

Package ‘ts2net’

October 14, 2022

Title From Time Series to Networks

Version 0.1.0

Description Transforming one or multiple time series into networks. This package is useful for complex systems modeling, time series data mining, or time series analysis using networks.

An introduction to the topic and the descriptions of the methods implemented in this package can be found in Mitchell (2006) <[doi:10.1016/j.artint.2006.10.002](https://doi.org/10.1016/j.artint.2006.10.002)>, Silva and Zhao (2016) <[doi:10.1007/978-3-319-17290-3](https://doi.org/10.1007/978-3-319-17290-3)>, and Silva et al. (2021) <[doi:10.1002/widm.1404](https://doi.org/10.1002/widm.1404)>.

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URL <https://github.com/lferreira/ts2net>

BugReports <https://github.com/lferreira/ts2net/issues>

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dataset_sincos_generate

Sin-Cos data set generator. This function generates a set of sine and cosine time series. This function is used as example of the package application.

Description

Sin-Cos data set generator. This function generates a set of sine and cosine time series. This function is used as example of the package application.

Usage

```
dataset_sincos_generate(
  num_sin_series = 25,
  num_cos_series = 25,
  x_max = 8 * pi,
  ts_length = 100,
  jitter_amount = 0.1,
  return_x_values = FALSE
)
```

Arguments

`num_sin_series` Integer. Number of sine time series

`num_cos_series` Integer. Number of cosine time series

`x_max` Float. Max x value in $\sin(x)$ or $\cos(x)$.

`ts_length` Integer. Time series length.

`jitter_amount` Float. The total amount of jitter added to each time series.

`return_x_values` Boolean. If positive, returns a list of data frames with x and y values.

Value

A list with all time series. First the `num_sin_series` sine time series followed by the `num_cos_series` cosine time series.

`dist_file_parts_merge` *Merge parts of distances stored in files.*

Description

The functions `tsdist_dir_parallel` and `tsdist_parts_parallel` calculate part of the distance matrix D. The results of the multiple calls of these functions are normally stored in RDS or csv files. This function merges these files and construct a distance matrix D.

Usage

```
dist_file_parts_merge(list_files, dir_path, num_elements, file_type = "RDS")
```

Arguments

`list_files` A list of files with distances.

`dir_path` If `list_files` was not passed, than this function uses this parameter to read the files in this directory.

`num_elements` The number of time series in the data set. The number of elements defines the number of rows and columns in the distance matrix D.

`file_type` The extension of the files where the distances are stored. It can be "RDS" (default) or "csv". The RDS files should be data frames composed by three columns `i,j`, and `dist`. This format is preferred because it is a compact file. The other option is a "csv" also containing the `i,j`, and `dist` columns.

Value

Distance matrix `D`

`dist_matrix_normalize` *Normalize a distance/similarity matrix.*

Description

Normalize a distance/similarity matrix.

Usage

```
dist_matrix_normalize(D, to = c(0, 1))
```

Arguments

`D` Distance/similarity matrix

`to` An array of two elements `c(min_value, max_value)` representing the interval where the elements of `dist_matrix` will be normalized to.

Value

Normalized matrix

`dist_parts_merge` *Merge parts of distances stored in data frames.*

Description

The functions `tsdist_dir_parallel` and `tsdist_parts_parallel` calculate part of the distance matrix `D`. This function merges these files and construct a distance matrix `D`.

Usage

```
dist_parts_merge(list_dfs, num_elements)
```

Arguments

`list_dfs` A list of data frames. Each data frame should have three columns `i,j`, and `dist`.

`num_elements` The number of time series in the data set. The number of elements defines the number of rows and columns in the distance matrix `D`.

Value

Distance matrix D

dist_percentile	<i>Returns the distance value that corresponds to the desired percentile. This function is useful when the user wants to generate networks with different distance functions but with the same link density.</i>
-----------------	--

Description

Returns the distance value that corresponds to the desired percentile. This function is useful when the user wants to generate networks with different distance functions but with the same link density.

Usage

```
dist_percentile(D, percentile = 0.1, is_D_symetric = TRUE)
```

Arguments

D	distance matrix
percentile	(Float) The desired percentile of lower distances.
is_D_symetric	(Boolean)

Value

Distance percentile value.

events_from_ts	<i>Extract events from a time series.</i>
----------------	---

Description

This function transforms an time series (array) into a binary time series where 1 means a event and 0 means no event.

Usage

```
events_from_ts(
  ts,
  th,
  method = c("greater_than", "lower_than", "top_percentile", "lower_percentile",
    "highest", "lowest"),
  return_marked_times = FALSE
)
```

Arguments

ts	Array. Time series
th	A threshold (if 'method=greater_than' or '=lower_than'), or the percentile (if 'method=top_percentile' or '=lower_percentile'), or the total number (if 'method=highest' or '=lowest').
method	String. One of following options: * 'greater_than': All values greater or equal to 'th'. * 'lower_than': All values lower or equal to 'th'. * 'top_percentile': Values greater than the 'th' percentile. * 'highest': The top 'th' values. * 'lowest': The lower 'th' values.
return_marked_times	Return the time indices (marked points) where the events occur.

Value

An event (binary, 1: event, 0 otherwise) time series

net_enn	<i>Construct an epsilon-network from a distance matrix.</i>
---------	---

Description

Construct an epsilon-network from a distance matrix.

Usage

```
net_enn(
  D,
  eps,
  treat_NA_as = 1,
  is_dist_symetric = TRUE,
  weighted = FALSE,
  invert_dist_as_weight = TRUE,
  add_col_rownames = TRUE
)
```

Arguments

D	Distance matrix
eps	the threshold value to be considered a link. Only values lower or equal to epsilon become 1.
treat_NA_as	A numeric value, usually 1, that represent NA values in the distance matrix
is_dist_symetric	Boolean, TRUE (default) if dist is symmetric
weighted	Boolean, TRUE will create a weighted network

invert_dist_as_weight

Boolean, if weighted == TRUE, then the weights become 1 - distance. This is the default behavior since most network measures interpret higher weights as stronger connection.

add_col_rownames

Boolean. If TRUE (default), it uses the column and row names from dist matrix as node labels.

Value

a igraph network

net_enn_approx	<i>Construct an approximated epsilon neighbor network (faster, but approximated) from a distance matrix. Some actual nearest neighbors may be omitted.</i>
----------------	--

Description

Construct an approximated epsilon neighbor network (faster, but approximated) from a distance matrix. Some actual nearest neighbors may be omitted.

Usage

```
net_enn_approx(D, eps, ...)
```

Arguments

D	Distance matrix
eps	(Integer) k nearest-nearest neighbors where each time series will be connected to
...	Other parameters to frNN() function from dbscan package.

Value

Approximated epsilon nearest-neighbor network

net_knn	<i>Construct a knn-network from a distance matrix.</i>
---------	--

Description

Construct a knn-network from a distance matrix.

Usage

```
net_knn(D, k, num_cores = 1)
```

Arguments

D	Distance matrix
k	(Integer) k nearest-nearest neighbors where each time series will be connected to
num_cores	(Integer) Number of cores to use.

Value

k nearest-neighbor network

net_knn_approx	<i>Construct an approximated knn-network (faster, but approximated) from a distance matrix.</i>
----------------	---

Description

Construct an approximated knn-network (faster, but approximated) from a distance matrix.

Usage

```
net_knn_approx(D, k, ...)
```

Arguments

D	Distance matrix
k	(Integer) k nearest-nearest neighbors where each time series will be connected to
...	Other parameters to kNN() function from dbscan package.

Value

Approximated k nearest-neighbor network

net_weighted	<i>Creates a weighted network.</i>
--------------	------------------------------------

Description

A link is created for each pair of nodes, except if the distance is maximum (1). In network science, stronger links are commonly represented by high values. For this reason, the link weights returned are $1 - D$.

Usage

```
net_weighted(D, invert_dist_as_weight = TRUE)
```

Arguments

D	Distance matrix. All values must be between [0,1].
invert_dist_as_weight	Boolean, if weighted == TRUE, then the weights become $1 - \text{distance}$. This is the default behavior since most network measures interpret higher weights as stronger connection.

Value

Fully connected network

random_ets	<i>Random event time series generator</i>
------------	---

Description

It generates an event time series with length `ts_length` with `num_events` events considering a uniform probability distribution.

Usage

```
random_ets(ts_length, num_events, return_marked_times = FALSE)
```

Arguments

ts_length	Time series Length
num_events	The number of events
return_marked_times	Return the time indices (marked points) where the events occur.

Value

An event (binary, 1: event, 0 otherwise) time series

tsdist_ccf	<i>Cross-correlation distance</i>
------------	-----------------------------------

Description

Minimum correlation distance considering a +/- max lag (lag_max)

Usage

```
tsdist_ccf(
  ts1,
  ts2,
  type = c("correlation", "covariance"),
  cor_type = "abs",
  directed = FALSE,
  lag_max = 10,
  return_lag = FALSE
)
```

Arguments

ts1	Array. Time series 1
ts2	Array. Time series 2
type	String. "correlation" or "covariance" to be used (type) in the ccf function.
cor_type	String. "abs" (default), "+", or "-". "abs" considers the correlation absolute value. "+" only positive correlations and "-" only negative correlations.
directed	Boolean. If FALSE (default), the lag interval [-lag_max,+lag_max] is considered. Otherwise, [-lag_max,0] is considered.
lag_max	Integer. Default = 10.
return_lag	Also returns the time lag that leads to the shortest distances.

Value

Distance

tsdist_cor	<i>Absolute, positive, or negative correlation distance.</i>
------------	--

Description

Considering r the person correlation coefficient, this function returns either $1 - \text{abs}(r)$ if `cor_type=="abs"`, $1 - \text{pmax}(0, r)$ if `cor_type == "+"`, or $1 - \text{pmax}(0, r * -1)$ if `cor_type == "-"`. Another possibility is to run a significance test to verify if the r is significant.

Usage

```
tsdist_cor(ts1, ts2, cor_type = "abs", sig_test = FALSE, sig_level = 0.01, ...)
```

Arguments

ts1	Array. Time series 1
ts2	Array. Time series 2
cor_type	String. "abs" (default), "+", or "-". "abs" considers the correlation absolute value. "+" only positive correlations and "-" only negative correlations.
sig_test	Run a statistical test. Return 0 if significant or 1 otherwise.
sig_level	The significance level to test if correlation is significant.
...	Additional parameters to cor.test() function.

Value

Real value [0,1] where 0 means perfect positive (or negative if positive_cor==FALSE) correlation and 1 no positive (or negative if positive_cor==FALSE) correlation.

tsdist_dtw	<i>Dynamic Time Warping (DTW) distance.</i>
------------	---

Description

This function is a wrapper for the dtw() function from the dtw package.

Usage

```
tsdist_dtw(ts1, ts2, ...)
```

Arguments

ts1	Array. Time series 1
ts2	Array. Time series 2
...	Additional parameters for the dtw() function from the dtw package.

Value

DTW distance

 tsdist_es

Event synchronization distance test.

Description

Quiroga, R. Q., Kreuz, T., & Grassberger, P. (2002). Event synchronization: a simple and fast method to measure synchronicity and time delay patterns. *Physical review E*, 66(4), 041904.

Usage

```
tsdist_es(
  ets1,
  ets2,
  tau_max = +Inf,
  method = c("quiroga", "boers"),
  sig_test = FALSE,
  reps = 100,
  sig_level = 0.01
)
```

Arguments

ets1	Event time series 1 (one means an event, or zero otherwise)
ets2	Event time series 2 (one means an event, or zero otherwise)
tau_max	The maximum tau allowed ()
method	"quiroga" (default) for the default co-occurrence count and normalization or "boers" for the co-occurrence count with tau_max and no normalization.
sig_test	Run a significance test. Return 0 if significant or 1 otherwise.
reps	Number of repetitions to construct the confidence interval
sig_level	The significance level to test if correlation is significant.

Details

Boers, N., Goswami, B., Rheinwalt, A., Bookhagen, B., Hoskins, B., & Kurths, J. (2019). Complex networks reveal global pattern of extreme-rainfall teleconnections. *Nature*, 566(7744), 373-377.

Value

distance

tsdist_mic	<i>Maximal information coefficient (MIC) distance.</i>
------------	--

Description

This function transforms the MIC function (from minerva package) into a distance function.

Usage

```
tsdist_mic(ts1, ts2)
```

Arguments

ts1	Array. Time series 1
ts2	Array. Time series 2

Value

Distance

tsdist_nmi	<i>Normalized mutual information distance</i>
------------	---

Description

Calculates the normalized mutual information (NMI) and returns it as distance $1 - \text{NMI}$.

Usage

```
tsdist_nmi(  
  ts1,  
  ts2,  
  nbins = c("sturges", "freedman-diaconis", "scott"),  
  normalization = c("sum", "min", "max", "sqrt"),  
  method = "emp"  
)
```

Arguments

ts1	Array. Time series 1
ts2	Array. Time series 2
nbins	The number of bins used for the discretization of both time series. It can be a positive integer or a string with one of the three rules "sturges" (default), "freedman-diaconis", or "scott".

normalization	The mutual information (I) normalization method. Options are "sum" (default) $1-(2I/(h1+h2))$, "min" $1-(I/\min(h1,h2))$, "max" $1-(I/\max(h1,h2))$, and "sqrt" $1-(I/\sqrt{h1*h2})$.
method	The name of the entropy estimator used in the functions <code>mutinformation()</code> and <code>entropy()</code> from the <code>infotheo</code> package.

Value

Distance

tsdist_voi	<i>Variation of Information distance</i>
------------	--

Description

The variation of information (VoI) is a distance function based on mutual information.

Usage

```
tsdist_voi(
  ts1,
  ts2,
  nbins = c("sturges", "freedman-diaconis", "scott"),
  method = "emp"
)
```

Arguments

ts1	Array. Time series 1
ts2	Array. Time series 2
nbins	The number of bins used for the discretization of both time series. It can be a positive integer or a string with one of the three rules "sturges" (default), "freedman-diaconis", or "scott".
method	The name of the entropy estimator used in the functions <code>mutinformation()</code> and <code>entropy()</code> from the <code>infotheo</code> package.

Value

Distance

tsdist_vr	<i>van Rossum distance</i>
-----------	----------------------------

Description

This function compares the times which the events occur e.g., time indices where the time series values are different than zero. Note that the intensity does not matter but if there is an event or not. This function also performs a statistical test using a shuffling approach to test significance. This implementation uses the fmetric function from the mmpp package.

Usage

```
tsdist_vr(ets1, ets2, tau = 1, sig_test = FALSE, reps = 100, sig_level = 0.01)
```

Arguments

ets1	Event time series 1 (one means an event, or zero otherwise)
ets2	Event time series 2 (one means an event, or zero otherwise)
tau	Parameter for filtering function (See fmetric function from mmpp package.)
sig_test	Run a statistical test. Return 0 if significant or 1 otherwise.
reps	Number of repetitions to construct the confidence interval
sig_level	The significance level to test if correlation is significant.

Value

distance

tsnet_rn	<i>Construct the recurrence network from a time series.</i>
----------	---

Description

This function constructs the recurrence matrix of the time series using the function 'rqa()' from **nonlinearTseries** package.

Usage

```
tsnet_rn(x, radius, embedding.dim, time.lag = 1, do.plot = FALSE, ...)
```

Arguments

<code>x</code>	Array. Time series
<code>radius</code>	Maximum distance between two phase-space points to be considered a recurrence.
<code>embedding.dim</code>	Integer denoting the dimension in which we shall embed the time.series. If missing, the embedding dimensions is estimated using ‘estimateEmbeddingDim()’ from nonlinearTseries . The constructed igraph network has the estimated dimension (and other info) as a parameter. For example: <code>net\$embedding_dim</code>
<code>time.lag</code>	Integer denoting the number of time steps that will be use to construct the Takens’ vectors.
<code>do.plot</code>	Boolean. Show recurrence plot (default = FALSE)
<code>...</code>	Other parameters to ‘rqa()’ from nonlinearTseries

Value

recurrence network

tsnet_vg

Construct the visibility graph from a time series

Description

TODO: weights

Usage

```
tsnet_vg(x, method = c("nvg", "hvg"), limit = +Inf, num_cores = 1)
```

Arguments

<code>x</code>	Array. Time series
<code>method</code>	String. Construction method: "nvg" (default) for Natural visibility graph, "hvg" horizontal visibility graph.
<code>limit</code>	Positive integer. The maximum temporal distance (indexes) allowed in the visibility. This parameter limits the max visibility.
<code>num_cores</code>	Number of cores (default = 1).

Value

visibility graph

tssim_event_sync	<i>Event synchronization measure</i>
------------------	--------------------------------------

Description

This function is an adapted version of the coocmetric function from the package mmpp. The differences are the introduction of a tau_max limitation factor and the optional normalization.

Usage

```
tssim_event_sync(  
  tts1,  
  tts2,  
  tau_max = 1,  
  normalization = c("both", "min", "none")  
)
```

Arguments

tts1	Time indices marking events in time series 1
tts2	Time indices marking events in time series 2
tau_max	Max tau to be considered
normalization	Forms of normalization after the co-occurrence count. Possible values "both" (default), "min", and "none". The Default is "both", the original normalization defined by Quiroga et al: $\sqrt{N1*N2}$. This normalization might be problematic when both time series have very different number of events. Another possibility is to normalize the count by the "min" length between both series. The interpretation now takes into account only the series with less events. For example, considering two series, one with many events and another with just a single event, the results can be 1 (total sync). The option "none" means no normalization and the method returns the total count of synchronized events.

Details

Quiroga, R. Q., Kreuz, T., & Grassberger, P. (2002). Event synchronization: a simple and fast method to measure synchronicity and time delay patterns. *Physical review E*, 66(4), 041904.

Boers, N., Goswami, B., Rheinwalt, A., Bookhagen, B., Hoskins, B., & Kurths, J. (2019). Complex networks reveal global pattern of extreme-rainfall teleconnections. *Nature*, 566(7744), 373-377.

Value

Synchronization-based similarity

`ts_dist`*Calculate distances between pairs of time series in a list.*

Description

This function calculates the distance between all combinations of time series in the list and returns a distance matrix. This function is usually the first try and might work if the number of time series and their length are not too big.

Usage

```
ts_dist(  
  ts_list,  
  dist_func = tsdist_cor,  
  is_symetric = TRUE,  
  error_value = NaN,  
  warn_error = TRUE,  
  num_cores = 1,  
  ...  
)
```

Arguments

<code>ts_list</code>	List of time series (arrays).
<code>dist_func</code>	Function to be applied to all combinations of time series. This function should have at least two parameters for each time series. Ex: <code>function(ts1, ts2) cor(ts1, ts2)</code>
<code>is_symetric</code>	Boolean. If the distance function is symmetric.
<code>error_value</code>	The value returned if an error occur when calculating a the distance for a pair of time series.
<code>warn_error</code>	Boolean. If TRUE (default), a warning will rise when an error occur during the calculations.
<code>num_cores</code>	Numeric. Number of cores
<code>...</code>	Additional parameters for <code>measureFunc</code>

Value

A distance or similarity matrix M whose position M_{ij} corresponds to distance or similarity value between time series i and j .

ts_dist_part	<i>Calculate distances between pairs of time series in part of a list.</i>
--------------	--

Description

This function is particularly useful to run in parallel as jobs in a cluster (HPC). It returns a data frame with elements (i,j) and a distance value calculated for the time series i and j. Not all the elements are calculated but just a part of the total combinations of time series in the list. This function load all the time series in the memory to make the calculations faster. However, if the time series are too long and/or the dataset is huge, it might represent a memory problem. In this case, `dist_dir_parallel()` is more recommended.

Usage

```
ts_dist_part(
  ts_list,
  num_part,
  num_total_parts,
  combinations,
  dist_func = tsdist_cor,
  isSymetric = TRUE,
  error_value = NaN,
  warn_error = TRUE,
  simplify = TRUE,
  num_cores = 1,
  ...
)
```

Arguments

ts_list	List of time series.
num_part	Numeric positive between 1 and the total number of parts (<code>num_total_parts</code>). This value corresponds to the part (chunk) of the total number of parts to be calculated.
num_total_parts	Numeric positive corresponding the total number of parts.
combinations	A list composed by arrays of size 2 indicating the files indices to be compared. If this parameter is passed, then the function does not split all the possibilities and does not use the parameters <code>num_part</code> and <code>num_total_parts</code> . This parameter is useful when the number of combinations is very high and this functions is called several times (high <code>num_total_parts</code>). In this case, instead of calculating all the combinations in each call, the user can calculate it once and pass it via this parameter.
dist_func	Function to be applied to all combinations of time series. This function should have at least two parameters for each time series. Ex: <code>function(ts1, ts2) cor(ts1, ts2)</code>

isSymetric	Boolean. If the distance function is symmetric.
error_value	The value returned if an error occur when calculating a the distance for a pair of time series.
warn_error	Boolean. If TRUE (default), a warning will rise when an error occur during the calculations.
simplify	Boolean. If FALSE, returns a list of one (if isSymetric == FALSE) or two elements (if isSymetric == TRUE).
num_cores	Numeric. Number of cores
...	Additional parameters for measureFunc

Value

A data frame with elements (i,j) and a distance value calculated for the time series i and j.

ts_dist_part_file	<i>Calculate distances between pairs of time series stored in files.</i>
-------------------	--

Description

This function works similarly as `dist_parts_parallel()`. The difference is that it reads the time series from RDS files in a directory. The advantage of this approach is that it does not load all the time series in memory but reads them only when necessary. This means that this function requires much less memory and should be preferred when memory consumption is a concern, e.g., huge data set or very long time series. The disadvantage of this approach is that it requires a high number of file read operations which considerably takes more time during the calculations. **IMPORTANT:** the file order is very important so it is highly recommended to use numeric names, e.g., 0013.RDS.

Usage

```
ts_dist_part_file(
  input_dir,
  num_part,
  num_total_parts,
  combinations,
  measureFunc = tsdist_cor,
  isSymetric = TRUE,
  error_value = NaN,
  warn_error = TRUE,
  simplify = TRUE,
  num_cores = 1,
  ...
)
```

Arguments

input_dir	Directory path for the directory with time series files (RDS)
num_part	Numeric positive between 1 and the total number of parts (num_total_parts). This value corresponds to the part (chunk) of the total number of parts to be calculated.
num_total_parts	Numeric positive corresponding the total number of parts.
combinations	A list composed by arrays of size 2 indicating the files indices to be compared. If this parameter is passed, then the function does not split all the possibilities and does not use the parameters num_part and num_total_parts.
measureFunc	Function to be applied to all combinations of time series. This function should have at least two parameters for each time series. Ex: function(ts1, ts2) cor(ts1, ts2)
isSymetric	Boolean. If the distance function is symmetric.
error_value	The value returned if an error occur when calculating a the distance for a pair of time series.
warn_error	Boolean. If TRUE (default), a warning will rise when an error occur during the calculations.
simplify	Boolean. If FALSE, returns a list of one (if isSymetric == FALSE) or two elements (if isSymetric == TRUE).
num_cores	Numeric. Number of cores
...	Additional parameters for measureFunc

Value

A data frame with elements (i,j) and a distance value calculated for the time series i and j. Each index corresponds to the order where the files are listed.

ts_to_windows	<i>Extract time windows from a time series</i>
---------------	--

Description

This function is useful when constructing a network from a single time series. The returned list can be directly used to calculate the distance matrix D with ts_dist().

Usage

```
ts_to_windows(x, width, by = 1)
```

Arguments

x	time series
width	window length
by	Window step. This is the number of values in and out during the window rollover process.

Value

List of windows

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