# Package 'dagmc'

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BNLearnScorer CachedScorer CalculateAcceptanceRates CalculateNodeMoveNeighbourhood CalculateSplitJoinNeighbourhood

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BNLearnScorer 3

## **Description**

A thin wrapper on the bnlearn::score function.

## Usage

```
BNLearnScorer(node, parents, ...)
```

#### **Arguments**

node Name of node to score.

parents The parent nodes of node.

... The ellipsis is used to pass other parameters to the scorer.

## **Examples**

```
data <- bnlearn::learning.test
BNLearnScorer('A', c('B', 'C'), data = data)
BNLearnScorer('A', c(), data = data)
BNLearnScorer('A', vector(), data = data)
BNLearnScorer('A', NULL, data = data)
BNLearnScorer('A', c('B', 'C'), data = data, type = "bde", iss = 100)
BNLearnScorer('A', c('B', 'C'), data = data, type = "bde", iss = 1)</pre>
```

CachedScorer

This builds the score cache. It can be used for problems where the score only changes as a function of (node, parents).

# Description

This builds the score cache. It can be used for problems where the score only changes as a function of (node, parents).

# Usage

```
CachedScorer(scorer, max_size = NULL)
```

# **Arguments**

scorer A scorer.

max\_size Not implemented. Maximum number of scores to store in the cache. If the total

number of combinations is greater than this number then the cache follows a

least recently used replacement policy.

## **Examples**

```
scorer <- CreateScorer(data = bnlearn::learning.test)
cached_scorer <- CachedScorer(scorer)
cached_scorer('A', c('B', 'C'))</pre>
```

CalculateAcceptanceRates

Calculate acceptance rates per proposal.

## **Description**

This makes the assumption that the proposal has saved a variable "proposal\_used" and mcmc has saved a variable 'accept'.

## Usage

```
CalculateAcceptanceRates(chains, group_by = NULL)
```

#### **Arguments**

chains MCMC chains.

group\_by Vector of strings that are in c("chain", "proposal\_used"). Default is NULL which

will return the acceptance rates marginalised over chains and the proposal used.

## Value

Summary of acceptance rates per grouping.

 ${\tt CalculateNodeMoveNeighbourhood}$ 

Calculate neighbourhood for node move.

## **Description**

Calculate neighbourhood for node move.

## Usage

CalculateNodeMoveNeighbourhood(partitioned\_nodes)

# **Arguments**

partitioned\_nodes

Labelled partition.

 ${\tt CalculateSplitJoinNeighbourhood}$ 

Calculate neighbourhood for the split or join proposal.

# Description

The number of split combinations prescribed by KP15 is ambiguous when a partition element has only 1 node. A split for a partition element with 1 node results in a proposal to stay still, as such I remove that proposal.

# Usage

CalculateSplitJoinNeighbourhood(partitioned\_nodes)

## **Arguments**

 ${\tt CalculateStayStillNeighbourhood}$ 

Calculate neighbourhood for staying still.

## **Description**

Calculate neighbourhood for staying still.

# Usage

CalculateStayStillNeighbourhood(partitioned\_nodes)

# **Arguments**

partitioned\_nodes

A labelled partition.

 ${\tt CalculateSwapAdjacentNodeNeighbourhood}$ 

Calculate neighbourhood for swapping nodes.

# Description

Calculate neighbourhood for swapping nodes.

## Usage

 ${\tt CalculateSwapAdjacentNodeNeighbourhood(partitioned\_nodes)}$ 

## **Arguments**

```
partitioned_nodes
```

Labelled partition.

 ${\tt CalculateSwapNodeNeighbourhood}$ 

Calculate neighbourhood for swapping nodes.

# Description

Calculate neighbourhood for swapping nodes.

## Usage

 ${\tt CalculateSwapNodeNeighbourhood(partitioned\_nodes)}$ 

## **Arguments**

```
partitioned_nodes
```

Labelled partition.

CheckBlacklistObeyed Check blacklist obeyed.

# Description

If an edge between two nodes is blacklisted in Partition MCMC the adjacent partition element cannot be the only direct node for it's blacklisted child.

## Usage

```
CheckBlacklistObeyed(partitioned_nodes, blacklist = NULL, nodes = NULL)
```

CheckWhitelistObeyed

## Arguments

partitioned\_nodes

Labelled partition.

blacklist A data.frame of (parent, child) pairs representing edges that cannot be in the

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DAG.

nodes A vector of node names to check. Default is to check all child nodes in the

blacklist.

CheckWhitelistObeyed Check whitelist is obeyed.

## Description

Check whitelist is obeyed.

#### Usage

CheckWhitelistObeyed(partitioned\_nodes, whitelist = NULL, nodes = NULL)

## **Arguments**

partitioned\_nodes

Labelled partition.

whitelist A data.frame of (parent, child) pairs representing edges that must be in the DAG.

nodes A vector of node names to check. Default is to check all child nodes in the

whitelist.

CollectDags Collect DAGs.

#### **Description**

Get the unique set of DAGs along with their log\_score. It also estimates the normalised log score assuming  $\tilde{Z} = \Sigma_s^S p(G_s) p(D|G_s)$  where  $\{G_1, G_2, G_3, ..., G_S\}$  is the set of unique DAGs in the the chain. This assumes that you have captured the most probable graphs, such that  $\tilde{Z}$  is approximately equal to the true evidence  $Z = \Sigma_{G \in \mathcal{G}} p(G) p(D|G)$  where you sum across all possible graphs  $\mathcal{G}$ . This also makes the assumption that the scoring method used is proportional to the posterior probability;  $\operatorname{score}(G,D) \propto p(G) p(G|D)$ .

## Usage

```
CollectDags(post_chain)
```

#### **Arguments**

post\_chain A chain that includes a DAG per sample.

8 CreateScorer

#### Value

dag\_collection A list with entries: dag: A list of all unique DAGs within the sample. log\_score: A vector with the log\_scores for each DAG. log\_evidence: A numeric value representing the evidence  $log(\tilde{Z}) = \Sigma_s^S log(p(G_s)p(D|G_s))$ . log\_norm\_score: A vector of normalised log\_scores for each DAG  $G_s$  using  $p(G_s|D) = log(p(G_s)p(D|G_s)) - log(\tilde{Z})$ .

CreateScorer

Scorer constructor.

## **Description**

Scorer constructor.

## Usage

```
CreateScorer(
   scorer = BNLearnScorer,
   ...,
   max_parents = Inf,
   blacklist = NULL,
   whitelist = NULL,
   cache = FALSE
)
```

# Arguments

A scorer function that takes (node, parents) as parameters. Default is BNLearn-Scorer.

Parameters to pass to scorer.

The maximum number of allowed parents. Default is infinite.

A boolean matrix of (parent, child) pairs where TRUE represents edges that cannot be in the DAG. Default is NULL which represents no blacklisting.

Whitelist

A boolean matrix of (parent, child) pairs where TRUE represents edges that must be in the DAG. Default is NULL which represents no whitelisting.

cache A boolean to indicate whether to build the cache. The cache only works for

problems where the scorer only varies as a function of (node, parents). Default

is FALSE.

## **Examples**

```
scorer <- CreateScorer(data = bnlearn::asia)</pre>
```

DefaultProposal 9

DefaultProposal	Default proposal constructor.
Der dar tri r oposar	Dejanti proposat constructor.

# Description

Default proposal constructor.

# Usage

```
DefaultProposal(p = c(0.33, 0.33, 0.165, 0.165, 0.01), verbose = TRUE)
```

## **Arguments**

-	n	<b>Probability</b>	for each	proposal	in the	order (split	ioin.	node	move.	swap nod	e.

swap\_adjacent, stay\_still).

verbose Boolean flag to record proposal used.

FindChangedNodes	Find nodes with changed parent combinations between different la-
-	belled partitions.

# Description

TODO: This is quite slow. From the proposal we should be able to determine the nodes that need to be rescored rather than finding them using this function.

# Usage

```
{\tt FindChangedNodes(old\_partitioned\_nodes, new\_partitioned\_nodes, scorer)}
```

# Arguments

```
\begin{tabular}{ll} old\_partitioned\_nodes \\ Labelled partition. \\ new\_partitioned\_nodes \\ Labelled partition. \\ \end{tabular}
```

scorer Scorer object.

## Value

Vector of changed nodes.

10 GetEmptyDAG

## **Examples**

```
scorer = CreateScorer()

old_dag <- UniformlySampleDAG(LETTERS[1:5])
old_partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(old_dag)

new_dag <- UniformlySampleDAG(LETTERS[1:5])
new_partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(new_dag)

changed_nodes <- FindChangedNodes(old_partitioned_nodes, new_partitioned_nodes, scorer)</pre>
```

FlattenChains

Flatten list of chains.

# Description

Flatten list of chains.

# Usage

FlattenChains(chains)

# Arguments

chains

MCMC chains.

 ${\tt GetEmptyDAG}$ 

Get an empty DAG given a set of nodes.

## **Description**

Get an empty DAG given a set of nodes.

# Usage

GetEmptyDAG(nodes)

## **Arguments**

nodes

A vector of node names.

## Value

An adjacency matrix with elements designated as (parent, child).

 ${\tt GetLowestPairwiseScoringEdges}$ 

Preprocessing for blacklisting. Get the lowest pairwise scoring edges.

## **Description**

Get the lowest c pairwise scoring edges represented as a blacklist matrix. This blacklisting procedure is motivated by Koller & Friedman (2003).

# Usage

GetLowestPairwiseScoringEdges(scorer, n\_retain)

## **Arguments**

scorer A scorer object.

n\_retain An integer representing the number of edges to retain.

#### Value

A boolean matrix of (parent, child) pairs for blacklisting.

## References

1. Koller D, Friedman N. Being Bayesian about network structure. A Bayesian approach to structure discovery in Bayesian networks. Mach Learn. 2003;50(1):95–125.

GetNodePartition

Get a node's partition element number.

# Description

Get a node's partition element number.

## Usage

GetNodePartition(partitioned\_nodes, node)

# Arguments

partitioned\_nodes

Labelled partition.

node Node name.

#### Value

Node's partition element number.

12 GetOrderedPartition

 ${\tt GetNumberOfPartitions} \ \ \textit{Get number of partitions}.$ 

# Description

Calculate the number of partitions for a given labelled partition. This is 'm' in Kuipers & Moffa (2015).

# Usage

GetNumberOfPartitions(partitioned\_nodes)

# Arguments

partitioned\_nodes

Labelled partition.

 ${\tt GetOrderedPartition}$ 

Get ordered labelled partition.

# Description

Calculate the ordered partition. Denoted as lamba in Kuipers & Moffa (2015).

# Usage

GetOrderedPartition(partitioned\_nodes)

# **Arguments**

partitioned\_nodes

Labelled partition.

#### Value

Ordered partition.

GetParentCombinations 13

GetParentCombinations Get parent combinations for a given node.

# Description

Get parent combinations for a given node.

# Usage

GetParentCombinations(partitioned\_nodes, node, scorer)

# **Arguments**

partitioned\_nodes

Labelled partition.

node

Node name.

scorer

A scorer object.

# Value

List of parent combinations.

 $\label{thm:continuous} {\it GetPartitionedNodesFromAdjacencyMatrix} \\ {\it Map\ DAG\ to\ a\ labelled\ partition}.$ 

# Description

This partitions nodes into levels of outpoints as explained in Section 4.1 of Kuipers & Moffa 2015. This takes an adjacency matrix and returns a data.frame of (partition, node) pairs

## Usage

GetPartitionedNodesFromAdjacencyMatrix(adjacency)

# **Arguments**

adjacency

Adjacency matrix.

# Value

Labelled partition for the given adjacency matrix.

14 GetRestrictedParents

GetPartitionNodes

Get nodes in a partition element.

# Description

Get nodes in a partition element.

# Usage

```
GetPartitionNodes(partitioned_nodes, elements)
```

## **Arguments**

partitioned\_nodes

Labelled partition.

elements

An integer or vector of integers for the partition element number.

 ${\tt GetRestrictedNodes}$ 

Get nodes that have restricted parents.

# Description

Get nodes that have restricted parents.

# Usage

```
GetRestrictedNodes(list)
```

# Arguments

list

A black or white list.

GetRestrictedParents

Get black or white listed parents.

# **Description**

Get black or white listed parents.

# Usage

```
GetRestrictedParents(node, listed = NULL)
```

## **Arguments**

node

The name of the node to get white or black listed parents.

listed

A black or white list.

LogSumExp 15

LogSumExp	Log-Sum-Exponential calculation using the trick that limits underflow
	issues.

# Description

Log-Sum-Exponential calculation using the trick that limits underflow issues.

# Usage

```
LogSumExp(x)
```

# Arguments

X

A vector of numeric.

## Value

Log-Sum-Exponential (LSE) of x.

NodeMove

Node move proposal.

# Description

Node move proposal.

# Usage

```
NodeMove(partitioned_nodes)
```

# Arguments

```
partitioned_nodes
```

Labelled partition.

16 **PartitionMCMC** 

OrderPartitionedNodes Order partitioned nodes.

# Description

Order partitioned nodes.

# Usage

OrderPartitionedNodes(partitioned\_nodes)

## **Arguments**

partitioned\_nodes

Labelled partition.

# Value

Labelled partitioned in descending partition element order.

PartitionMCMC Transition objects. A one step implementation of partition MCMC.

This acts as a constructor.

# Description

This is a constructor for a single Partition MCMC step. The function constructs an environment with the proposal and verbose flag. It then returns a function which takes the current\_state and a scorer object.

# Usage

```
PartitionMCMC(proposal = NULL, verbose = TRUE)
```

## **Arguments**

Proposal function. Default is the DefaultProposal. proposal

verbose Flag to pass meme information.

## Value

Function that takes the current state and scorer that outputs a new state.

PartitionSplitJoin 17

## **Examples**

```
dag <- UniformlySampleDAG(c('A', 'B', 'C', 'D', 'E', 'F'))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)

scorer <- list(
    scorer = BNLearnScorer,
    parameters = list(data = bnlearn::learning.test)
    )

current_state <- list(
    state = partitioned_nodes,
    log_score = ScoreLabelledPartition(partitioned_nodes, scorer)
    )

pmcmc <- PartitionMCMC(proposal = PartitionSplitJoin)
pmcmc(current_state, scorer)</pre>
```

PartitionSplitJoin

Partition split or join constructor.

# Description

Partition split or join constructor.

#### Usage

```
PartitionSplitJoin(partitioned_nodes)
```

# Arguments

partitioned\_nodes

Labelled partition.

PlotScoreTrace

Plot the score trace.

## **Description**

Plot the score trace.

# Usage

```
PlotScoreTrace(
  chains,
  attribute = "log_score",
  n_burnin = 0,
  same_plot = TRUE,
  col = NULL,
  ...
)
```

18 ProposeNodeMove

## **Arguments**

chains MCMC chains.

attribute Name of attribute to plot. Default is "log\_score".

n\_burnin Number of steps to remove as burnin.

same\_plot Whether to plot on the same figure or on multiple figures.

col A string representing a color for a single chain or a vector of strings to cycle

through for multiple chains.

... Extra parameters to pass to the plot and graphics::line functions.

PostProcessChains Analysis of chains. Equilibrium states.

#### **Description**

This allows you to remove a burnin and thin the chains after processing.

## Usage

PostProcessChains(chains, n\_burnin = 0, n\_thin = 1)

#### **Arguments**

chains MCMC chains.

n\_burnin Number of steps to remove at the start as a burnin. Default is 0.

n\_thin Number of steps between retained states. Default is 1.

ProposeNodeMove Propose individual node movement.

#### **Description**

This proposes that a single node selected uniformly can either:

- 1. Move to any current partition.
- 2. Move to any gap between or at the ends of the partitions.

Any of these moves are possible and are selected uniformly with two exceptions:

- 1. The selected node cannot move into adjacent gaps if it originated from a single node partition.
- 2. The selected node cannot move to the immediately higher gap if it originated from a two node partition.

#### Usage

ProposeNodeMove(partitioned\_nodes)

## **Arguments**

```
partitioned_nodes Labelled partition.
```

## **Examples**

```
dag <- UniformlySampleDAG(c('A', 'B', 'C', 'D', 'E', 'F'))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)
ProposeNodeMove(partitioned_nodes)</pre>
```

ProposePartitionSplitJoin

Propose a split or join of two partitions.

# Description

This is the 'Basic Move' (i.e. algorithm 1) in Kuipers & Moffa (2015). There is a caveat in that the split proposal for a partition with one element is ambiguous, as a split for such a partition element results in a stay still proposal. Such a proposal has been removed.

# Usage

```
ProposePartitionSplitJoin(partitioned_nodes)
```

#### **Arguments**

```
partitioned_nodes
A labelled partition.
```

## Value

A proposed labelled partition.

## **Examples**

```
dag <- UniformlySampleDAG(c('A', 'B', 'C', 'D', 'E', 'F'))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)
ProposePartitionSplitJoin(partitioned_nodes)</pre>
```

 ${\tt ProposeStayStill}$ 

Propose that the partition stays still.

# Description

Propose that the partition stays still.

# Usage

```
ProposeStayStill(partitioned_nodes)
```

# Arguments

```
partitioned_nodes
```

A labelled partition.

## Value

A proposed labelled partition.

## **Examples**

```
\label{eq:dag} $$ $$ dag <- UniformlySampleDAG(c('A', 'B', 'C', 'D', 'E', 'F'))$ $$ partitioned\_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag) $$ ProposeStayStill(partitioned\_nodes) $$
```

ProposeSwapAdjacentNode

Propose that two nodes swap partition elements.

## **Description**

Propose that two nodes swap partition elements.

# Usage

ProposeSwapAdjacentNode(partitioned\_nodes)

## **Arguments**

```
partitioned_nodes
```

labelled partition.

## Value

A proposed labelled partition.

ProposeSwapNode 21

## **Examples**

```
dag <- UniformlySampleDAG(c('A', 'B', 'C', 'D', 'E', 'F'))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)
ProposeStayStill(partitioned_nodes)</pre>
```

ProposeSwapNode

Propose that two nodes swap partition elements.

## **Description**

Propose that two nodes swap partition elements.

## Usage

ProposeSwapNode(partitioned\_nodes)

## **Arguments**

```
partitioned_nodes
```

labelled partition.

#### Value

A proposed labelled partition.

## **Examples**

```
dag <- UniformlySampleDAG(c('A', 'B', 'C', 'D', 'E', 'F'))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)
ProposeStayStill(partitioned_nodes)</pre>
```

 ${\tt SampleChain}$ 

Sample a single chain.

# Description

Sample a single chain.

#### Usage

```
SampleChain(n_results, init_state, transition, scorer, n_thin = 1)
```

## Arguments

 $n\_results$  Number of saved states.

init\_state An initial state that can be passed to transition.

transition A transition function. scorer A scorer object.

n\_thin Number of steps between saved states.

22 SampleChains

## **Examples**

```
data = bnlearn::learning.test

dag <- UniformlySampleDAG(colnames(data))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)

scorer_1 <- list(
    scorer = BNLearnScorer,
    parameters = list(data = data)
    )

results <- SampleChain(10, partitioned_nodes, PartitionMCMC(), scorer_1)</pre>
```

SampleChainDAGs

Sampled DAG from chains.

# Description

Sampled DAG from chains.

## Usage

```
SampleChainDAGs(chains, scorer)
```

## **Arguments**

chains MCMC chains. scorer Scorer object.

#### Value

Chains with sample dags and their corresponding score.

SampleChains

Sample multiple chains in parallel.

## **Description**

Sample multiple chains in parallel.

# Usage

```
SampleChains(
   n_results,
   init_state,
   transition,
   scorer,
   n_thin = 1,
   n_parallel_chains = 2
```

## **Arguments**

n\_results Number of saved states per chain.

init\_state An initial state that can be passed to transition. This can be a single state or a

list of states for each parallel chain.

transition A transition function. scorer A scorer object.

n\_thin Number of steps between saved states.

n\_parallel\_chains

Number of chains to run in parallel. Default is 2.

#### Value

List of results.

# **Examples**

```
data = bnlearn::learning.test

dag <- UniformlySampleDAG(colnames(data))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)

scorer <- list(
    scorer = BNLearnScorer,
    parameters = list(data = data)
    )

results <- SampleChains(10, partitioned_nodes, PartitionMCMC(), scorer)</pre>
```

SampleDAGFromLabelledPartition

Sample a DAG from a labelled partition.

# Description

Sample a DAG from a labelled partition.

#### Usage

SampleDAGFromLabelledPartition(partitioned\_nodes, scorer)

## **Arguments**

partitioned\_nodes

Labelled partition.

scorer Scorer object.

## Value

A matrix with elements of (parent, child) cells with 1 representing and edge and 0 otherwise.

24 ScoreDiff

## **Examples**

```
data <- bnlearn::learning.test

dag <- UniformlySampleDAG(colnames(data))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)

scorer <- CreateScorer(data = data)

SampleDAGFromLabelledPartition(partitioned_nodes, scorer = scorer)</pre>
```

ScoreDAG

Score DAG.

## **Description**

Score DAG.

## Usage

```
ScoreDAG(dag, scorer)
```

# Arguments

dag Adjacency matrix of (parent, child) entries with 1 denoting an edge and 0 other-

wise.

scorer Scorer object.

## Value

Log of DAG score.

ScoreDiff

Calculate the difference in log scores between two labelled partitions.

## **Description**

Calculate the difference in log scores between two labelled partitions.

# Usage

```
ScoreDiff(
  old_partitioned_nodes,
  new_partitioned_nodes,
  scorer,
  rescore_nodes = NULL
)
```

ScoreLabelledPartition 25

#### **Arguments**

#### Value

Log of score difference between two labelled partitions.

## **Examples**

```
data <- bnlearn::learning.test

old_dag <- UniformlySampleDAG(names(data))
old_partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(old_dag)

new_dag <- UniformlySampleDAG(names(data))
new_partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(new_dag)

scorer <- list(
    scorer = BNLearnScorer,
    parameters = list(data = data)
    )

ScoreDiff(old_partitioned_nodes, new_partitioned_nodes, scorer = scorer)</pre>
```

ScoreLabelledPartition

Score labelled partition by adding the log scores for each node.

## **Description**

Score labelled partition by adding the log scores for each node.

## Usage

```
ScoreLabelledPartition(partitioned_nodes, scorer)
```

## **Arguments**

#### Value

Log of the node score.

26 ScoreNode

## **Examples**

```
data <- bnlearn::learning.test

dag <- UniformlySampleDAG(names(data))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)

scorer <- list(
    scorer = BNLearnScorer,
    parameters = list(data = data)
    )

ScoreLabelledPartition(partitioned_nodes, scorer)</pre>
```

ScoreNode

Score node by marginalising over parent combinations.

## **Description**

Score node by marginalising over parent combinations.

# Usage

```
ScoreNode(partitioned_nodes, node, scorer)
```

# Arguments

```
\begin{tabular}{ll} partitioned\_nodes & Labelled partition. \\ node & The node name. \\ scorer & A scorer object. \\ \end{tabular}
```

## Value

Log of the node score.

#### **Examples**

```
data <- bnlearn::learning.test

dag <- UniformlySampleDAG(names(data))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)

scorer <- list(
    scorer = BNLearnScorer,
    parameters = list(data = data)
    )

ScoreNode(partitioned_nodes, 'A', scorer)</pre>
```

Score Table Node 27

ScoreTableNode

Calculate score tables for (node, parents) combinations.

# Description

Calculate score tables for (node, parents) combinations.

# Usage

```
ScoreTableNode(partitioned_nodes, node, scorer)
```

## **Arguments**

```
partitioned_nodes
```

Labelled partition.

node Name of node. scorer Scorer object.

#### Value

List of log\_scores for each combination in parent\_combinations.

## **Examples**

```
data <- bnlearn::learning.test

dag <- UniformlySampleDAG(names(data))
partitioned_nodes <- GetPartitionedNodesFromAdjacencyMatrix(dag)

scorer <- list(
    scorer = BNLearnScorer,
    parameters = list(data = data)
    )

ScoreTableNode(partitioned_nodes, 'A', scorer)</pre>
```

StayStill

 ${\it StayStill\ proposal}.$ 

## **Description**

StayStill proposal.

## Usage

```
StayStill(partitioned_nodes)
```

## **Arguments**

```
partitioned_nodes
```

Labelled partition.

SwapAdjacentNode

Swap nodes from adjacent partition elements proposal.

# Description

Swap nodes from adjacent partition elements proposal.

## Usage

```
SwapAdjacentNode(partitioned_nodes)
```

## **Arguments**

```
partitioned_nodes
```

Labelled partition.

SwapNode

Swap node proposal.

# Description

Swap node proposal.

## Usage

SwapNode(partitioned\_nodes)

## **Arguments**

partitioned\_nodes

Labelled partition.

UniformlySampleDAG

Uniformly sample DAG given a set of nodes.

# Description

Uniformly sample DAG given a set of nodes.

# Usage

UniformlySampleDAG(nodes)

# Arguments

nodes

A vector of node names.

## Value

Adjacency matrix with elements designated as (parent, child).

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