# Package 'SID'

December 6, 2023

Type Package

Title Structural Intervention Distance
Version 1.1
<b>Date</b> 2023-12-03
Encoding UTF-8
Imports pealg, igraph, RBGL, Matrix, methods
<b>Description</b> The code computes the structural intervention distance (SID) between a true directed acyclic graph (DAG) and an estimated DAG. Definition and details about the implementation can be found in J. Peters and P. Bühlmann: "Structural intervention distance (SID) for evaluating causal graphs", Neural Computation 27, pages 771-799, 2015 <doi:10.1162 neco_a_00708="">.</doi:10.1162>
<pre>URL https://github.com/fkgruber/SID_cran</pre>
License FreeBSD
Repository CRAN
NeedsCompilation no
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<b>Date/Publication</b> 2023-12-06 10:10:02 UTC
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SID-package

Structural Intervention Distance (SID) (package)

### **Description**

The package allows to compute the structural intervention distance (SID) between two graphs. This (pre-)metric evaluates graphs from a causal inference point of view. The package also contains a function for computing the Hamming distance and to generate random DAGs.

#### **Details**

Package: SID
Type: Package
Version: 1.0
Date: 2015-03

Date: 2015-03-07 License: FreeBSD

Computing the Structural Intervention Distance. The algorithm can be called with structIntervDist.

#### Author(s)

Jonas Peters <jonas.peters@tuebingen.mpg.de>

#### References

J. Peters, P. B\"uhlmann: Structural intervention distance (SID) for evaluating causal graphs, Neural Computation 27, pages 771-799, 2015

#### See Also

randomDAG structIntervDist hammingDist

computePathMatrix

auxiliary file for SID: computes a path matrix efficiently (can probably be made faster)

# **Description**

auxiliary file for SID: computes a path matrix efficiently (can probably be made faster) This function takes an adjacency matrix G from a DAG and computes a path matrix for which entry (i,j) being one means that there is a directed path from i to j the diagonal will also be one.

#### Usage

computePathMatrix(G, spars=FALSE)

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#### **Arguments**

G graph.

spars boolean indicating whether G is a sparse matrix.

#### Value

pathMatrix

#### Author(s)

Jonas Peters <jonas.peters@tuebingen.mpg.de>

#### See Also

structIntervDist computePathMatrix2

computePathMatrix2 auxiliary file for SID: computes a path matrix efficiently (can probably

be made faster)

# **Description**

auxiliary file for SID: computes a path matrix efficiently (can probably be made faster) The only difference to the function computePathMatrix is that this function changes the graph by removing all edges that leave condSet. If condSet is empty, it just returns PathMatrix1.

# Usage

```
computePathMatrix2(G,condSet,PathMatrix1, spars=FALSE)
```

# **Arguments**

G graph.

condSet set of variables
PathMatrix1 PathMatrix1

spars boolean indicating whether the matrices are sparse

#### Value

pathMatrix

# Author(s)

Jonas Peters <jonas.peters@tuebingen.mpg.de>

#### See Also

structIntervDist computePathMatrix

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Hamming Distance

#### **Description**

Computes the Hamming distance between two graph objects.

#### Usage

```
hammingDist(G1, G2, allMistakesOne = TRUE)
```

#### **Arguments**

G1 p x p adjacency matrix of the first graph.

G2 p x p adjacency matrix of the second graph.

allMistakesOne boolean variable (TRUE/FALSE) that indicates whether all mistakes should be

counted as one. E.g., if it is set to FALSE the Hamming distance between X ->

Y and  $X \leftarrow Y$  is two and one if it set to TRUE.

#### **Details**

The Hamming distance between two graphs counts the number of edges, in which the graphs do not coincide. allMistakesOne determines whether a reversed edge counts as two or as one mistake. The Hamming distance is symmetric, that is hammingDist(G1,G2) = hammingDist(G2,G1).

#### Value

The method outputs the computed Hamming distance.

#### Author(s)

Jonas Peters <jonas.peters@tuebingen.mpg.de>

#### See Also

structIntervDist

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randomDAG	outputs the adjacency matrix of a randomly generated directed acyclic graph (DAG).
randomDAG	

# **Description**

After simulating a random topological order first (the order can also be prespecified), the algorithm connects two nodes with the porbability probConnect.

#### Usage

```
randomDAG(p, probConnect, causalOrder = sample(p, p, replace = FALSE))
```

# Arguments

p	number of nodes.
probConnect	probability of connecting two nodes, determines the sparsity of the graph. Choosing probConnect = $2/(p-1)$ , for example, leads to an expected number of p nodes.
causalOrder	OPTIONAL: causal or topological order of the nodes. If not provided, the topo-

logical order is chosen randomly.

# Value

 $p \times p$  adjacency matrix that describes a directed acyclic graph (DAG) with p nodes. The entry (i,j) equals one if and only if there is an edge from i to j.

# Author(s)

Jonas Peters <jonas.peters@tuebingen.mpg.de>

#### See Also

```
structIntervDist hammingDist
```

# **Examples**

```
randomDAG(p = 5, probConnect = 0.6)
```

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structIntervDist	Structural Intervention Distance (SID)
structIntervDist	Structural Intervention Distance (SIL

#### **Description**

computes the structural intervention distance (SID) between two graphs; it evaluates graphs from a causal inference point of view.

#### Usage

```
structIntervDist(trueGraph, estGraph, output = FALSE, spars = FALSE)
```

#### **Arguments**

trueGraph p x p adjacecncy matrix; it must be a directed acyclic graph (DAG)

estGraph again a p x p adjacecncy matrix; it can be either a DAG or a completed partially directed graph (CPDAG) that represents a Markov equivalence class.

output boolean (TRUE/FALSE) that indicates whether output shall be shown boolean (TRUE/FALSE) that indicates whether the input matrices trueGraph and estGraph are sparse. If matrices are indeed large and sparse this increases the speed of the algorithm.

# Details

The structural intervention distance (SID) considers each pair (i,j) and checks whether the parents of i in estGraph is a valid adjustment set in trueGraph for the causal effect from i to j. If it is, estimating the causal effect from i to j using parent adjustment in estGraph is a correct procedure, independently of the distribution. If it is not, the pair (i,j) contributes a mistake of one to the overall structural intervention distance.

The SID satisfies the properties of a pre-metric; we have SID(G,G) = 0 and  $0 \le SID(G,G) \le p(p-1)$ , where p is the number of nodes. The SID is not symmetric: in general, we have  $SID(G,G) \le SID(G,G)$ . Furthermore, the SID can be zero although the two graphs are not the same.

If estGraph is a completed partially directed acyclic graph (CPDAG), that is it describes a Markov equivalence class, the SID outputs a lower and an upper bound of the SID, that correspond to the best and worst DAG in the equivalence class, respectively.

#### Value

The main function is structIntervDist. It takes two DAGs (adjacency matrices) of the same dimension as input and provides a list with

sid	the SID
sidUpperBound	relevant if estGraph is a CPDAG: it is the highest SID that can be achieved by a graph within the Markov equivalence class ("worst DAG")
sidLowerBound	relevant if estGraph is a CPDAG: it is the lowest SID that can be achieved by a graph within the Markov equivalence class ("best DAG")
incorrectMat	the matrix of incorrect interventional distributions; its sum equals \$sid

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#### Author(s)

Jonas Peters <jonas.peters@tuebingen.mpg.de>

#### References

J. Peters, P. B\"uhlmann: Structural intervention distance (SID) for evaluating causal graphs, Neural Computation 27, pages 771-799, 2015

#### See Also

hammingDist randomDAG

#### **Examples**

```
G \leftarrow rbind(c(0,1,1,1,1),
           c(0,0,1,1,1),
           c(0,0,0,0,0),
           c(0,0,0,0,0),
           c(0,0,0,0,0)
H1 \leftarrow rbind(c(0,1,1,1,1))
            c(0,0,1,1,1),
            c(0,0,0,1,0),
            c(0,0,0,0,0),
            c(0,0,0,0,0)
H2 \leftarrow rbind(c(0,0,1,1,1))
            c(1,0,1,1,1),
            c(0,0,0,0,0),
            c(0,0,0,0,0),
            c(0,0,0,0,0)
cat("true DAG G:\n")
show(G)
cat("\n")
cat("=======","\n")
cat("\n")
cat("estimated DAG H1:\n")
show(H1)
sid <- structIntervDist(G,H1)</pre>
shd <- hammingDist(G,H1)</pre>
cat("SHD between G and H1: ",shd,"\n")
cat("SID between G and H1: ",sid$sid,"\n")
#cat("The matrix of incorrect interventional distributions is: ","\n")
#show(sid$incorrectMat)
cat("\n")
cat("=======","\n")
cat("\n")
cat("estimated DAG H2:\n")
show(H2)
sid <- structIntervDist(G,H2)</pre>
shd <- hammingDist(G,H2)</pre>
```

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```
cat("SHD between G and H2: ",shd,"\n")
cat("SID between G and H2: ",sid$sid,"\n")
cat("The matrix of incorrect interventional distributions is: ","\n")
show(sid$incorrectMat)

readline("The SID can also be applied to CPDAGs. Please press enter...")
cat("\n")
cat("\n")
cat("e=======","\n")
cat("\n")
cat("estimated CPDAG H1c:\n")
H1c <- rbind(c(0,1,1,1,1),c(1,0,1,1),c(1,1,0,1,0),c(1,1,1,0,0),c(1,1,0,0)))
show(H1c)
sid <- structIntervDist(G,H1c)
cat("SID between G and H1: \n")
cat("lower bound: ",sid$sidLowerBound," upper bound: ", sid$sidUpperBound, "\n")
cat("\n")</pre>
```

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