# Package 'dbnR'

June 19, 2024

```
Title Dynamic Bayesian Network Learning and Inference
Version 0.7.9
Description Learning and inference over dynamic Bayesian networks of arbitrary
      Markovian order. Extends some of the functionality offered by the 'bnlearn'
      package to learn the networks from data and perform exact inference.
      It offers three structure learning algorithms for dynamic Bayesian networks:
      Trabelsi G. (2013) <doi:10.1007/978-3-642-41398-8_34>, Santos F.P. and Maciel C.D. (2014)
      <a href="doi:10.1109/BRC.2014.6880957"><a href="doi:10.1109/BRC.2014.6880957"><a href="doi:10.1109/BRC.2014.6880957"></a>, Quesada D., Bielza C. and Larrañaga P. (2021)</a>
      <doi:10.1007/978-3-030-86271-8_14>. It also offers the possibility to perform
      forecasts of arbitrary length. A tool for visualizing the structure of the
      net is also provided via the 'visNetwork' package.
Depends R (>= 3.5.0), bnlearn (>= 4.5)
Imports data.table (>= 1.12.4), Rcpp (>= 1.0.2), magrittr (>= 1.5), R6
      (>= 2.4.1), stats (>= 3.6.0), MASS (>= 7.3-55)
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      3.6.0), graphics (>= 3.6.0), testthat (>= 2.1.0)
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```

Type Package

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AIC.dbn

Calculate the AIC of a dynamic Bayesian network

#### **Description**

Generic method for calculating the Akaike information criterion (AIC) of a "dbn" S3 object given some data. Calls bnlearn's AIC underneath.

#### Usage

```
## S3 method for class 'dbn'
AIC(object, ..., k)
```

#### **Arguments**

object the structure of the network

... additional parameters for the network scoring

k the penalty parameter

#### Value

the AIC score of the network

AIC.dbn.fit

Calculate the AIC of a dynamic Bayesian network

#### **Description**

Generic method for calculating the Akaike information criterion (AIC) of a "dbn.fit" S3 object given some data. Calls bnlearn's AIC underneath.

#### Usage

```
## S3 method for class 'dbn.fit'
AIC(object, ..., k)
```

#### **Arguments**

object the fitted network

... additional parameters for the network scoring

k the penalty parameter

#### Value

the AIC score of the network

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all.equal.dbn

Check if two network structures are equal to each other

# **Description**

Generic method for checking the equality of two "dbn" S3 objects. Calls bnlearn's all.equal underneath.

# Usage

```
## S3 method for class 'equal.dbn'
all(target, current, ...)
```

# Arguments

target "dbn" object

current the other "dbn" object additional parameters

#### Value

boolean result of the comparison

all.equal.dbn.fit

Check if two fitted networks are equal to each other

#### Description

Generic method for checking the equality of two "dbn.fit" S3 objects. Calls bnlearn's all.equal underneath.

# Usage

```
## S3 method for class 'equal.dbn.fit'
all(target, current, ...)
```

#### **Arguments**

target "dbn.fit" object

current the other "dbn.fit" object
... additional parameters

#### Value

boolean result of the comparison

as.character.dbn 5

as.character.dbn

Convert a network structure into a model string

#### **Description**

Generic method for converting a "dbn" S3 object into a string. Calls bnlearn's as.character underneath.

#### Usage

```
## S3 method for class 'dbn'
as.character(x, ...)
```

#### **Arguments**

x a "dbn" object

... additional parameters

#### Value

string representing the DBN model

BIC.dbn

Calculate the BIC of a dynamic Bayesian network

# **Description**

Generic method for calculating the Bayesian information criterion (BIC) of a "dbn" S3 object given some data. Calls bnlearn's BIC underneath.

# Usage

```
## S3 method for class 'dbn'
BIC(object, ...)
```

#### **Arguments**

object the structure of the network
... additional parameters for the network scoring

#### Value

the BIC score of the network

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BIC.dbn.fit

Calculate the BIC of a dynamic Bayesian network

#### **Description**

Generic method for calculating the Bayesian information criterion (BIC) of a "dbn.fit" S3 object given some data. Calls bnlearn's BIC underneath.

#### Usage

```
## S3 method for class 'dbn.fit'
BIC(object, ...)
```

#### **Arguments**

object the fitted network

... additional parameters for the network scoring

#### Value

the BIC score of the network

calc\_mu

Calculate the mu vector from a fitted BN or DBN

# Description

Given a "bn.fit" or a "dbn.fit" object, calculate the mu vector of the equivalent multivariate Gaussian distribution. Front end of a C++ function.

#### Usage

```
calc_mu(fit)
```

# Arguments

fit a bn.fit or dbn.fit object

#### Value

a named numeric vector of the means of each variable

calc\_sigma 7

# **Examples**

```
dt_train <- dbnR::motor[200:2500]
net <- bnlearn::mmhc(dt_train)
fit <- bnlearn::bn.fit(net, dt_train, method = "mle-g")
mu <- dbnR::calc_mu(fit)

f_dt_train <- dbnR::fold_dt(dt_train, size = 2)
net <- dbnR::learn_dbn_struc(dt_train, size = 2)
fit <- dbnR::fit_dbn_params(net, f_dt_train)
mu <- dbnR::calc_mu(fit)</pre>
```

calc\_sigma

Calculate the sigma covariance matrix from a fitted BN or DBN

#### Description

Given a "bn.fit" or a "dbn.fit" object, calculate the sigma covariance matrix of the equivalent multivariate Gaussian distribution. Front end of a C++ function.

#### **Usage**

```
calc_sigma(fit)
```

#### **Arguments**

fit

a bn.fit or dbn.fit object

#### Value

a named numeric covariance matrix of the nodes

#### **Examples**

```
dt_train <- dbnR::motor[200:2500]
net <- bnlearn::mmhc(dt_train)
fit <- bnlearn::bn.fit(net, dt_train, method = "mle-g")
sigma <- dbnR::calc_sigma(fit)

f_dt_train <- dbnR::fold_dt(dt_train, size = 2)
net <- dbnR::learn_dbn_struc(dt_train, size = 2)
fit <- dbnR::fit_dbn_params(net, f_dt_train)
sigma <- dbnR::calc_sigma(fit)</pre>
```

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coef.dbn.fit

Extracts the coefficients of a DBN

# Description

Generic method for "dbn.fit" S3 objects. Calls bnlearn underneath.

# Usage

```
## S3 method for class 'dbn.fit'
coef(object, ...)
```

# Arguments

object the fitted network
... additional parameters

#### Value

the coefficients of the network

degree

Calculates the degree of a list of nodes

# **Description**

#' Generic method for calculating the degree of a list of nodes in a BN or a DBN. Calls bnlearn's degree underneath. I have to redefine the generic and mask the original for it to work on both bn and dbn objects without the user having to import bnlearn.

#### Usage

```
degree(object, Nodes, ...)
```

#### **Arguments**

```
object a "bn", "dbn", "bn.fit" or "dbn.fit" object

Nodes which nodes to check
```

... additional parameters

#### Value

the degree of the nodes

filtered\_fold\_dt 9

filtered_fold_dt Fold a dataset avoiding overlapping of different time series	
---	--

#### **Description**

If the dataset that is going to be folded contains several different time series instances of the same process, folding it could introduce false rows with data from different time series. Given an id variable that labels the different instances of a time series inside a dataset and a desired size, this function folds the dataset and avoids mixing data from different origins in the same instance.

boolean that decides whether or not the id\_var column is deleted

#### Usage

```
filtered_fold_dt(dt, size, id_var, clear_id_var = TRUE)
```

# **Arguments**

dt data.table to be folded
size the size of the data.table
id\_var the variable that labels each individual instance of the time series

#### Value

the filtered data.table

clear\_id\_var

#### **Examples**

```
dt <- dbnR::motor[201:2500]
dt[, n_sec := rep(seq(46), each = 50)] # I'll create secuences of 50 instances each
f_dt <- dbnR::fold_dt(dt, size = 2)
dim(f_dt)
f_dt <- dbnR::filtered_fold_dt(dt, size = 2, id_var = "n_sec")
dim(f_dt) # The filtered folded dt has a row less for each independent secuence</pre>
```

filter\_same\_cycle

Filter the instances in a data.table with different ids in each row

#### **Description**

Given an id variable that labels the different instances of a time series inside a dataset, discard the rows that have values from more than 1 id.

#### Usage

```
filter_same_cycle(f_dt, size, id_var)
```

10 fitted.dbn.fit

#### **Arguments**

f\_dt folded data.table

size the size of the data.table

id\_var the variable that labels each individual instance of the time series

#### Value

the filtered data.table

# **Examples**

```
\label{eq:dt} $dt <-dbnR::motor[201:2500]$ $dt[, n_sec := rep(seq(46), each = 50)] $\# I'll create secuences of 50 instances each $f_dt <-dbnR::fold_dt(dt, size = 2)$ $f_dt[50, .SD, .SDcols = c("n_sec_t_0", "n_sec_t_1")]$ $f_dt <-dbnR::filter_same_cycle(f_dt, size = 2, id_var = "n_sec")$ $f_dt[50, .SD, .SDcols = c("n_sec_t_0", "n_sec_t_1")]$ $$
```

fitted.dbn.fit

Extracts the fitted values of a DBN

# **Description**

Generic method for "dbn.fit" S3 objects. Calls bnlearn underneath.

# Usage

```
## S3 method for class 'dbn.fit'
fitted(object, ...)
```

#### **Arguments**

object the fitted network
... additional parameters

#### Value

the fitted values of the network

fit\_dbn\_params 11

fit\_dbn\_params

Fits a markovian n DBN model

# Description

Fits the parameters of the DBN via MLE. The "mu" vector of means and the "sigma" covariance matrix are set as attributes of the dbn.fit object for future exact inference.

#### Usage

```
fit_dbn_params(net, f_dt, ...)
```

#### Arguments

```
net the structure of the DBN

f_dt a folded data.table

... additional parameters for the bn.fit function
```

#### Value

a "dbn.fit" S3 object with the fitted net

#### **Examples**

```
size = 3
dt_train <- dbnR::motor[200:2500]
net <- learn_dbn_struc(dt_train, size)
f_dt_train <- fold_dt(dt_train, size)
fit <- fit_dbn_params(net, f_dt_train, method = "mle-g")</pre>
```

fold\_dt

Widens the dataset to take into account the t previous time slices

#### **Description**

This function will widen the dataset to put the t previous time slices in each row, so that it can be used to learn temporal arcs in the second phase of the dmmhc.

#### Usage

```
fold_dt(dt, size)
```

#### **Arguments**

dt the data.table to be treated

size number of time slices to unroll. Markovian 1 would be size 2

forecast\_ts

#### Value

the extended data.table

# **Examples**

```
data(motor)
size <- 3
f_dt <- fold_dt(motor, size)</pre>
```

forecast\_ts

Performs forecasting with the GDBN over a dataset

# Description

Given a dbn.fit object, the size of the net and a folded dataset, performs a forecast over the initial evidence taken from the dataset.

#### Usage

```
forecast_ts(
   dt,
   fit,
   size = NULL,
   obj_vars,
   ini = 1,
   len = dim(dt)[1] - ini,
   rep = 1,
   num_p = 50,
   print_res = TRUE,
   plot_res = TRUE,
   mode = "exact",
   prov_ev = NULL
)
```

# Arguments

dt	data.table object with the TS data
fit	dbn.fit object
size	number of time slices of the net. Deprecated, will be removed in the future
obj_vars	variables to be predicted
ini	starting point in the dataset to forecast.
len	length of the forecast
rep	number of times to repeat the approximate forecasting
num_p	number of particles in the approximate forecasting
print_res	if TRUE prints the mae and sd metrics of the forecast

```
plot_res if TRUE plots the results of the forecast

mode "exact" for exact inference, "approx" for approximate

prov_ev variables to be provided as evidence in each forecasting step
```

#### Value

a list with the original time series values and the results of the forecast

#### **Examples**

```
generate_random_network_exp
```

Generate a random DBN and a sampled dataset

#### Description

This function generates both a random DBN and a dataset that can be used to learn its structure from data. It's intended for experimental use.

#### Usage

```
generate_random_network_exp(
    n_vars,
    size,
    min_mu,
    max_mu,
    min_sd,
    max_sd,
    min_coef,
    max_coef,
    seed = NULL
)
```

learn\_dbn\_struc

# **Arguments**

n_vars	number of desired variables per time-slice
size	desired size of the networks
min_mu	minimum mean allowed for the variables
max_mu	maximum mean allowed for the variables
min_sd	minimum standard deviation allowed for the variables
max_sd	maximum standard deviation allowed for the variables
min_coef	minimum coefficient allowed for the parent nodes
max_coef	maximum coefficient allowed for the parent nodes
seed	the seed of the experiment

#### Value

a list with the original network structure and the sampled dataset

learn\_dbn\_struc Learns the structure of a markovian n DBN model from data

# Description

Learns a gaussian dynamic Bayesian network from a dataset. It allows the creation of markovian n nets rather than only markov 1.

# Usage

```
learn_dbn_struc(dt, size = 2, method = "dmmhc", f_dt = NULL, ...)
```

#### **Arguments**

dt	the data.frame or data.table to be used
size	number of time slices of the net. Markovian 1 would be size 2
method	the structure learning method of choice to use
f_dt	previously folded dataset, in case some specific rows have to be removed after the folding
	additional parameters for rsmax2 function

#### Value

```
a "dbn" S3 object with the structure of the network
```

# Examples

```
data("motor")
net <- learn_dbn_struc(motor, size = 3)</pre>
```

logLik.dbn 15

logLik.dbn

Calculate the log-likelihood of a dynamic Bayesian network

#### **Description**

Generic method for calculating the log-likelihood of a "dbn" S3 object given some data. Calls bnlearn's logLik underneath.

# Usage

```
## S3 method for class 'dbn'
logLik(object, dt, ...)
```

#### Arguments

object the structure of the network

dt the dataset to calculate the score of the network ... additional parameters for the network scoring

#### Value

the log-likelihood score of the network

logLik.dbn.fit

Calculate the log-likelihood of a dynamic Bayesian network

#### Description

Generic method for calculating the log-likelihood of a "dbn.fit" S3 object given some data. Calls bnlearn's logLik underneath.

# Usage

```
## S3 method for class 'dbn.fit'
logLik(object, dt, ...)
```

#### **Arguments**

object the fitted network

dt the dataset to calculate the score of the network ... additional parameters for the network scoring

#### Value

the log-likelihood score of the network

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mean.dbn.fit	Average the parameters of multiple dbn.fit objects with identical structures

#### **Description**

Generic method for "dbn.fit" S3 objects. Calls bnlearn underneath.

# Usage

```
## S3 method for class 'dbn.fit' mean(x, ...)
```

#### **Arguments**

x the fitted network... additional parameters

# Value

the averaged parameters

motor	Multivariate time series dataset on the temperature of an electric mo-
	tor

# **Description**

Data from several sensors on an electric motor that records different benchmark sessions of measurements at 2 Hz. The dataset is reduced to 3000 instances from the 60th session in order to include it in the package for testing purposes. For the complete dataset, refer to the source.

#### Usage

```
data(motor)
```

#### **Format**

An object of class data.table (inherits from data.frame) with 3000 rows and 11 columns.

#### **Source**

Kaggle, <a href="https://www.kaggle.com/wkirgsn/electric-motor-temperature">https://www.kaggle.com/wkirgsn/electric-motor-temperature</a>

mvn\_inference 17

mvn\_inference

Performs inference over a multivariate normal distribution

#### **Description**

Given some evidence, this function performs inference over a multivariate normal distribution. After converting a Gaussian linear network to its MVN form, this kind of inference can be performed. It's recommended to use predict\_dt functions instead unless you need a more flexible inference method.

#### Usage

```
mvn_inference(mu, sigma, evidence)
```

#### **Arguments**

mu the mean vector

sigma the covariance matrix

evidence a single row data.table or a named vector with the values and names of the

variables given as evidence

#### Value

a list with the posterior mean and covariance matrix

#### **Examples**

```
size = 3
data(motor)
dt_train <- motor[200:2500]
dt_val <- motor[2501:3000]
obj <- c("pm_t_0")

net <- learn_dbn_struc(dt_train, size)
f_dt_train <- fold_dt(dt_train, size)
f_dt_val <- fold_dt(dt_val, size)
ev <- f_dt_val[1, .SD, .SDcols = obj]
fit <- fit_dbn_params(net, f_dt_train, method = "mle-g")

pred <- mvn_inference(calc_mu(fit), calc_sigma(fit), ev)</pre>
```

18 nodes<-

nodes

Returns a list with the names of the nodes of a BN or a DBN

#### **Description**

Generic method for obtaining the names of the nodes in a BN or a DBN. Calls bnlearn's nodes underneath. I have to redefine the generic and mask the original for it to work on both bn and dbn objects without the user having to import bnlearn.

#### Usage

```
nodes(object, ...)
```

#### **Arguments**

```
object a "bn", "dbn", "bn.fit" or "dbn.fit" object
... additional parameters
```

# Value

the names of the nodes

nodes<-

Relabel the names of the nodes of a BN or a DBN

# Description

Generic method for renaming the nodes in a BN or a DBN. Calls bnlearn's nodes<- underneath. I have to redefine the generic and mask the original for it to work on both bn and dbn objects without the user having to import bnlearn.

#### Usage

```
nodes(object) <- value</pre>
```

#### **Arguments**

```
object a "bn", "dbn", "bn.fit" or "dbn.fit" object
```

value a list with the new names

#### Value

the modified object

plot.dbn 19

plot.dbn

Plots a dynamic Bayesian network

# Description

Generic method for plotting the "dbn" S3 objects. Calls plot\_dynamic\_network underneath.

#### Usage

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

#### **Arguments**

x the structure of the network.

... additional parameters for the visualization of a DBN

plot.dbn.fit

Plots a fitted dynamic Bayesian network

# Description

Generic method for plotting the "dbn.fit" S3 objects. Calls plot\_dynamic\_network underneath.

# Usage

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

# Arguments

x the structure of the network.

... additional parameters for the visualization of a DBN

plot\_dynamic\_network Plots a dynamic Bayesian network in a hierarchical way

#### **Description**

To plot the DBN, this method first computes a hierarchical structure for a time slice and replicates it for each slice. Then, it calculates the relative position of each node with respect to his equivalent in the first slice. The result is a net where each time slice is ordered and separated from one another, where the leftmost slice is the oldest and the rightmost represents the present time. This function is also called by the generic plot function of "dbn" and "dbn.fit" S3 objects.

#### Usage

```
plot_dynamic_network(
   structure,
   offset = 200,
   subset_nodes = NULL,
   reverse = FALSE
)
```

#### **Arguments**

structure the structure or fit of the network.

offset the blank space between time slices

subset\_nodes a vector containing the names of the subset of nodes to plot

reverse reverse to the classic naming convention of the nodes. The oldest time-slice will

now be t\_0 and the most recent one t\_n. Only for visualization purposes, the network is unmodified underneath. If using subset\_nodes, remember that t\_0 is

now the oldest time-slice.

#### Value

the visualization of the DBN

#### **Examples**

```
size = 3
dt_train <- dbnR::motor[200:2500]
net <- learn_dbn_struc(dt_train, size)
plot_dynamic_network(net)</pre>
```

plot\_static\_network 21

plot\_static\_network

Plots a Bayesian network in a hierarchical way

#### **Description**

This function calculates the levels of each node and then plots them in a hierarchical layout in visNetwork. Can be used in place of the generic plot function offered by bnlearn for "bn" and "bn.fit" S3 objects.

#### Usage

```
plot_static_network(structure)
```

#### **Arguments**

structure

the structure or fit of the network.

# **Examples**

```
dt_train <- dbnR::motor[200:2500]
net <- bnlearn::mmhc(dt_train)
plot_static_network(net)
fit <- bnlearn::bn.fit(net, dt_train, method = "mle-g")
plot_static_network(fit) # Works for both the structure and the fitted net</pre>
```

predict.dbn.fit

Performs inference in every row of a dataset with a DBN

#### **Description**

Generic method for predicting a dataset with a "dbn.fit" S3 objects. Calls predict\_dt underneath.

#### Usage

```
## S3 method for class 'dbn.fit'
predict(object, ...)
```

#### **Arguments**

```
object a "dbn.fit" object
```

... additional parameters for the inference process

#### Value

a data.table with the prediction results

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predict\_bn

Performs inference over a fitted GBN

#### Description

Performs inference over a Gaussian BN. It's thought to be used in a map for a data.table, to use as evidence each separate row. If not specifically needed, it's recommended to use the function predict\_dt instead. This function is deprecated and will be removed in a future version.

#### Usage

```
predict_bn(fit, evidence)
```

# Arguments

fit the fitted bn

evidence values of the variables used as evidence for the net

#### Value

a data.table with the predictions

#### **Examples**

```
size = 3
data(motor)
dt_train <- motor[200:2500]
dt_val <- motor[2501:3000]
net <- learn_dbn_struc(dt_train, size)
f_dt_train <- fold_dt(dt_train, size)
f_dt_val <- fold_dt(dt_val, size)
fit <- fit_dbn_params(net, f_dt_train, method = "mle-g")
res <- f_dt_val[, predict_bn(fit, .SD), .SDcols = c("pm_t_0", "coolant_t_0"), by = 1:nrow(f_dt_val)]</pre>
```

predict\_dt

Performs inference over a test dataset with a GBN

#### **Description**

This function performs inference over each row of a folded data.table, plots the results and gives metrics of the accuracy of the predictions. Given that only a single row is predicted, the horizon of the prediction is at most 1. This function is also called by the generic predict method for "dbn.fit" objects. For long term forecasting, please refer to the forecast\_ts function.

#### Usage

```
predict_dt(fit, dt, obj_nodes, verbose = T, look_ahead = F)
```

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# **Arguments**

fit	the fitted bn
dt	the test dataset
obj_nodes	the nodes that are going to be predicted. They are all predicted at the same time
verbose	if TRUE, displays the metrics and plots the real values against the predictions
look_ahead	boolean that defines whether or not the values of the variables in t_0 should be used when predicting, even if they are not present in obj_nodes. This decides if look-ahead bias is introduced or not.

#### Value

a data.table with the prediction results for each row

#### **Examples**

```
size = 3
data(motor)
dt_train <- motor[200:900]</pre>
dt_val <- motor[901:1000]</pre>
# With a DBN
obj <- c("pm_t_0")
net <- learn_dbn_struc(dt_train, size)</pre>
f_dt_train <- fold_dt(dt_train, size)</pre>
f_dt_val <- fold_dt(dt_val, size)</pre>
fit <- fit_dbn_params(net, f_dt_train, method = "mle-g")</pre>
res <- suppressWarnings(predict_dt(fit, f_dt_val, obj_nodes = obj, verbose = FALSE))</pre>
# With a Gaussian BN directly from bnlearn
obj <- c("pm")
net <- bnlearn::mmhc(dt_train)</pre>
fit <- bnlearn::bn.fit(net, dt_train, method = "mle-g")</pre>
res <- suppressWarnings(predict_dt(fit, dt_val, obj_nodes = obj, verbose = FALSE))</pre>
```

print.dbn

Print method for "dbn" objects

# Description

Generic print method for "dbn" S3 objects. Calls bnlearn's print underneath

# Usage

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

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# **Arguments**

x the "dbn" object... additional parameters

print.dbn.fit

Print method for "dbn.fit" objects

# Description

Generic print method for "dbn.fit" S3 objects. Calls bnlearn's print underneath

# Usage

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

# Arguments

x the "dbn.fit" object
... additional parameters

rbn.dbn.fit

Simulates random samples from a fitted DBN

# Description

Generic method for "dbn.fit" S3 objects. Calls bnlearn's rbn underneath.

#### Usage

```
rbn.dbn.fit(x, n, ...)
```

# Arguments

x the fitted networkn number of samples... additional parameters

#### Value

the sampled dataset

reduce\_freq 25

reduce_freq	Reduce the frequency of the time series data in a data.table
-------------	--

#### **Description**

In a time series dataset, there is a time difference between one row and the next one. This function reduces the number of rows from its current frequency to the desired one by averaging batches of rows. Instead of the frequency in Hz, the number of seconds between rows is asked (Hz = 1/s).

#### Usage

```
reduce_freq(dt, obj_freq, curr_freq, id_var = NULL)
```

#### Arguments

dt the original data.table

obj\_freq the desired number of seconds between rows

curr\_freq the number of seconds between rows in the original dataset

id\_var optional variable that labels different time series in a dataset, to avoid averaging

values from different processes

#### Value

the data.table with the desired frequency

#### **Examples**

```
# Let's assume that the dataset has a frequency of 4Hz, 0.25 seconds between rows
dt <- dbnR::motor
dim(dt)
# Let's change the frequency to 2Hz, 0.5 seconds between rows
dt <- reduce_freq(dt, obj_freq = 0.5, curr_freq = 0.2)
dim(dt)</pre>
```

residuals.dbn.fit

Returns the residuals from fitting a DBN

#### **Description**

Generic method for "dbn.fit" S3 objects. Calls bnlearn underneath.

#### Usage

```
## S3 method for class 'dbn.fit'
residuals(object, ...)
```

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#### **Arguments**

object the fitted network
... additional parameters

#### Value

the residuals of fitting the network

score

Computes the score of a BN or a DBN

# Description

Generic method for computing the score of a BN or a DBN. Calls bnlearn's nodes underneath. I have to redefine the generic and mask the original for it to work on both bn and dbn objects without the user having to import bnlearn.

#### Usage

```
score(object, ...)
```

#### **Arguments**

```
object a "bn" or "dbn" object additional parameters
```

#### Value

the score of the network

shift\_values

Move the window of values backwards in a folded dataset row

#### Description

This function moves the values in t\_0, t\_1, ..., t\_n-1 in a folded dataset row to t\_1, t\_2, ..., t\_n. All the variables in t\_0 will be inputed with NAs and the obtained row can be used to forecast up to any desired point.

# Usage

```
shift_values(f_dt, row)
```

sigma.dbn.fit 27

# Arguments

f\_dt a folded dataset

row the index of the row that is going to be processed

#### Value

a one row data.table the shifted values

# **Examples**

```
dt <- dbnR::motor
f_dt <- dbnR::fold_dt(dt, size = 2)
s_row <- dbnR::shift_values(f_dt, row = 500)</pre>
```

sigma.dbn.fit

Returns the standard deviation of the residuals from fitting a DBN

# Description

Generic method for "dbn.fit" S3 objects. Calls bnlearn underneath.

#### Usage

```
## S3 method for class 'dbn.fit'
sigma(object, ...)
```

# Arguments

```
object the fitted network
... additional parameters
```

#### Value

the standard deviation residuals of fitting the network

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 $smooth\_ts$ 

Performs smoothing with the GDBN over a dataset

#### **Description**

Given a dbn.fit object, the size of the net and a folded dataset, performs a smoothing of a trajectory. Smoothing is the opposite of forecasting: given a starting point, predict backwards in time to obtain the time series that generated that point.

#### Usage

```
smooth_ts(
   dt,
   fit,
   size = NULL,
   obj_vars,
   ini = dim(dt)[1],
   len = ini - 1,
   print_res = TRUE,
   plot_res = TRUE,
   prov_ev = NULL
)
```

# Arguments

dt	data.table object with the TS data
fit	dbn.fit object
size	number of time slices of the net. Deprecated, will be removed in the future
obj_vars	variables to be predicted. Should be in the oldest time step
ini	starting point in the dataset to smooth
len	length of the smoothing
print_res	if TRUE prints the mae and sd metrics of the smoothing
plot_res	if TRUE plots the results of the smoothing
prov_ev	variables to be provided as evidence in each smoothing step. Should be in the oldest time step

#### Value

a list with the original values and the results of the smoothing

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#### **Examples**

time\_rename

Renames the columns in a data.table so that they end in '\_t\_0'

# Description

This will rename the columns in a data.table so that they end in '\_t\_0', which will be needed when folding the data.table. If any of the columns already ends in '\_t\_0', a warning will be issued and no further operation will be done. There is no need to use this function to learn a DBN unless some operation with the variable names wants to be done prior to folding a dataset.

#### Usage

```
time_rename(dt)
```

# **Arguments**

dt

the data.table to be treated

#### Value

the renamed data.table

#### **Examples**

```
data("motor")
dt <- time_rename(motor)</pre>
```

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[[<-.dbn.fit

Replacement function for parameters inside DBNs

#### **Description**

Generic parameter replacement method for "dbn.fit" S3 objects. Calls bnlearn underneath.

#### Usage

```
## S3 replacement method for class 'dbn.fit' x[[name]] \leftarrow value
```

# Arguments

x the fitted network

name of the node to replace its parameters

value the new parameters

#### Value

the modified network

\$<-.dbn.fit</pre>

Replacement function for parameters inside DBNs

#### **Description**

Generic parameter replacement method for "dbn.fit" S3 objects. Calls bnlearn underneath.

#### Usage

```
## S3 replacement method for class 'dbn.fit' xname <- value
```

# Arguments

x the fitted network

name of the node to replace its parameters

value the new parameters

#### Value

the modified network

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