# Package 'BNDataGenerator'

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R topics documented:	
BNDataGenerator-package	2
big_letters	2
BN_Data_Generator	3
=	4
	4
fromto_to_mat	5
6 -	6
— <i>•</i>	7
	8
	8
mat_to_fromto	9
toss_value	(

2 big\_letters

BNDataGenerator-package

Data Generator based on Bayesian Network Model

## **Description**

Data Generator based on Bayesian Network Model

## **Details**

Package: BN\_Data\_Generator

Type: Package Version: 1.0

Date: 2014-12-28 License: GPL (>=2)

## 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(asia) or demo(topology).

big\_letters

Letters built

## Description

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

## Usage

big\_letters(len)

BN\_Data\_Generator 3

## **Arguments**

len Get or set the length of letters.

#### Author(s)

## **Examples**

```
big_letters(26)
big_letters(50)
big_letters(100)
big_letters(999)
```

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(asia) or demo(topology).

 $C_M_WO_WC$ 

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.

#### 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.

fromto\_to\_mat 5

#### 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(asia) or demo(topology).

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

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.
```

## 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)

is\_acyclic 7

#### 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(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)

#### See Also

is\_DAG

8 make\_topology

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)
```

mat\_to\_fromto 9

## **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

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)

10 toss\_value

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

times Number of tosses. num\_of\_cases Cardinality.

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)
```

## **Index**

```
*Topic Bayesian Network
    BN_Data_Generator, 3
    BNDataGenerator-package, 2
*Topic letters
    big_letters, 2
big_letters, 2
BN_Data_Generator, 3
{\tt BNDataGenerator}
        (BNDataGenerator-package), 2
BNDataGenerator-package, 2
C_M_WO_WC, 4
check_cardinality, 4
fromto_to_mat, 5
gen_asia, 6
is_acyclic, 7, 8
is_DAG, 7, 8
make_topology, 8
mat_to_fromto, 9
toss_value, 10
```