. To adjust and/or evalute DAGs for biasing paths, use dag.adjust, dag.draw for drawing a DAG. dag.search uses brute.search to evaluate all possible adjustment sets, allowing the identification of minimal sufficient adjustment sets using msas. dag.sim simulates data (normally distributed or binary) according to the causal structure given by a DAG object.

In version 1.2.0, generic S3 methods (print, plot, summary) for dagR-DAGs were implemented, but the original functions summary_dagRdag to summarize and dag.draw to plot a DAG object were preserved for backwards compatibility. Export functions to other packages were added upon a reviewer request.

Several helper functions currently are not hidden and should later be made internal.

Please see the NEWS file for version changes and known open issues.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Breitling LP, Duan C, Dragomir AD, Luta G (2022). Using dagR to identify minimal sufficient adjustment sets and to simulate data based on directed acyclic graphs. Int J Epidemiol 50(6):1772-1777 <doi: 10.1093/ije/dyab167>.

Duan C, Dragomir AD, Luta G, Breitling LP (2022). Reflection on modern methods: Understanding bias and data analytical strategies through DAG-based data simulations. Int J Epidemiol 50(6):2091-2097 <doi: 10.1093/ije/dyab096>.

Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic research. Epidemiology 10(1):37-48.

add.arc

Add an arc to a DAG.

Description

Conveniently add an arc to an existing DAG.

Usage

add.arc(dag, arc, type = 0)

Arguments

dag	The DAG to which an arc should be added.
arc	A vector of length 2, indicating from which node (first element) to which node (second element) the arc is to go. Note: the node numbering follows the numbering of the existing DAG (as shown in dag.draw with option numbering=T), not the numbering of dag.init.
type	0 (=default) for a directed arc, 1 for an undirected association.

Value

A DAG with the arc (and corresponding arc.type) added, and with the path-related variables (paths, pathsN, path.status, searchType, searchRes) removed.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

rm.arc, add.node, rm.node

4

add.node

Description

Conveniently adds a node to an existing DAG, inserting its coordinates and label before the outcome node. Also updates the arcs correspondingly.

Usage

add.node(dag, name = "unknown", type = 1, x = NA, y = NA)

Arguments

dag	The DAG to which the node is to be added.
name	Label for the node (defaults to "unknown").
type	Type of node (1=covariable, 2=unknown); defaults to 1.
x	X coordinate for the node position.
у	Y coordinate for the node position.

Details

If no x and y coordinates are provided, the function places the node in an arbitrary position, slightly different with each additional node, so that one can more easily replace the nodes afterwards using dag.move.

Value

A DAG with the new node added.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

See Also

rm.node, add.arc, rm.arc

addAngle

Description

Adds two radian angles together and applies modulus 2*pi. This is internally called by smoothArc, though hardly needed.

Usage

addAngle(a, b)

Arguments

а	Angle 1 in radian.
b	Angle 2 in radian.

Value

numeric value [0, 2*pi).

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

smoothArc

allCombs

Create all combinations of the elements of a vector.

Description

Creates a matrix with all combinations of 1 to all elements of the vector provided. Elements to occur in all combinations can be specified. This is internally called by brute.search.

Usage

allCombs(x, force = c(), trace = FALSE)

Arguments

х	A vector of elements of which combinations are to be formed.
force	A vector of elements that are supposed to occur in each combination.
trace	A boolean indicating if some output should be printed (TRUE) or not (FALSE=default)

angle

Value

A matrix with one combination per row. For the shorter combinations, the columns to the right are filled up with NA.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

brute.search

angle

Calculate radian angle of line between two points.

Description

Calculates the radian angle of the line connecting two points. Internally called by smoothArc.

Usage

angle(A, B)

Arguments

A	Vector of length two indicating the coordinates of the first point.
В	Vector of length two indicating the coordinates of the second point.

Value

A numeric value [0, 2*pi).

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

smoothArc, addAngle

anglePoint

Description

Calculates the coordinates of the point that is at a specific radian angle in a specific distance from a source point. Internally called by smoothArc.

Usage

anglePoint(A, angl, len)

Arguments

A	Vector of length two with the coordinates of the source point.
angl	Radian angle indicating into which direction the new point is to be calculated.
len	The distance at which the new point is situated from the source point.

Value

A vector of length two with the coordinates of the new point.

Note

Another pretty superfluous helper function...

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

smoothArc

assoc.exists Check if

Check if association between two DAG nodes exists.

Description

Checks if an association between two DAG nodes already exists, i.e. does not need to be introduced when adjusting for a shared child etc. Internally called by dag.adjustment.

Usage

assoc.exists(dag, a, b)

brute.search

Arguments

dag	The DAG to be dealt with.
а	First node.
b	Second node.

Value

A boolean indicating whether or not an association between first node and second node already exists.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

brute.search

Evaluate all possible adjustment sets of a DAG.

Description

Evaluates all adjustment sets of a DAG, optionally including adjustment sets including "unknown" nodes. If the DAG has a non-empty adjustment set, only adjustment sets including these adjustment variables are evaluated.

Usage

```
brute.search(dag, allow.unknown = FALSE, trace = TRUE, stop = 0)
```

Arguments

dag	The DAG to be evaluated.
allow.unknown	Boolean indicating "unknown" nodes should be featured in the adjustment sets to be evaluated (TRUE) or not (FALSE=default).
trace	Boolean indicating if some output should be produced (TRUE=default).
stop	If =0, all eligible adjustment sets are evaluated. If =1, evaluations are stopped after the first sufficient adjustment set has been evaluated. Defaults to 0.

Value

A dataframe with the first columns (X1..Xn) indicating the variables in the respective adjustment set evaluated. The column totalPaths indicates the number of paths found when adjusting for the respective set, and openPaths indicates the number of biasing paths.

The output produced by brute.search allows to manually identify sufficient and minimal sufficient adjustment sets, which in the future should preferably be done by a helper summary function. The evaluation of a complicated DAG like demo.dag2 can take quite some time, and future functions should either employ more intelligent algorithms to search specifically for sufficient sets, or they should allow e.g. the evaluation of adjustment sets of specific sizes.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

dag.adjust Adjust an existing DAG for covariables.

Description

Looks for associations introduced by adjusting for the covariables specified, then looks for biasing paths, and finally evaluates these paths.

Usage

dag.adjust(dag, A = c())

Arguments

dag	The DAG to be adjusted (or evaluated).
A	Vector indicating the adjustment set. The numbering is according to the nodes vector of the DAG, which is shown e.g. in the legend of a DAG drawn by dag.draw. This numbering is different (+1) from the one used in dag.init, because the nodes vector also contains the exposure at position 1 (in contrast to the covariables vector used in dag.init)!

Details

If the adjustment set is empty, the function only looks for biasing paths and evalutes these.

Value

A DAG with the adjustment set A, and possibly with additional associations introduced by adjustment, biasing paths found, and the status of these.

If adjustment set is not empty, searchType and searchRes are set to NULL.

Note

CAVE: Do not apply this to an already adjusted DAG, since this might not be handled appropriately (see documentation of dag.adjustment called by dag.adjust).

Note

dag.adjustment

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Breitling LP, Duan C, Dragomir AD, Luta G (2022). Using dagR to identify minimal sufficient adjustment sets and to simulate data based on directed acyclic graphs. Int J Epidemiol 50(6):1772-1777.

Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic research. Epidemiology 10(1):37-48.

See Also

dag.adjustment, find.paths, eval.paths

dag.adjustment Adjust a DAG for one or more variables.

Description

Identifies the associations introduced by adjustment for the variables specified, and returns the DAG with these associations added. Note that this is called internally by dag.adjust, which makes sure that biasing paths are looked for and evaluted afterwards. Thus, dag.adjustment should 1.) *not* be called directly, and 2.) *not* be called on an already adjusted DAG!

Usage

```
dag.adjustment(dag, A=NULL)
```

Arguments

dag	The DAG to be adjusted.
A	The adjustment set to be applied.

Details

The adjustment set A specified when calling dag.adjustment overrules the adjustment variables that are present in the DAG. To keep these in the adjustment set, one has to add them to A.

Value

A DAG with A as the adjustment set and the associations introduced by adjustment for A added to the DAG.

Note

You should *not* use dag.adjustment on an already adjusted DAG, since it cannot identify associations that had been introduced by the earlier adjustment. If the new adjustment set does not include the adjustment variables present in the first set, the new DAG might feature associations that actually only would be introduced when adjusting for the variables featured in the first but not second adjustment set.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

See Also

dag.adjust, find.paths, eval.paths

dag.ancestors

Identify ancestors of DAG nodes.

Description

This identifies those nodes in a DAG that are ancestors of the nodes specified, i.e. acc. to the model depicted by the DAG they causally precede those nodes. Internally called by dag.adjustment in the context of finding associations introduced by adjustment.

Usage

```
dag.ancestors(dag, A)
```

Arguments

dag	The DAG to be evaluated.
A	A vector of nodes for which ancestors are to be identified.

Value

A vector indicating which nodes are ancestors of those in A. Note that A actually is included at the beginning of the vector.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

dag.adjust, dag.adjustment

12

dag.draw

Draw a DAG.

Description

Draws a DAG defined in an object of class dagRdag (as of dagR version 1.2.0, the generic function plot.dagRdag can be used for this purpose, but dag.draw is maintained for backwards compatibility). The nodes are represented by 'C' (covariables; numbered with subscripts) and 'U' (un-known/unmeasured covariables; numbered with subscripts), 'X' and 'Y' (exposure and outcome, respectively). A legend presents the names of the nodes. The X->Y arc is marked with a questionmark as the relationship of interest. Adjusted variables are under- and over-lined. Undirected associations are drawn with dashed lines. If paths have been identified (and evaluated), these (and their status) are written next to the legend.

Usage

dag.draw(dag, legend = TRUE, paths = TRUE, numbering = FALSE, p = FALSE, alt.symb = TRUE, noxy = 0, ...)

Arguments

dag	The DAG to be drawn.
legend	Boolean indicating whether a node legend should be included.
paths	Boolean indicating whether paths (and their status) should be written.
numbering	Boolean indicating whether the arcs should be numbered in the DAG.
р	Boolean indicating whether the curving points of undirected associations should be drawn.
alt.symb	Boolean indicating if the alternative node symbols (dag\$symbols) should be used. Note that especially the legends and paths will not be formatted nicely if these symbols are longer strings.
noxy	Integer to indicate if the X->Y should not be drawn (0=default; 1=no arc; 2=arc, but no question mark).
	Currently not used.

Value

Returns the DAG (for whatever reason...).

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Breitling LP, Duan C, Dragomir AD, Luta G (2022). Using dagR to identify minimal sufficient adjustment sets and to simulate data based on directed acyclic graphs. Int J Epidemiol 50(6):1772-1777.

Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic research. Epidemiology 10(1):37-48.

See Also

dag.letter, garrows, smoothArc, dag.legend, write.paths

t up a new DAG.

Description

Allows setting up a new DAG. See the demo.dag0 to demo.dag6 functions for some example specifications.

Usage

```
dag.init(outcome = NULL, exposure = NULL, covs = c(), arcs = c(),
assocs = c(), xgap = 0.04, ygap = 0.05, len = 0.1, y.name = NULL,
x.name = NULL, cov.names = c(), symbols = NULL, ...)
```

Arguments

outcome	Currently not used!
exposure	Currently not used!
covs	Vector including an integer for each covariable to be in the DAG (1 for a "stan- dard" covariable, 2 for an unknown/unmeasured one).
arcs	Vector of duplets of integers, in which nodes from which an arc or undirected association is to emenate are followed by those to which it is to point. To refer to the exposure, use 0, to refer to the outcome, use -1, to refer to covariables, use and element of 1:length(covs).
assocs	A vector of same length as covs, with 0 indicating directed arcs, 1 indicating undirected associations.
xgap	How much x space is to be left between arc ends and nodes when drawing?
ygap	How much y space is to be left between arc ends and nodes when drawing?
len	Length of arrow whiskers when drawing.
y.name	Label of outcome.

dag.init

x.name	Label of exposure.
cov.names	Vector of covariable labels.
symbols	Vector of alternative node symbols. Longer symbols will not be formatted nicely. Note that the first element refers to the exposure, the following ones to the covariables, the last one to the outcome.
	Currently not used.

Value

A DAG (objects of class dagRdag). Check out some of the demonstration DAGs for details. The DAG is actually a list object, with elements cov.types (the covs vector, with 0 put in front, and -1 at the end).; x and y (coordinates for drawing the nodes, initially set up more or less in a half-circle above the x->y arc); arc (the arcs, transformed into a matrix); arc.type (the assocs vector); curve.x and curve.y (if associations are featured, these provide the coordinates through which to curve); xgap, ygap, len (the respective drawing parameters); symbols (alternative node symbols); version (dagR version).

Note

CAVE: The numbering of the covariables and arc coordinates is different here than in the functions later used on the DAG (e.g. add.arc, dag.adjust)! The functions generally work according to the indexing of the R objects that they handle. Whereas for dag.init the n covariable nodes are numbered 1:n, the node vector of the resulting DAG will also contain the exposure node at the beginning and the outcome node at the end, i.e. it will go from 1: (n+2) with the covariables at 2:n+1. summary_dagRdag will show the latter numbering. *Example*: when adjusting for the first covariable, dag.adjust must be handed the adjustment set A=2, as the first covariable will occupy the second node (the first node is occupied by the exposure).

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic research. Epidemiology 10(1):37-48.

See Also

dag.draw

Examples

#dag.init(covs = c(1, 1), arcs = c(0, 2, 1, 2, 1, 0, -1, 2))

dag.legend

Description

Lists the DAG symbols along with their names/labels below a DAG drawn.

Usage

dag.legend(dag, lx = -0.15, ly = -0.075, alt.symb = TRUE)

Arguments

dag	The DAG for which the legend is needed.
lx	X coordinate for repositioning legend.
ly	Y coordinate for repositioning legend.
alt.symb	Boolean indicating if the alternative node symbols (dag\$symbols) should be used. Note that the formatting is not changed, i.e. longer symbols will not be formatted nicely.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

dag.draw,write.paths

dag.letter

Write a DAG node symbol.

Description

Writes the node symbols, 'X' and 'Y' for exposure and outcome, 'C' and 'U' (with consecutive subscripts) for known and unknown covariables. Since v1.1.2, alt.symb allows the use of custom node symbols. Unknownness is identified by either node name 'unknown' or covariable type '2' in the DAG object. Note that adjusted nodes are marked by bar and underline; this currently does not apply to those marked as unknown.

Usage

```
dag.letter(dag, letter, x, y, alt.symb = TRUE)
```

dag.letter2

Arguments

dag	The DAG for which a node is to be written.
letter	The node that is of interest.
х	X position.
У	Y position.
alt.symb	Boolean indicating if custom symbols (dag\$symbols) should be used if available.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

dag.draw,dag.legend,write.paths

dag.letter2

Return a DAG node symbol.

Description

Similar to dag.letter(), but returning a string to label a DAG node. Adjusted nodes are marked by a preceding underscore.

Usage

dag.letter2(dag, letter, alt.symb)

Arguments

dag	The dagRdag object for which a node symbol is to be returned.
letter	The number of the node for which the symbol (often a single letter) is to be returned.
alt.symb	If TRUE, the alternative node symbols of the DAG object will be used.

Value

A string containing the DAG letter or alternative symbol.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

dag.letter

dag.move

Description

This allows to reposition a node or association curving point of a DAG graphically. First, select a node or curving point by left-clicking close to it. Then reposition it to any other position by left-clicking. Once you are happy with the new position, right-click to exit.

Usage

dag.move(dag)

Arguments

dag The DAG to be modified.

Value

The same DAG, but with the feature repositioned.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

dag.search

Evaluate possible adjustment sets of a DAG.

Description

Currently, this simply is a wrapper for brute.search, which returns the input DAG with the results of brute.search and a string describing the search setup.

Usage

```
dag.search(dag, type = "brute", allow.unknown = FALSE, trace = FALSE, stop = 0)
```

Arguments

dag	DAG to be evaluated.
type	Type of search to be performed. Currently, only =brute is possible.
allow.unknown	See brute.search.
trace	See brute.search.
stop	See brute.search.

dag.sim

Value

The DAG with components searchType and searchRes added.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

See Also

brute.search

dag.sim

Simulate data based on a DAG.

Description

Simulates data according to a DAG object. This function may be replaced by dag.sim2 in the future.

Usage

Arguments

dag	The DAG object according to which data is to be simulated.
b	Vector of coefficients defining the direct effects of the DAG arcs.
bxy	Coefficient defining the direct effect of main exposure X on outcome Y.
n	Number of observations to be simulated.
mu	Vector of means that are to be simulated for the different DAG nodes. For binary nodes without an ancestor, the mean is taken as the prevalence to be simulated. For binary nodes with ancestors, the mean is similarly interpreted (see details in Value section).
binary	Vector indicating which nodes are to be continuous (=0) and binary (=1).
stdev	Vector of standard deviations for each node. For nodes without ancestors, con- tinuous data are drawn from a Normal distribution with this standard deviation. For nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation values.
naming	If =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named $X1Xn$.
seed	Seed to initialize the random number generator.
verbose	If =TRUE, additional output is given during the simulation, in particular show- ing the different calculation steps.

A dataframe with n (rows) observations featuring simulated data for each node (columns) in the DAG. Simulation steps: 1. simulate data for nodes i without ancestors, drawing from Normal distribution with mean mu[i] and stdev[i] (continuous node), or drawing from Bernoulli events with probability mu[i] (binary node). 2. simulate data for nodes i for which all ancestors already have been simulated by multiplying the ancestor values with the corresponding arc coefficients and summing them up, shifting the resulting values to the mean mu[i] specified for the currently simulated node (logit-transformed if binary), then adding noise drawn from a Normal distribution with mean 0 and standard deviation stdev[i], finally using the inverse logit of the resulting values as success probabilities for simulating binary data if node is binary.

Note

Undirected arcs are ignored in these simulations!

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP, Duan C, Dragomir AD, Luta G (2022). Using dagR to identify minimal sufficient adjustment sets and to simulate data based on directed acyclic graphs. Int J Epidemiol 50(6):1772-1777.

Duan C, Dragomir AD, Luta G, Breitling LP (2022). Reflection on modern methods: Understanding bias and data analytical strategies through DAG-based data simulations. Int J Epidemiol 50(6):2091-2097.

See Also

dag.sim2

dag.sim2

Simulate data based on a DAG.

Description

Simulates data according to a DAG object.

Usage

20

Value

dag.sim2

lambda = NA, binary = NA, naming = 2, seed = NA, verbose = FALSE)

Arguments

scale). bxy Coefficient defining the direct effect of main exposure X on outcome Y (or linear scale). n Number of observations to be simulated. distr 0 for Normal distribution continuous nodes, 1 for binary nodes simulated from logistic model, 1.1 for binary nodes simulated from logistic model, 2.1 for binary nodes simulated from linear risk difference model, 2.1 for binary nodes simulated from linear risk difference model, 2.1 for binary nodes simulated from linear risk difference model, 2.1 for binary nodes simulated for the different DAG nodes: For normally distributed continuous variables, the overall mean simulated. For binary variables w/ distr=1 or distr=2, overall proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For nodes without ancestors, continuous data are drawn from a Normal distribution with this standard deviation. For continuous nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation value. If used on binary variables with ancestors, this would analogously add residua noise to the calculated predictor, diluting the direct effects. nu Not used. lambda Not used. binary For backwards compatibility: Vector indicating which nodes are to be contin- uous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued. naming If=2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1Xn		
scale). bxy Coefficient defining the direct effect of main exposure X on outcome Y (or linear scale). n Number of observations to be simulated. distr 0 for Normal distribution continuous nodes, 1 for binary nodes simulated from logistic model, 1.1 for binary nodes simulated from linear risk difference model, 2.1 for binary nodes simulated from linear risk difference model (see mu) mu Vector of means that are to be simulated for the different DAG nodes: For normally distributed continuous variables, the overall mean simulated. For binary variables w/ distr=1 or distr=2, overall proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For nodes without ancestors, continuous data are drawn from a Normal distribution with this standard deviation. For continuous nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation value. If used on binary variables with ancestors, this would analogously add residua noise to the calculated predictor, diluting the direct effects. nu Not used. lambda Not used. binary For backwards compatibility: Vector indicating which nodes are to be continuous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued. naming If =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1Xn seed Seed to inini	dag	The DAG object according to which data is to be simulated.
n Number of observations to be simulated. distr 0 for Normal distribution continuous nodes, 1 for binary nodes simulated from logistic model (see mu), 2 for binary nodes simulated from logistic model (see mu), 2 for binary nodes simulated from linear risk difference model (see mu) mu Vector of means that are to be simulated for the difference model (see mu) For normally distributed continuous variables, the overall mean simulated. For binary variables w/ distr=1 or distr=2, overall proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For normally distributed continuous data are drawn from a Normal distribution with this standard deviation. For continuous nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation value. If used on binary variables with ancestors, this would analogously add residua noise to the calculated predictor, diluting the direct effects. nu Not used. lambda Not used. lambda Not used. lambda Not used. naming If =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1Xn seed Seed to initialize the random number generator. verbose If =TRUE, additional output is given during the simulation, in particular show-	b	Vector of coefficients defining the direct effects of the DAG arcs (on linear scale).
distr0 for Normal distribution continuous nodes, 1 for binary nodes simulated from logistic model, 1.1 for binary nodes simulated from linear risk difference model, 2.1 for binary nodes simulated from linear risk difference model (see mu)muVector of means that are to be simulated for the different DAG nodes: For normally distributed continuous variables, the overall mean simulated. For binary variables w/ distr=1 or distr=2, overall proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For nodes without ancestors, continuous data are drawn from a Normal distribution with this standard deviation. For continuous nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation value. If used on binary variables with ancestors, this would analogously add residual noise to the calculated predictor, diluting the direct effects.nuNot used.lambdaNot used.binaryFor backwards compatibility: Vector indicating which nodes are to be continuous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued.namingIf =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1XnseedSeed to initialize the random number generator.If given during the simulation, in particular show-	bxy	Coefficient defining the direct effect of main exposure X on outcome Y (on linear scale).
1 for binary nodes simulated from logistic model, 1.1 for binary nodes simulated from logistic model (see mu), 2 for binary nodes simulated from linear risk difference model, 2.1 for binary nodes simulated from linear risk difference model (see mu) mu Vector of means that are to be simulated for the different DAG nodes: For normally distributed continuous variables, the overall mean simulated. For binary variables w/ distr=1 or distr=2, overall proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated. For nodes without ancestors, continuous data are drawn from a Normal distribution with this standard deviation. For continuous nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation value. If used on binary variables with ancestors, this would analogously add residuat noise to the calculated predictor, diluting the direct effects. nu Not used. lambda Not used. binary For backwards compatibility: Vector indicating which nodes are to be continuous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued. naming If =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1Xn seed <td>n</td> <td>Number of observations to be simulated.</td>	n	Number of observations to be simulated.
For normally distributed continuous variables, the overall mean simulated. For binary variables w/ distr=1 or distr=2, overall proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated in the reference category (sum of coef-weighted predictors =0).stdevVector of standard deviations for each node. For nodes without ancestors, continuous data are drawn from a Normal distribution with this standard deviation. For continuous nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation value. If used on binary variables with ancestors, this would analogously add residua noise to the calculated predictor, diluting the direct effects.nuNot used.lambdaNot used.binaryFor backwards compatibility: Vector indicating which nodes are to be continuous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued.namingIf =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1XnseedSeed to initialize the random number generator.verboseIf =TRUE, additional output is given during the simulation, in particular show-	distr	 for binary nodes simulated from logistic model, for binary nodes simulated from logistic model (see mu), for binary nodes simulated from linear risk difference model,
stdevVector of standard deviations for each node. For nodes without ancestors, continuous data are drawn from a Normal distribution with this standard deviation. For continuous nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation value. If used on binary variables with ancestors, this would analogously add residual noise to the calculated predictor, diluting the direct effects.nuNot used.lambdaNot used.binaryFor backwards compatibility: Vector indicating which nodes are to be continuous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued.namingIf =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1XnseedSeed to initialize the random number generator.verboseIf =TRUE, additional output is given during the simulation, in particular show-	mu	For normally distributed continuous variables, the overall mean simulated. For binary variables w/ distr=1 or distr=2, overall proportion of successes simulated. For binary variables w/ distr=1.1 or distr=2.1, proportion of successes simulated
lambdaNot used.binaryFor backwards compatibility: Vector indicating which nodes are to be continuous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued.namingIf =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1XnseedSeed to initialize the random number generator.verboseIf =TRUE, additional output is given during the simulation, in particular show-	stdev	Vector of standard deviations for each node. For nodes without ancestors, continuous data are drawn from a Normal distribution with this standard deviation. For continuous nodes with ancestors, this is the standard deviation of the residual noise that is added to the calculated observation value. If used on binary variables with ancestors, this would analogously add residual
binaryFor backwards compatibility: Vector indicating which nodes are to be continuous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued.namingIf =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1XnseedSeed to initialize the random number generator.verboseIf =TRUE, additional output is given during the simulation, in particular show-	nu	Not used.
uous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued.namingIf =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named X1XnseedSeed to initialize the random number generator.verboseIf =TRUE, additional output is given during the simulation, in particular show-	lambda	Not used.
output dataframe. Otherwise, the output dataframe variables are named X1XnseedSeed to initialize the random number generator.verboseIf =TRUE, additional output is given during the simulation, in particular show-	binary	For backwards compatibility: Vector indicating which nodes are to be continuous (=0) and binary (=1). If given, this is passed to argument "distr" and a warning is issued.
verbose If =TRUE, additional output is given during the simulation, in particular show-	naming	If =2, the alternative DAG node symbols are used for naming the variables in the output dataframe. Otherwise, the output dataframe variables are named $X1Xn$.
	seed	Seed to initialize the random number generator.
	verbose	If =TRUE, additional output is given during the simulation, in particular show- ing the different calculation steps.

Value

A dataframe with n (rows) observations featuring simulated data for each node (columns) in the DAG.

Simulation steps:

1. simulate data for nodes i without ancestors, drawing from Normal distribution with mean mu[i] and stdev[i] (continuous node), or drawing from Bernoulli events with probability mu[i] (binary node).

2. simulate data for nodes i for which all ancestors already have been simulated by multiplying the ancestor values with the corresponding arc coefficients and summing them up, shifting the resulting values to the mean mu[i] (exceptions: distr=1.1 or distr=2.1, as detailed in "mu" above) specified for the currently simulated node (logit-transformed if binary based on logistic model), then adding noise drawn from a Normal distribution with mean 0 and standard deviation stdev[i], finally using the resulting values (inverse logit, if binary based on logistic model) as success probabilities for simulating binary data if node is binary.

As the noise is added after shifting to the mean, the mean of the simulated data will not be exact. Also, the noise is added before calculating descendant nodes, i.e. it is sort of true inter-individual variation, rather than measurement error.

For the risk difference model, the success probability calculated by summing the weighted ancestors can easily be <0 (or >1). If this happens, the probability is set to 0 (or 1), and a warning is issued.

Note

Undirected arcs are ignored in these simulations.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP, Duan C, Dragomir AD, Luta G (2022). Using dagR to identify minimal sufficient adjustment sets and to simulate data based on directed acyclic graphs. Int J Epidemiol 50(6):1772-1777.

Duan C, Dragomir AD, Luta G, Breitling LP (2022). Reflection on modern methods: Understanding bias and data analytical strategies through DAG-based data simulations. Int J Epidemiol 50(6):2091-2097.

See Also

dag.sim

dagR2dagitty

Description

Translates a DAG as defined in a dagRdag object created by dagR into the dagitty package format. Node labeling follows the rules used for plotting dagRdag objects, but adjusted nodes are marked by a preceding underscore instead of under- and over-line.

Usage

dagR2dagitty(x, alt.symb = TRUE, only.code = TRUE)

Arguments

x	The dagR DAG to be translated.
alt.symb	Boolean indicating if the alternative node symbols should be used.
only.code	If TRUE, a string with R dagitty function call is returned, which should be checked by the user (and possibly edited as required) before running it to create an equivalent dagitty DAG. If FALSE and the dagitty package has been installed and loaded, the dagitty function is called directly and the resulting dagitty DAG is returned.

Value

Either a string containing dagitty syntax to translate the dagR DAG into dagitty format, or a dagitty object.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Breitling LP, Duan C, Dragomir AD, Luta G (2022). Using dagR to identify minimal sufficient adjustment sets and to simulate data based on directed acyclic graphs. Int J Epidemiol 50(6):1772-1777.

https://cran.r-project.org/package=dagitty

See Also

dag.letter2

Description

Initializes a simple DAG used during the dagR development phase.

Usage

demo.dag0()

Value

Returns a DAG.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

demo.dag1, demo.dag2, demo.dag3, demo.dag4, demo.dag5, demo.dag6

demo.dag1

Set up demo DAG #1.

Description

Initializes a classical "M DAG" useful for demonstrating harmful adjustment. The DAG is motivated by figure 3 in Fleischer (2008) and also featured in Breitling (2010).

Usage

demo.dag1()

Value

Returns a DAG.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Fleischer NL, Diez Roux AV (2008). Using directed acyclic graphs to guide analyses of neighbourhood health effects: an introduction. J Epidemiol Community Health 62:842-846.

See Also

demo.dag0, demo.dag2, demo.dag3, demo.dag4, demo.dag5, demo.dag6

demo.dag2

Set up demo DAG #2.

Description

Initializes a more complex DAG, motivated by Shrier (2008). This DAG was used to examine the performance of brute.search and has been featured in Breitling (2010).

Usage

demo.dag2()

Value

Returns a DAG.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587. Shrier I, Platt RW (2008). Reducing bias through directed acyclic graphs. BMC Med Res Methodol 8:70

See Also

demo.dag0, demo.dag1, demo.dag3, demo.dag4, demo.dag5, demo.dag6

Description

Initializes a DAG motivated by the manual for the software DAG v0.11 (Knüppel 2009). This DAG has been featured in Breitling (2010).

Usage

demo.dag3()

Value

Returns a DAG.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587. Knüppel S (2009). DAG v0.11 documentation (Oct 21, 2009). https://hsz.dife.de/dag/

See Also

demo.dag0, demo.dag1, demo.dag2, demo.dag4, demo.dag5, demo.dag6

demo.dag4 Set up demo DAG #4.

Description

Initializes a miscellaneous DAG. What happens if you adjust for the exposure's child?

Usage

```
demo.dag4()
```

Value

Returns a DAG.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

See Also

demo.dag0, demo.dag1, demo.dag2, demo.dag3, demo.dag5, demo.dag6

demo.dag5

Set up demo DAG #5.

Description

Initializes a miscellaneous DAG. What happens if you adjust for the outcome's child?

Usage

demo.dag5()

Value

Returns a DAG.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

demo.dag0, demo.dag1, demo.dag2, demo.dag3, demo.dag4, demo.dag6

demo.dag6

Set up demo DAG #6.

Description

Initializes a miscellaneous DAG. What happens if you adjust for the collider?

Usage

demo.dag6()

Value

Returns a DAG.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

demo.dag0, demo.dag1, demo.dag2, demo.dag3, demo.dag4, demo.dag5

demo.dag7

Set up demo DAG #7.

Description

Initializes a DAG motivated by the manual for the software DAG v0.11 (Kn\"uppel 2009). This DAG has been featured in Breitling (2010). The DAG is the same as DAG #3, but #7 demonstrates the use of alternative node symbols.

Usage

demo.dag7()

Value

Returns a DAG.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587. Kn\"uppel S (2009). DAG v0.11 documentation (Oct 21, 2009). https://hsz.dife.de/dag/

See Also

demo.dag3

distPoints

Description

Another rather superfluous helper function, internally used by smoothArc. Calculates the distance between two points.

Usage

distPoints(A, B)

Arguments

А	Vector of length two, indicating x and y of first point.
В	Vector of length two, indicating x and y of second point.

Value

Distance between the two points.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

eval.paths

Evaluate potentially biasing paths in a DAG.

Description

This essentially implements the graphical algorithm described in Greenland (1999) to identify open "backdoor" (or not strictly backdoor, but potentially biasing) paths in a DAG. Paths are identified as being 'open', 'blocked by collider', or 'blocked by adjustment'. If both latter conditions apply, 'blocked by collider' is returned.

Usage

```
eval.paths(dag)
```

Arguments

dag

A DAG to which find.paths has already been applied (e.g. within dag.adjust).

Details

This function identifies a collider-blocked path as 'blocked by collider' even if it has been unblocked by adjusting for the collider. One could argue that this should not be the case. However, the biasing seems to be sufficiently represented in the DAG by the introduction of the association "jumping" the collider and potentially opening biasing paths.

Value

A DAG with component path.status added.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic research. Epidemiology 10(1):37-48.

See Also

dag.adjust, find.paths

find.paths	Find potentially biasing paths in a DAG.	
------------	--	--

Description

This identifies paths linking exposure and outcome in a DAG. Forward paths (including a directed arc emanating from the exposure) are *not* identified.

Usage

find.paths(dag)

Arguments

dag A DAG for which paths should be found.

Value

A DAG with components pathsN (number of paths identified) and paths (matrix with each row describing one path by indicating the arcs forming the path; ends with NA as some other function recognize the end of the path that way) added.

30

garrows

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587. Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic research. Epidemiol-

ogy 10(1):37-48.

See Also

dag.adjust, eval.paths

garrows

Draw a directed arc in a DAG.

Description

Internally called by dag.draw for drawing directed arcs.

Usage

garrows(x0, y0, x1, y1, xgap, ygap, len = 0.1)

Arguments

x0	X coordinate of origin.
уØ	Y coordinate of origin.
x1	X coordinate of target node.
y1	Y coordinate of target node.
xgap	Space between node and arc ends on x axis.
ygap	Space between node and arc ends on y axis.
len	Length of arrow whiskers (default=0.1).

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

See Also

dag.draw, smoothArc

inAngle

Description

Another rather superfluous helper function, calculating the radian angle between two radian angles. Internally called by smoothArc.

Usage

inAngle(a, b)

Arguments

а	Radian angle 1.
b	Radian angle 2.

Value

Numeric in range from -pi to pi.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

smoothArc

is.acyclic

Check if a DAG actually is acyclic.

Description

This function checks for each node in a DAG whether backtracing arcs leading to it results in an "infinite recursion" error indicating that there actually is a cyclic part in the DAG (which then obviously seems not to be a DAG).

Usage

is.acyclic(dag, maxSecs=NA)

Arguments

dag	The DAG to be check.
maxSecs	maximum time before function aborts;

is.in

Value

A list with two elements. acyclic is a boolean indicating whether the DAG is acyclic (=TRUE) or contains a cyclic component (=FALSE). nodewise is a vector containing 1 boolean per node in the DAG, TRUE indicating that backtracing from this node does not lead to a cyclic component, FALSE indicating that backtracing from this node leads to a cyclic component.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

is.in

Check if a specific numeric value occurs in a vector.

Description

Another trivial helper function, called internally by eval.paths. It checks whether the specified (numeric) value is part of a specified vector of (numeric) values.

Usage

is.in(x, c = NULL)

Arguments

Х	A numeric value, for which the presence in a vector is to be checked.
с	A vector of numeric values.

Value

Boolean; TRUE if value is present, FALSE if not.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

eval.paths

is.unknown

Description

Another helper function, internally used by brute.search. It checks whether the node specified is of type=2 or is named 'unknown'.

Usage

is.unknown(x, dag)

Arguments

х	The node of interest.
dag	The DAG to be evaluated.

Value

TRUE if unknown (acc. to type or name), FALSE otherwise.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

See Also

brute.search

msas

Identify minimal sufficient adjustment sets.

Description

Evaluates DAG adjustment sets identified by a dag.search (or brute.search) for minimal sufficiency by counting for each sufficient adjustment set A how many smaller sufficient ones that are contained in A exist.

Usage

```
msas(adjSets)
```

Arguments

adjSets The searchRes component of a DAG (or the output of brute.search, which is used by dag.adjust to produce searchRes).

plot.dagRdag

Value

A vector containing a -1 for each insufficient adjustment set, and for sufficient ones the number of smaller sufficient ones contained in it.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Breitling LP, Duan C, Dragomir AD, Luta G (2022). Using dagR to identify minimal sufficient adjustment sets and to simulate data based on directed acyclic graphs. Int J Epidemiol 50(6):1772-1777.

Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic research. Epidemiology 10(1):37-48.

Knüppel S, Stang A (2010). DAG Program: identifying minimal sufficient adjustment sets. Epidemiology 21(1):159.

See Also

viv, summary.dagRdag

plot.dagRdag Function to draw a DAG

Description

Generic function to draw a directed acyclic graph in an object of class dagRdag. This essentially passes the DAG object to the function dag.draw, which is maintained for backwards compatibility.

Usage

S3 method for class 'dagRdag'
plot(x, y, ...)

Arguments

Х	Object of class dagRdag to be passed to dag.draw.
У	Currently not used.
	Other arguments to be passed to dag.draw.

Details

For all available arguments, see documentation of dag.draw.

Value

The DAG object is returned.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Breitling LP, Duan C, Dragomir AD, Luta G (2022). Using dagR to identify minimal sufficient adjustment sets and to simulate data based on directed acyclic graphs. Int J Epidemiol 50(6):1772-1777.

See Also

dag.draw

print.dagRdag

Prints the raw contents of an object of class dagRdag.

Description

Generic function print code for class dagRdag. This uses the default print method for list objects and points the user to the availability of the more convenient summary method.

Usage

S3 method for class 'dagRdag'
print(x, ...)

Arguments

х	An object of class dagRdag.
	Other arguments passed to the print routine.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

36

rm.arc

Description

Conveniently remove an arc from an existing DAG.

Usage

rm.arc(dag, arc)

Arguments

dag	The DAG from which to remove the arc.
arc	A single integer, indicating which arc is to be removed (refering to the respective
	row of the dag\$arc matrix).

Value

A DAG with the arc specified removed along with the corresponding attributes like arc types, curves, and path evaluation variables.

Note

The numbering of the arcs can be visualized by applying dag.draw with the option "number-ing=TRUE".

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

add.arc,add.node,rm.node

rm.node

Remove a node from a DAG.

Description

Conveniently remove a node from an existing DAG.

Usage

rm.node(dag, node)

Arguments

dag	The DAG from which to remove the node.
node	A single integer, indicating which node is to be removed.

Value

A DAG with the node specified removed, along with the corresponding attributes and dependent variables, i.e. arcs involving this node are also removed, and the numbering of the nodes (and their occurrence in arcs) is corrected accordingly.

Note: Search components (searchType, searchRes) of the DAG currently are generally set to NULL, even if no path is removed. This is for simplicity, because the node numbers would need to be changed eg. in the searchRes variables etc.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

add.node, rm.arc, add.arc

smoothArc

Draw an undirected assocation in a DAG.

Description

This draws a dashed connection between two points, curving it so that it goes through a third point. This is internally used by dag.draw to draw associations.

Usage

smoothArc(A, B, C, res = 20, gap = 0.05, p = FALSE)

Arguments

A	Vector of length 2, providing xy coordinates of first point.
В	Vector of length 2, providing xy coordinates of second point.
С	Vector of length 2, indicating xy coordinates through which the association should be curved.
res	How smooth should the curve be drawn?
gap	How far from point A and B should the line end?
р	If TRUE, the point through which the curve goes is drawn (this is to allow better moving it with dag.move).

Note

In the version 1.0.1 distributed as online supplemental material with Breitling (2010), the function contains arbitrary default values used during development.

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

See Also

dag.draw,dag.move

summary.dagRdag Summarize a DAG.

Description

Generic function summary() for class dagRdag.

Usage

S3 method for class 'dagRdag'
summary(object, ...)

Arguments

object	An object of class dagRdag.
	Currently not used.

Details

Summarizes according to what functions have been applied to the DAG. It does not itself call dag.search and the like. Exception: is calls is.acyclic (with maxSecs=5). This function passes the object to summary_dagRdag, which is preserved for backwards compati-

Author(s)

bility.

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic research. Epidemiology 10(1):37-48.

Kn\"uppel S, Stang A (2010). DAG Program: identifying minimal sufficient adjustment sets. Epidemiology 21(1):159.

summary_dagRdag Summarize a DAG.

Description

Generic function summary() working code for class dagRdag, which is used by package dagR from version 1.1.1 on. From version 1.2.0, summary.dagRdag() is available as a generic function, but summary_dagRdag is preserved for backwards compatibility.

Usage

summary_dagRdag(dag)

Arguments

dag

An object of class dagRdag.

Details

Summarizes according to what functions have been applied to the DAG. It does not itself call dag.search and the like. Exception: is calls is.acyclic (with maxSecs=5).

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

References

Breitling LP (2010). dagR: a suite of R functions for directed acyclic graphs. Epidemiology 21(4):586-587.

Greenland S, Pearl J, Robins JM (1999). Causal diagrams for epidemiologic reserach. Epidemiology 10(1):37-48.

Kn\"uppel S, Stang A (2010). DAG Program: identifying minimal sufficient adjustment sets. Epidemiology 21(1):159.

40

Description

Checks if all numeric elements of a vector occur also in another vector. It is internally used by msas to check if some adjustment set is contained in another one.

41

Usage

viv(v1, v2)

Arguments

v1	The vector whose occurrence in v2 is to be checked.
v2	The vector in which v1 might occur.

Details

If a value occurs more than once in v1, it is counted as contained in v2 if it appears there once. An empty v1 (consisting only of NA) is considered to be contained in any v2.

Value

TRUE if v1 occurs in v2, FALSE otherwise.

Author(s)

Lutz P Breitling <1.breitling@posteo.de>

See Also

msas

write.paths

Write the paths into a DAG drawing.

Description

Writes the paths into a DAG drawing, using the symbols ('C', 'U', 'X', 'Y') used in the drawing, indicating directed arcs by '<' and '>', undirected ones by '-'. Since version 1.1.2, alt.symb allow usage of custom node symbols, though multi-character symbols will not be formatted well. Adjusted variables are under- and over-lined. If the paths have been evaluated using eval.paths, the status are also written.

viv

Usage

write.paths(dag, px = 0.5, py = -0.06, alt.symb = TRUE)

Arguments

be used.
ł

Author(s)

Lutz P Breitling <l.breitling@posteo.de>

See Also

dag.draw,find.paths,eval.paths,dag.legend

Index

* package dagR-package, 2 add.arc, 4, 5, 37, 38 add.node, 4, 5, 37, 38 addAngle, 6, 7 allCombs, 6 angle,7 anglePoint, 8 assoc.exists, 8 brute.search, 3, 7, 9, 18, 19, 34 dag.adjust, 3, 10, 12, 30, 31 dag.adjustment, 8, 11, 11, 12 dag.ancestors, 12dag.draw, 3, 13, 15-17, 31, 36, 39, 42 dag.init, *3*, 14 dag.legend, 14, 16, 17, 42 dag.letter, *14*, 16, *17* dag.letter2, 17, 23 dag.move, 18, 39 dag.search, 3, 18 dag.sim, 3, 19, 22 dag.sim2, 20, 20 dagR (dagR-package), 2 dagR-package, 2 dagR2dagitty, 23 demo.dag0, 24demo.dag1, 24demo.dag2, 25demo.dag3, 26demo.dag4, 26demo.dag5, 27demo.dag6, 27demo.dag7, 28distPoints, 29 eval.paths, 11, 12, 29, 31, 33, 42 find.paths, 11, 12, 30, 30, 42

garrows, 14, 31
inAngle, 32
is.acyclic, 32
is.in, 33
is.unknown, 34
msas, 3, 34, 41
plot.dagRdag, 35
print.dagRdag, 36
rm.arc, 4, 5, 37, 38
rm.node, 4, 5, 37, 37
smoothArc, 6-8, 14, 31, 32, 38
summary_dagRdag, 35, 39
summary_dagRdag, 40
viv, 35, 41
write.paths, 14, 16, 17, 41