

Package ‘jti’

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Title Junction Tree Inference

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Description Minimal and memory efficient implementation of the junction tree algorithm using the Lauritzen-Spiegelhalter scheme;
S. L. Lauritzen and D. J. Spiegelhalter (1988)
<<https://www.jstor.org/stable/2345762?seq=1>>.

Depends R (>= 3.5.0)

URL <https://github.com/mlindsk/jti>

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jti-package

jti: Junction Tree Inference

Description

Minimal and memory efficient implementation of the junction tree algorithm using the Lauritzen-Spiegelhalter scheme. S. L. Lauritzen and D. J. Spiegelhalter (1988).

Author(s)

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See Also

Useful links:

- <https://github.com/mlindsk/jti>

asia

Asia

Description

Small synthetic data set from Lauritzen and Spiegelhalter (1988) about lung diseases (tuberculosis, lung cancer or bronchitis) and visits to Asia. This copy of the data was taken from the R package "bnlearn" where all values "yes" have been converted to "y" and all values "no" have been converted to "n".

Usage

```
asia
```

Format

An object of class `tbl_df` (inherits from `tbl`, `data.frame`) with 5000 rows and 8 columns.

Details

D (**dyspnea**)

T (**tuberculosis**)

L (**lung cancer**)

B (**bronchitis**)

A (**visit to Asia**)

S (**smoking**)

X (**chest C-ray**)

E (**tuberculosis vs cancer/bronchitis**)

References

[bnlearn-asia](#)

asia2

Asia2

Description

See the `asia` data for information. This version, has class `bn_fit`.

Usage

```
asia2
```

Format

An object of class `list` of length 8.

References

[bnlearn-asia](#)

<code>bnfit_to_cpts</code>	<i>bnfit to cpts</i>
----------------------------	----------------------

Description

Convert a `bn.fit` object (a list of `cpts` from the `bnlearn` package) into a list of ordinary array-like `cpts`

Usage

```
bnfit_to_cpts(x)
```

Arguments

<code>x</code>	A <code>bn.fit</code> object
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<code>compile</code>	<i>Compile information</i>
----------------------	----------------------------

Description

Compiled objects are used as building blocks for junction tree inference

Usage

```
compile(x, evidence = NULL, root_node = "", joint_vars = NULL, tri = "minimal")
```

```
## S3 method for class 'cpt_list'
```

```
compile(x, evidence = NULL, root_node = "", joint_vars = NULL, tri = "min_nei")
```

Arguments

<code>x</code>	An object returned from <code>cpt_list</code>
<code>evidence</code>	A named vector. The names are the variables and the elements are the evidence.
<code>root_node</code>	A node for which we require it to live in the root clique (the first clique).
<code>joint_vars</code>	A vector of variables for which we require them to be in the same clique. Edges between all these variables are added to the moralized graph.
<code>tri</code>	The optimization strategy used for triangulation. Either one of <code>'min_nei'</code> , <code>'min_fill'</code> , <code>'min_sp'</code> , <code>'sparse'</code> , <code>'minimal'</code>

Details

The Junction Tree Algorithm performs both a forward and inward message passing (collect and distribute). However, when the forward phase is finished, the root clique potential is guaranteed to be the joint pmf over the variables involved in the root clique. Thus, if it is known in advance that a specific variable is of interest, the algorithm can be terminated after the forward phase. Use the `root_node` to specify such a variable and specify `propagate = "collect"` in the junction tree algorithm function `jt`.

Moreover, if interest is in some joint pmf for variables that end up being in different cliques these variables must be specified in advance using the `joint_vars` argument. The compilation step then adds edges between all of these variables to ensure that at least one clique contains all of them.

Evidence can be entered either at compile stage or after compilation. Hence, one can also combine evidence from before compilation with evidence after compilation. `Before` refers to entering evidence in the `'compile'` function and `after` refers to entering evidence in the `'jt'` function.

Examples

```
cpt1 <- cpt_list(asia2)
cp1 <- compile(cpt1, evidence = c(bronc = "yes"), joint_vars = c("bronc", "tub"))
print(cp1)
dim_names(cp1)
plot(get_graph(cp1))
```

cpt_list

Conditional probability list

Description

A check and conversion of cpts to be used in the junction tree algorithm

Usage

```
cpt_list(x, g = NULL)

## S3 method for class 'list'
cpt_list(x, g = NULL)

## S3 method for class 'data.frame'
cpt_list(x, g)
```

Arguments

`x` Either a named list with cpts in form of array-like object(s) where names must be the child node or a `data.frame`

`g` Either a directed acyclic graph (DAG) as an `igraph` object or a decomposable graph as an `igraph` object. If `x` is a list, `g` must be `NULL`. The procedure then deduce the graph from the conditional probability tables.

Examples

```

library(igraph)
e1 <- matrix(c(
  "A", "T",
  "T", "E",
  "S", "L",
  "S", "B",
  "L", "E",
  "E", "X",
  "E", "D",
  "B", "D"),
  nc = 2,
  byrow = TRUE
)

g <- igraph::graph_from_edgelist(e1)
cl <- cpt_list(asia, g)

print(cl)
dim_names(cl)
names(cl)
plot(get_graph(cl))

```

dim_names

Cpt list getters

Description

Getter methods for cpt list objects

Usage

```

dim_names(x)

## S3 method for class 'cpt_list'
dim_names(x)

## S3 method for class 'cpt_list'
names(x)

## S3 method for class 'charge'
dim_names(x)

## S3 method for class 'charge'
names(x)

```

Arguments

x cpt_list or a compiled object

get_cliques *Return the cliques of a junction tree*

Description

Return the cliques of a junction tree

Usage

```
get_cliques(x)

## S3 method for class 'jt'
get_cliques(x)

## S3 method for class 'charge'
get_cliques(x)

get_clique_root(x)

## S3 method for class 'jt'
get_clique_root(x)
```

Arguments

x A junction tree object, jt.

See Also

[jt](#)

Examples

```
# See Example 5 and 6 of the 'jt' function
```

 get_graph

Get graph

Description

Retrieve the graph from

Usage

```
get_graph(x)
```

```
## S3 method for class 'charge'
```

```
get_graph(x)
```

```
## S3 method for class 'cpt_list'
```

```
get_graph(x)
```

Arguments

x cpt_list or a compiled object

Value

A graph as an igraph object

jt

Junction Tree

Description

Construction of a junction tree and message passing

Usage

```
jt(x, evidence = NULL, flow = "sum", propagate = "full")
```

```
## S3 method for class 'charge'
```

```
jt(x, evidence = NULL, flow = "sum", propagate = "full")
```

Arguments

x An object return from compile
 evidence A named vector. The names are the variables and the elements are the evidence
 flow Either "sum" or "max"
 propagate Either "no", "collect" or "full".

Details

Evidence can be entered either at compile stage or after compilation. Hence, one can also combine evidence from before compilation with evidence after compilation. Before refers to entering evidence in the 'compile' function and after refers to entering evidence in the 'jt' function.

Value

A jt object

See Also

[query_belief](#), [mpe](#), [get_cliques](#), [get_clique_root](#), [propagate](#)

Examples

```
# Setting up the network
# -----

library(igraph)
el <- matrix(c(
  "A", "T",
  "T", "E",
  "S", "L",
  "S", "B",
  "L", "E",
  "E", "X",
  "E", "D",
  "B", "D"),
  nc = 2,
  byrow = TRUE
)

g <- igraph::graph_from_edgelist(el)
plot(g)
# -----

# Data
# ----
# We use the asia data; see the man page (?asia)

# Compilation
# -----
cl <- cpt_list(asia, g) # Checking and conversion
cp <- compile(cl)

# After the network has been compiled, the graph has been triangulated and
# moralized. Furthermore, all conditional probability tables (CPTs) has been
# designated one of the cliques (in the triangulated and moralized graph).

# Example 1: sum-flow without evidence
# -----
```

```

jt1 <- jt(cp)
plot(jt1)
print(jt1)
query_belief(jt1, c("E", "L", "T"))
query_belief(jt1, c("B", "D", "E"), type = "joint")

# Notice, that jt1 is equivalent to:
# jt1 <- jt(cp, propagate = "no")
# jt1 <- propagate(jt1, prop = "full")
# That is; it is possible to postpone the actual propagation

# Example 2: sum-flow with evidence
# -----

e2 <- c(A = "y", X = "n")
jt2 <- jt(cp, e2)
query_belief(jt2, c("B", "D", "E"), type = "joint")

# Notice that, the configuration (D,E,B) = (y,y,n) has changed
# dramatically as a consequence of the evidence

# We can get the probability of the evidence:
query_evidence(jt2)

# Example 3: max-flow without evidence
# -----

jt3 <- jt(cp, flow = "max")
mpe(jt3)

# Example 4: max-flow with evidence
# -----

e4 <- c(T = "y", X = "y", D = "y")
jt4 <- jt(cp, e4, flow = "max")
mpe(jt4)

# Notice, that T, E, S, B, X and D has changed from "n" to "y"
# as a consequence of the new evidence e4

# Example 5: specifying a root node and only collect to save run time
# -----

cp5 <- compile(cpt_list(asia, g), root_node = "X")
jt5 <- jt(cp5, propagate = "collect")
query_belief(jt5, get_clique_root(jt5), "joint")

# We can only query from the root clique now (clique 1)
# but we have ensured that the node of interest, "X", does indeed live in
# this clique

```

```
# Example 6: Compiling from a list of conditional probabilities
# -----

# * We need a list with CPTs which we extract from the asia2 object
#   - the list must be named with child nodes
#   - The elements need to be array-like objects

cl <- cpt_list(asia2)
cp6 <- compile(cl)

# Inspection; see if the graph correspond to the cpts
# g <- get_graph(cp6)
# plot(g)

# This time we specify that no propagation should be performed
jt6 <- jt(cp6, propagate = "no")

# We can now inspect the collecting junction tree and see which cliques
# are leaves and parents
plot(jt6)
get_cliques(jt6)
get_clique_root(jt6)

leaves(jt6)
unlist(parents(jt6))

# That is;
# - clique 2 is parent of clique 1
# - clique 3 is parent of clique 4 etc.

# Next, we send the messages from the leaves to the parents
jt6 <- send_messages(jt6)

# Inspect again
plot(jt6)

# Send the last message to the root and inspect
jt6 <- send_messages(jt6)
plot(jt6)

# The arrows are now reversed and the outwards (distribute) phase begins
leaves(jt6)
parents(jt6)

# Clique 2 (the root) is now a leaf and it has 1, 3 and 6 as parents.

# Finishing the message passing
jt6 <- send_messages(jt6)
jt6 <- send_messages(jt6)

# Queries can now be performed as normal
query_belief(jt6, c("either", "tub"), "joint")
```

leaves

Query Parents or Leaves in a Junction Tree

Description

Return the clique indices of current parents or leaves in a junction tree

Usage

```
leaves(jt)
```

```
## S3 method for class 'jt'
```

```
leaves(jt)
```

```
parents(jt)
```

```
## S3 method for class 'jt'
```

```
parents(jt)
```

Arguments

jt A junction tree object, jt.

See Also

[jt](#), [get_cliques](#)

Examples

```
# See example 6 in the help page for the jt function
```

mpe

Most Probable Explanation

Description

Returns the most probable explanation given the evidence entered in the junction tree

Usage

```
mpe(x)
```

```
## S3 method for class 'jt'
```

```
mpe(x)
```

Arguments

x A junction tree object, jt, with max-flow.

See Also

[jt](#)

Examples

```
# See the 'jt' function
```

new_mpd	<i>Maximal Prime Decomposition</i>
---------	------------------------------------

Description

Find the maximal prime decomposition and its associated junction tree

Usage

```
new_mpd(graph)
```

Arguments

graph Neighbor matrix

Value

- prime_ints: a list with the prime components, - flawed: indicating which prime components that are triangulated - jt_collect: the MPD junction tree prepared for collecting

Examples

```
library(igraph)
e1 <- matrix(c(
  "A", "T",
  "T", "E",
  "S", "L",
  "S", "B",
  "L", "E",
  "E", "X",
  "E", "D",
  "B", "D"),
  nc = 2,
  byrow = TRUE
)

g <- igraph::graph_from_edgelist(e1, directed = FALSE)
```

```
A <- igraph::as_adjacency_matrix(g, sparse = FALSE)
new_mpd(A)
```

plot.jt *A plot method for junction trees*

Description

A plot method for junction trees

Usage

```
## S3 method for class 'jt'
plot(x, ...)
```

Arguments

x A junction tree object, jt.
... For S3 compatability. Not used.

See Also

[jt](#)

print.charge *A print method for compiled objects*

Description

A print method for compiled objects

Usage

```
## S3 method for class 'charge'
print(x, ...)
```

Arguments

x A compiled object
... For S3 compatability. Not used.

See Also

[jt](#)

print.cpt_list	<i>A print method for cpt lists</i>
----------------	-------------------------------------

Description

A print method for cpt lists

Usage

```
## S3 method for class 'cpt_list'  
print(x, ...)
```

Arguments

x	A cpt_list object
...	For S3 compatability. Not used.

See Also

[compile](#)

print.jt	<i>A print method for junction trees</i>
----------	--

Description

A print method for junction trees

Usage

```
## S3 method for class 'jt'  
print(x, ...)
```

Arguments

x	A junction tree object, jt.
...	For S3 compatability. Not used.

See Also

[jt](#)

propagate	<i>Propagation of junction trees</i>
-----------	--------------------------------------

Description

Given a junction tree object, propagation is conducted

Usage

```
propagate(x, prop = "full")

## S3 method for class 'jt'
propagate(x, prop = "full")
```

Arguments

x	A junction tree object jt
prop	Either "collect" or "full".

See Also

[jt](#)

Examples

```
# See Example 1 in the 'jt' function
```

query_belief	<i>Query probabilities</i>
--------------	----------------------------

Description

Get probabilities from a junction tree object

Usage

```
query_belief(x, nodes, type = "marginal")

## S3 method for class 'jt'
query_belief(x, nodes, type = "marginal")
```

Arguments

x	A junction tree object, jt.
nodes	The nodes for which the probability is desired
type	Either 'marginal' or 'joint'

See Also

[jt](#), [get_cliques](#), [leaves](#), [parents](#)

Examples

```
# See example 6 in the help page for the jt function
```

triangulate	<i>Triangulate a Bayesian network</i>
-------------	---------------------------------------

Description

Given a list of CPTs, this function finds a triangulation

Usage

```
triangulate(x, joint_vars = NULL, tri = "min_fill")
```

```
## S3 method for class 'cpt_list'
```

```
triangulate(x, joint_vars = NULL, tri = "min_fill")
```

Arguments

x	An object returned from <code>cpt_list</code>
joint_vars	A vector of variables for which we require them to be in the same clique. Edges between all these variables are added to the moralized graph.
tri	The optimization strategy used for triangulation. Either one of 'min_nei', 'min_fill', 'min_sp', 'sparse', 'minimal'

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