Package 'BNDataGenerator'

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 BNDataGenerator-package

Data Generator based on Bayesian Network Model

Description

Data Generator based on Bayesian Network Model

Details

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Package: BN_Data_Generator

Type: Package Version: 1.0

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Author(s)

References

Jae-seong Yoo, (2014), "A Study on Comparison of Bayesian Network Structure Learning Algorithms for Selecting Appropriate Models", M.S. thesis, Department of Statistics, Korea University, Seoul.

See Also

See demo(gen_asia) or demo(topology).

big_letters

Letters built

Description

The 26 or more lower-case letters of the Roman alphabet;

Usage

```
big_letters(len)
```

Arguments

len

Get or set the length of letters.

Author(s)

Examples

```
big_letters(26)
big_letters(50)
```

big_letters(100)

big_letters(999)

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BN_Data_Generator Data Generator based on Bayesian Network Model

Description

Data Generator based on Bayesian Network Model

Usage

```
BN_Data_Generator(arcs_mat, Probs, n, nodename=NULL, cardinality=NULL)
```

Arguments

arcs_mat A matrix that determines the arcs.

Probs The conditional probabilities.

n Data size.

nodename The names of each nodes.

cardinality The cardinality of each nodes.

Author(s)

References

Jae-seong Yoo, (2014), "A Study on Comparison of Bayesian Network Structure Learning Algorithms for Selecting Appropriate Models", M.S. thesis, Department of Statistics, Korea University, Seoul.

See Also

See demo(gen_asia) or demo(topology).

check_cardinality A checker of needs of input_Prob follows cardinality

Description

A checker of needs of input_Prob follows cardinality.

Usage

```
check_cardinality(arcs_mat, nodename=NULL, cardinality=NULL)
```

Arguments

arcs_mat A matrix that determines the arcs.

nodename The names of each nodes.

cardinality The cardinality of each nodes.

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Author(s)

C_M_WO_WC

Correct, Missing, Wrongly Oriented, Wronglyy Corrected Arcs

Description

The existence of the known network structures allows us to define three important terms which indicate the performance of the algorithm (in terms of the number of graphical errors in the learnt structure).

Usage

```
C_M_WO_WC(target_arcs_mat, learnt_arcs_mat)
```

Arguments

target_arcs_mat

A matrix of known network structure.

learnt_arcs_mat

A matrix of learnt network structure.

Value

C (Correct Arcs)

Edges present in the original network and in the learnt network structure.

M (Missing Arcs)

Edges present in the original network but not in the learnt network structure.

WO (Wrongly Oriented Arcs)

Edges present in the learnt network structure, but having opposite orientation when compared with the corresponding edge in the original network structure.

WC (Wrongly Corrected Arcs)

Edges not present in the original network but included in the learnt network structure.

Author(s)

References

X.-w. Chen, G. Anantha, and X. Wang, (2006), An effective structure learning method for constructing gene networks, Bioinformatics, Vol. 22, 1367-1374.

See Also

See demo(gen_asia) or demo(topology).

gen_asia

fromto_to_mat

Convert from 'fromto' to 'matrix'

Description

Convert from 'fromto' to 'matrix' that determines the arcs.

Usage

```
fromto_to_mat(fromto, nodename)
```

Arguments

fromto A matrix form structured 'fromto' that determines the arcs.

nodename The names of each nodes.

Author(s)

gen_asia

Asia (synthetic) data based on a model set by Lauritzen and Spiegelhalter

Description

Small synthetic data set from Lauritzen and Spiegelhalter (1988) about lung diseases (tuberculosis, lung cancer or bronchitis) and visits to Asia.

Usage

```
gen_asia()
```

Value

D (dyspnoea) A two-level factor with levels yes and no.

T (tuberculosis)

A two-level factor with levels yes and no.

L (lung cancer)

A two-level factor with levels yes and no.

B (bronchitis) A two-level factor with levels yes and no.

A (visit to Asia)

A two-level factor with levels yes and no.

S (smoking) A two-level factor with levels yes and no.

X (chest X-ray)

A two-level factor with levels yes and no.

E (tuberculosis versus lung cancer/bronchitis)

A two-level factor with levels yes and no.

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Note

Lauritzen and Spiegelhalter (1988) motivate this example as follows: Shortness-of-breath (dyspnoea) may be due to tuberculosis, lung cancer or bronchitis, or none of them, or more than one of them. A recent visit to Asia increases the chances of tuberculosis, while smoking is known to be a risk factor for both lung cancer and bronchitis. The results of a single chest X-ray do not discriminate between lung cancer and tuberculosis, as neither does the presence or absence of dyspnoea. Standard learning algorithms are not able to recover the true structure of the network because of the presence of a node (E) with conditional probabilities equal to both 0 and 1. Monte Carlo tests seems to behave better than their parametric counterparts.

Author(s)

References

Lauritzen S, Spiegelhalter D (1988). "Local Computation with Probabilities on Graphical Structures and their Application to Expert Systems (with discussion)". Journal of the Royal Statistical Society: Series B (Statistical Methodology), 50(2), 157-224.

See Also

See demo(gen asia).

is_acyclic

Acyclic graphs

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(arcs_mat)
```

Arguments

arcs_mat

A matrix that determines the arcs.

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)

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See Also

is_DAG

is_DAG

Directed acyclic graphs

Description

This function tests whether the given graph is a DAG, a directed acyclic graph.

Usage

```
is_DAG(arcs_mat)
```

Arguments

arcs_mat

A matrix that determines the arcs.

Details

is_dag checks whether there is a directed cycle in the graph. If not, the graph is a DAG.

Value

A logical vector of length one.

Author(s)

See Also

```
is_acyclic
```

make_topology

Bayesian Networks with varying topologies

Description

Bayesian Networks with varying topologies (DAGs) with number of nodes.

Usage

```
make_topology(nodes, topology, Probs, nodename=NULL, cardinality=NULL)
```

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Arguments

nodes The number of nodes.

topology Geometric characteristic. Collapse, Line, Star, PseudoLoop, Diamond, Rhom-

bus.

Probs The conditional probabilities.

nodename The names of each nodes.

cardinality The cardinality of each nodes.

Details

The volume of the manifold is a geometric characteristic associated with the topology of Bayesian network. Each BN produces a different magnitude of the volume based on the DAG of Bayesian network. Collapse, Line, Star, PseudoLoop, Diamond, Rhombus.

Author(s)

References

Eitel J. M. L., (2008), An Information-geometric approach to learning Bayesian network topologies from data, Innovations in Bayesian Networks Studies in Computational Intelligence, Vol. 156, 187-217.

See Also

See demo(topology).

mat_to_fromto Co

Convert from 'matrix' to 'fromto'

Description

Convert from 'matrix' to 'fromto' that determines the arcs.

Usage

mat_to_fromto(arcs_mat)

Arguments

arcs_mat

A matrix that determines the arcs that determines the arcs.

Author(s)

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toss_value

Tossing a Cardinality

Description

Sets up a sample space for the experiment of tossing a cardinality repeatedly with the outcomes "Values".

Usage

```
toss_value(times, num_of_cases, makespace=FALSE)
```

Arguments

makespace Logical.

Value

A data frame, with an equally likely probs column if makespace is TRUE.

Note

It developed of 'tosscoin' function in prob package.

Author(s)

Examples

```
toss_value(1, 3)
toss_value(2, 3)
toss_value(3, 4, makespace = TRUE)
```

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