Package 'rEDM'

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Type Package

Title Empirical Dynamic Modeling ('EDM')

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Description An implementation of 'EDM' algorithms based on research software developed for internal use at the Sugihara Lab ('UCSD/SIO'). The package is implemented with 'Rcpp' wrappers around the 'cppEDM' library. It implements the 'simplex' projection method from Sugihara & May (1990) <doi:10.1038/344734a0>, the 'S-map' algorithm from Sugihara (1994) <doi:10.1098/rsta.1994.0106>, convergent cross mapping described in Sugihara et al. (2012) <doi:10.1126/science.1227079>, and, 'multiview embedding' described in Ye & Sugihara (2016) <doi:10.1126/science.aag0863>.

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LazyData true

LazyLoad yes

Imports methods, Rcpp (>= 1.0.1)

LinkingTo Rcpp, RcppThread

Suggests knitr, rmarkdown

VignetteBuilder knitr

NeedsCompilation yes

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R topics documented:

`block_3sp` *Time series for a three-species coupled model.*

Description

Time series generated from a discrete-time coupled Lotka-Volterra model exhibiting chaotic dynamics.

Usage

```
block_3sp
```

Format

A data frame with 198 rows and 10 columns:

`time` time index (# of generations)

`x_t` abundance of simulated species x at time t

`x_t-1` abundance of simulated species x at time $t-1$

`x_t-2` abundance of simulated species x at time $t-2$

`y_t` abundance of simulated species y at time t

`y_t-1` abundance of simulated species y at time $t-1$

`y_t-2` abundance of simulated species y at time $t-2$

`z_t` abundance of simulated species z at time t

`z_t-1` abundance of simulated species z at time $t-1$

`z_t-2` abundance of simulated species z at time $t-2$

`block_gp` *Deprecated functions*

Description

Deprecated functions.

Usage

```
block_gp(block,lib = c(1, NROW(block)),pred = lib,tp = 1,phi = 0,
v_e = 0,eta = 0,fit_params = TRUE,columns = NULL,target_column = 1,
stats_only = TRUE,save_covariance_matrix = FALSE,
first_column_time = FALSE,silent = FALSE, ...)
ccm_means(df, FUN = mean, ...)
tde_gp(time_series,lib = c(1, NROW(time_series)),pred = lib,
E = 1:10,tau = 1,tp = 1,phi = 0,v_e = 0,eta = 0,fit_params = TRUE,
stats_only = TRUE,save_covariance_matrix = FALSE,silent = FALSE, ...)
test_nonlinearity(ts,method = "ebisuzaki", num_surr = 200,
T_period = 1,E = 1, ...)
```

Arguments

`block`	not implemented
`lib`	not implemented
`pred`	not implemented
`tp`	not implemented
`phi`	not implemented
`v_e`	not implemented

eta	not implemented
fit_params	not implemented
columns	not implemented
target_column	not implemented
stats_only	not implemented
save_covariance_matrix	not implemented
first_column_time	not implemented
silent	not implemented
df	not implemented
FUN	not implemented
time_series	not implemented
E	not implemented
tau	not implemented
ts	not implemented
method	not implemented
num_surr	not implemented
T_period	not implemented
...	not implemented

Value

Not implemneted.

`block_lnlp` *Perform generalized forecasting using simplex projection or s-map*

Description

`block_lnlp` uses multiple time series given as input to generate an attractor reconstruction, and then applies the simplex projection or s-map algorithm to make forecasts. This method generalizes the `simplex` and `s_map` routines, and allows for "mixed" embeddings, where multiple time series can be used as different dimensions of an attractor reconstruction.

Usage

```
block_lnlp(block, lib = NULL, pred = NULL, norm = 2, method = c("simplex",
  "s-map"), tp = 1, num_neighbors = switch(match.arg(method),
  simplex = "e+1", `s-map` = 0), columns = NULL, target_column = 1,
  stats_only = TRUE, first_column_time = FALSE, exclusion_radius = NULL,
  epsilon = NULL, theta = NULL, silent = TRUE, save_smap_coefficients = FALSE)
```

Arguments

block	either a vector to be used as the time series, or a data.frame or matrix where each column is a time series
lib	a 2-column matrix, data.frame, 2-element vector or string of row indice pairs, where each pair specifies the first and last *rows* of the time series to create the library. If not specified, all available rows are used
pred	(same format as lib), but specifying the sections of the time series to forecast. If not specified, set equal to lib
norm	the distance measure to use. see 'Details'
method	the prediction method to use. see 'Details'
tp	the prediction horizon (how far ahead to forecast)
num_neighbors	the number of nearest neighbors to use. Note that the default value will change depending on the method selected. (any of "e+1", "E+1", "e + 1", "E + 1" will set this parameter to E+1 for each run.)
columns	either a vector with the columns to use (indices or names), or a list of such columns
target_column	the index (or name) of the column to forecast
stats_only	specify whether to output just the forecast statistics or to include the raw predictions for each run
first_column_time	indicates whether the first column of the given block is a time column (and therefore excluded when building the library)
exclusion_radius	excludes vectors from the search space of nearest neighbors if their *time index* is within exclusion_radius (NULL turns this option off)
epsilon	Not implemented
theta	the nonlinear tuning parameter (theta is only relevant if method == "s-map")
silent	prevents warning messages from being printed to the R console
save_smap_coefficients	specifies whether to include the s_map coefficients with the output

Details

The default parameters are set so that passing a vector as the only argument will use that vector to predict itself one time step ahead. If a matrix or data.frame is given as the only argument, the first column will be predicted (one time step ahead), using the remaining columns as the embedding. If the first column is not a time vector, 1:NROW will be used as time values.

norm = 2 (only option currently available) uses the "L2 norm", Euclidean distance:

$$distance(a, b) := \sqrt{\sum_i (a_i - b_i)^2}$$

method "simplex" (default) uses the simplex projection forecasting algorithm

method "s-map" uses the s-map forecasting algorithm

Value

A data.frame with components for the parameters and forecast statistics:

cols	embedding
tp	prediction horizon
nn	number of neighbors
num_pred	number of predictions
rho	correlation coefficient between observations and predictions
mae	mean absolute error
rmse	root mean square error
perc	percent correct sign
p_val	p-value that rho is significantly greater than 0 using Fisher's z-transformation
const_pred_rho	same as `rho`, but for the constant predictor
const_pred_mae	same as `mae`, but for the constant predictor
const_pred_rmse	same as `rmse`, but for the constant predictor
const_pred_perc	same as `perc`, but for the constant predictor
const_p_val	same as `p_val`, but for the constant predictor
model_output	data.frame with columns for the time index, observations, predictions, and estimated prediction variance (i

If "s-map" is the method, then the same, but with additional columns:

theta	the nonlinear tuning parameter
smap_coefficients	data.frame with columns for the s-map coefficients (if `save_smap_coefficients == TRUE`)
smap_coefficient_covariances	list of covariance matrices for the s-map coefficients (if `save_smap_coefficients == TRUE`)

Examples

```
block <- block_3sp
block_lnlp(block[,2:4])

block <- block_3sp
block_lnlp(block[,1:4], first_column_time = TRUE)

block <- block_3sp
block_lnlp(block, target_column = "x_t", columns = c("y_t", "z_t"), first_column_time = TRUE)

block <- block_3sp
x_t_pred = block_lnlp(block, columns = c("x_t", "y_t"), first_column_time = TRUE,
stats_only = FALSE)

block <- block_3sp
x_t_pred = block_lnlp(block, method = "s-map", theta = 3, columns =
c("x_t", "y_t"), first_column_time = TRUE, stats_only = FALSE, save_smap_coefficients = TRUE)
```

CCM *Convergent cross mapping using simplex projection*

Description

The state-space of a multivariate dynamical system (not a purely stochastic one) encodes coherent phase-space variable trajectories. If enough information is available, one can infer the presence or absence of cross-variable interactions associated with causal links between variables. CCM measures the extent to which states of variable Y can reliably estimate states of variable X. This happens only if X is causally influencing Y.

If cross-variable state predictability converges as more state-space information is provided, this indicates a causal link. CCM performs this cross-variable mapping using Simplex, with convergence assessed across a range of observational library sizes as described in *Sugihara et al. 2012.*

Usage

```
CCM(pathIn = "./", dataFile = "", dataFrame = NULL, pathOut = "./",
  predictFile = "", E = 0, Tp = 0, knn = 0, tau = -1, columns = "", target = "",
  libSizes = "", sample = 0, random = TRUE, replacement = FALSE, seed = 0,
  includeData = FALSE, verbose = FALSE, showPlot = FALSE)
```

Arguments

pathIn	path to dataFile.
dataFile	.csv format data file name. The first column must be a time index or time values. The first row must be column names.
dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named.
pathOut	path for predictFile containing output predictions.
predictFile	output file name.
E	embedding dimension.
Tp	prediction horizon (number of time column rows).
knn	number of nearest neighbors. If knn=0, knn is set to E+1.
tau	lag of time delay embedding specified as number of time column rows.
columns	string of whitespace separated column name(s) in the input data used to create the library.
target	column name in the input data used for prediction.
libSizes	string of 3 whitespace separated integer values specifying the intial library size, the final library size, and the library size increment.
sample	integer specifying the number of random samples to draw at each library size evaluation.
random	logical to specify random (TRUE) or sequential library sampling.

replacement	logical to specify sampling with replacement.
seed	integer specifying the random sampler seed. If seed=0 then a random seed is generated.
includeData	logical to include statistics and predictions for every prediction in the ensemble.
verbose	logical to produce additional console reporting.
showPlot	logical to plot results.

Details

CCM computes the X:Y and Y:X cross-mappings in parallel using threads.

Value

A data.frame with 3 columns. The first column is LibSize specifying the subsampled library size. Columns 2 and 3 report Pearson correlation coefficients for the prediction of X from Y, and Y from X.

References

Sugihara G., May R., Ye H., Hsieh C., Deyle E., Fogarty M., Munch S., 2012. Detecting Causality in Complex Ecosystems. Science 338:496-500.

Examples

```
data(sardine_anchovy_sst)
df <- CCM( dataFrame=sardine_anchovy_sst, E=3, Tp=0, columns="anchovy",
target="np_sst", libSizes="10 70 10", sample=100 )
```

---

ccm *Convergent cross mapping using simplex projection*

Description

ccm uses time delay embedding on one time series to generate an attractor reconstruction, and then applies the simplex projection algorithm to estimate concurrent values of another time series. This method is typically applied, varying the library sizes, to determine if one time series contains the necessary dynamic information to recover the influence of another, causal variable.

Usage

```
ccm(block, lib = NULL, pred = NULL, norm = 2, E = 1, tau = -1,
    tp = 0, num_neighbors = "e+1", lib_sizes = c(10, 75, 5),
    random_libs = TRUE, num_samples = 100, replace = FALSE, lib_column = 1,
    target_column = 2, first_column_time = FALSE, RNGseed = NULL,
    exclusion_radius = NULL, epsilon = NULL, stats_only = TRUE,
    silent = TRUE)
```

Arguments

block	either a vector to be used as the time series, or a data.frame or matrix where each column is a time series
lib	a 2-column matrix, data.frame, 2-element vector or string of row indice pairs, where each pair specifies the first and last *rows* of the time series to create the library. If not specified, all available rows are used
pred	(same format as lib), but specifying the sections of the time series to forecast. If not specified, set equal to lib
norm	the distance measure to use. see 'Details'
E	the embedding dimensions to use for time delay embedding
tau	the time-delay offset to use for time delay embedding
tp	the prediction horizon (how far ahead to forecast)
num_neighbors	the number of nearest neighbors to use. Note that the default value will change depending on the method selected. (any of "e+1", "E+1", "e + 1", "E + 1" will set this parameter to E+1 for each run
lib_sizes	three integers specifying the start, stop and increment index of library sizes
random_libs	indicates whether to use randomly sampled libs
num_samples	is the number of random samples at each lib size (this parameter is ignored if random_libs is FALSE)
replace	indicates whether to sample vectors with replacement
lib_column	name (index) of the column to cross map from
target_column	name (index) of the column to forecast
first_column_time	indicates whether the first column of the given block is a time column
RNGseed	will set a seed for the random number generator, enabling reproducible runs of ccm with randomly generated libraries
exclusion_radius	excludes vectors from the search space of nearest neighbors if their *time index* is within exclusion_radius (NULL turns this option off)
epsilon	not implemented
stats_only	specify whether to output just the forecast statistics or the raw predictions for each run
silent	prevents warning messages from being printed to the R console

Details

ccm runs both forward and reverse cross maps in seperate threads. Results are returned for both mappings. The default parameters are set so that passing a matrix as the only argument will use E = 1 (embedding dimension), and leave-one-out cross-validation over the whole time series to compute cross-mapping from the first column to the second column, letting the library size vary from 10 to 75 in increments of 5.

norm = 2 (only option currently available) uses the "L2 norm", Euclidean distance:

$$distance(a, b) := \sqrt{\sum_i (a_i - b_i)^2}$$

Value

If stats_only = TRUE: a data.frame with forecast statistics for both the forward and reverse mappings:

LibSize	library length (number of vectors)
x:y	cross mapped correlation coefficient between observations x and predictions y
y:x	cross mapped correlation coefficient between observations y and predictions x
E	embedding dimension
tau	time delay offset
tp	forecast interval
nn	number nearest neighbors

If stats_only = FALSE: a named list with the following items: settings:

LibMeans	data.frame with the mean bidirectional forecast statistics
CCM1_PredictStat	data.frame with forward mapped prediction statistics for each prediction of the ensemble
CCM1_Predictions	list of prediction result data.frame each forward mapped prediction of the ensemble
CCM2_PredictStat	data.frame with reverse mapped prediction statistics for each prediction of the ensemble
CCM2_Predictions	list of prediction result data.frame each reverse mapped prediction of the ensemble

CCM1_PredictStat and CCM2_PredictStat data.frames have columns:

N	prediction number
E	embedding dimension
nn	number of nearest neighbors
tau	embedding time delay offset
LibSize	library size
rho	correlation coefficient
RMSE	root mean square error
MAE	maximum absolute error
lib	column name of the library vector
target	column name of the target vector

Examples

```
anchovy_xmap_sst <- ccm(sardine_anchovy_sst, E = 3,
  lib_column = "anchovy", target_column = "np_sst",
  lib_sizes = c(10, 75, 5), num_samples = 100)
```

`circle` *2-D timeseries of a circle.*

Description

Time series of of circle in 2-D (sin and cos).

Usage

```
circle
```

Format

A data frame with 200 rows and 3 columns:

`Time` time index.

`x` sin component.

`y` cos component.

`ComputeError` *Compute error*

Description

`ComputeError` evaluates the Pearson correlation coefficient, mean absolute error and root mean square error between two numeric vectors.

Usage

```
ComputeError(obs, pred)
```

Arguments

`obs`	vector of observations.
`pred`	vector of predictions.

Value

A name list with components:

rho	Pearson correlation
MAE	mean absolute error
RMSE	root mean square error

Examples

```
data(block_3sp)
smplx <- Simplex( dataFrame=block_3sp, lib="1 99", pred="105 190", E=3,
columns="x_t", target="x_t")
err <- ComputeError( smplx$Observations, smplx$Predictions )
```

compute_stats *Compute performance metrics for predictions*

Description

Computes the rho, MAE, RMSE, perc, and p-val performance metrics

Arguments

observed	a vector of the observed values
predicted	a vector of corresponding predicted values

Value

A data.frame with components with various performance metrics:

num_pred	number of predictions
rho	correlation coefficient between observations and predictions
mae	mean absolute error
rmse	root mean square error
perc	percent correct sign
p_val	p-value that rho is significantly greater than 0 using Fisher's

Examples

```
compute_stats(rnorm(100), rnorm(100))
```

EDM *Empirical dynamic modeling*

Description

EDM provides tools for data-driven time series analyses. It is based on reconstructing multivariate state (or phase) space representations from uni or multivariate time series, then projecting state changes using various metrics applied to nearest neighbors.

EDM is a **Rcpp** interface to the cppEDM library of Empirical Dynamic Modeling tools. Functionality includes:

- Simplex projection (Sugihara and May 1990)
- Sequential Locally Weighted Global Linear Maps (S-map) (Sugihara 1994)
- Multivariate embeddings (Dixon et. al. 1999)
- Convergent cross mapping (Sugihara et. al. 2012)
- Multiview embedding (Ye and Sugihara 2016)

Details

Main Functions:

- Simplex - simplex projection
- SMap - S-map projection
- CCM - convergent cross mapping
- Multiview - multiview forecasting

Helper Functions:

- Embed - time delay embedding
- ComputeError - forecast skill metrics
- EmbedDimension - optimal embedding dimension
- PredictInterval - optimal prediction interval
- PredictNonlinear - evaluate nonlinearity

Author(s)

Maintainer: Joseph Park & Cameron Smith

Authors: Joseph Park, Cameron Smith, Ethan Deyle, Erik Saberski, George Sugihara

References

Sugihara G. and May R. 1990. Nonlinear forecasting as a way of distinguishing chaos from measurement error in time series. Nature, 344:734-741.

Sugihara G. 1994. Nonlinear forecasting for the classification of natural time series. Philosophical Transactions: Physical Sciences and Engineering, 348 (1688) : 477-495.

Dixon, P. A., M. Milicich, and G. Sugihara, 1999. Episodic fluctuations in larval supply. Science 283:1528-1530.

Sugihara G., May R., Ye H., Hsieh C., Deyle E., Fogarty M., Munch S., 2012. Detecting Causality in Complex Ecosystems. Science 338:496-500.

Ye H., and G. Sugihara, 2016. Information leverage in interconnected ecosystems: Overcoming the curse of dimensionality. Science 353:922-925.

Embed *Embed data with time lags*

Description

Embed performs Takens time-delay embedding on columns.

Usage

```
Embed(path = "./", dataFile = "", dataFrame = NULL, E = 0, tau = -1,
columns = "", verbose = FALSE)
```

Arguments

path	path to dataFile.
dataFile	.csv format data file name. The first column must be a time index or time values. The first row must be column names. One of dataFile or dataFrame are required.
dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named. One of dataFile or dataFrame are required.
E	embedding dimension.
tau	integer time delay embedding lag specified as number of time column rows.
columns	string of whitespace separated column name(s) in the input data to be embedded.
verbose	logical to produce additional console reporting.

Details

Each columns item will have E-1 time-lagged vectors created. The column name is appended with (t-n). For example, data columns X, Y, with E = 2 will have columns named X(t-0) X(t-1) Y(t-0) Y(t-1).

The returned data.frame does not have a time column. The returned data.frame is truncated by tau * (E-1) rows to remove state vectors with partial data (NaN elements).

Value

A data.frame with lagged columns. E columns for each variable specified in columns.

Examples

```
data(circle)
embed <- Embed( dataFrame = circle, E = 2, tau = -1, columns = "x y" )
```

EmbedDimension *Optimal embedding dimension*

Description

EmbedDimension uses Simplex to evaluate prediction accuracy as a function of embedding dimension.

Usage

```
EmbedDimension(pathIn = "./", dataFile = "", dataFrame = NULL, pathOut = "",
  predictFile = "", lib = "", pred = "", maxE = 10, Tp = 1, tau = -1,
  columns = "", target = "", embedded = FALSE, verbose = FALSE, numThreads = 4,
  showPlot = TRUE)
```

Arguments

pathIn	path to dataFile.
dataFile	.csv format data file name. The first column must be a time index or time values. The first row must be column names.
dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named.
pathOut	path for predictFile containing output predictions.
predictFile	output file name.
lib	string with start and stop indices of input data rows used to create the library of observations. A single contiguous range is supported.
pred	string with start and stop indices of input data rows used for predictions. A single contiguous range is supported.
maxE	maximum value of E to evalulate.
Tp	prediction horizon (number of time column rows).
tau	lag of time delay embedding specified as number of time column rows.
columns	string of whitespace separated column name(s) in the input data used to create the library.
target	column name in the input data used for prediction.
embedded	logical specifying if the input data are embedded.
verbose	logical to produce additional console reporting.
numThreads	number of parallel threads for computation.
showPlot	logical to plot results.

Value

A data.frame with columns E, rho.

Examples

```
data(TentMap)
E.rho <- EmbedDimension( dataFrame=TentMap, lib="1 100", pred="201 500",
columns="TentMap", target="TentMap", showPlot=FALSE)
```

EvergladesFlow *Water flow to NE Everglades*

Description

Cumulative weekly water flow into northeast Everglades from water control structures S12C, S12D and S333 from 1980 through 2005.

Usage

```
EvergladesFlow
```

Format

A data frame with 1379 rows and 2 columns:

`Date` Date.

`S12CD_S333_CFS` Cumulative weekly flow (CFS).

`Lorenz5D` *5-D Lorenz'96*

Description

5-D Lorenz'96 timeseries with F = 8.

Usage

```
Lorenz5D
```

Format

Data frame with 1000 rows and 6 columns

`Time` Time.

`V1` variable 1.

`V2` variable 2.

`V3` variable 3.

`V4` variable 4.

`V5` variable 5.

References

Lorenz, Edward (1996). Predictability - A problem partly solved, Seminar on Predictability, Vol. I, ECMWF.

MakeBlock *Make embedded data block*

Description

MakeBlock performs Takens time-delay embedding on columns. It is an internal function called by Embed that does not perform input error checking or validation.

Usage

```
MakeBlock(dataFrame, E = 0, tau = -1, columns = "")
```

Arguments

dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named.
E	embedding dimension.
tau	integer time delay embedding lag specified as number of time column rows.
columns	string of whitespace separated column name(s) in the input data to be embedded.

Details

Each columns item will have E-1 time-lagged vectors created. The column name is appended with (t-n). For example, data columns X, Y, with E = 2 will have columns named X(t-0) X(t-1) Y(t-0) Y(t-1).

The returned data.frame does not have a time column. The returned data.frame is truncated by tau * (E-1) rows to remove state vectors with partial data (NaN elements).

Value

A data.frame with lagged columns. E columns for each variable specified in columns.

Examples

```
data(TentMap)
embed <- MakeBlock(TentMap, 3, 1, "TentMap")
```

make_block *Make a time delay offset block*

Description

make_block generates a time offset block with the appropriate max_lag and tau. The first column is presumed to be a time or index vector, and is not included in the embedding.

Usage

```
make_block(block, columns = NULL, t = NULL, max_lag = 3, tau = -1, lib =
NULL, restrict_to_lib = TRUE)
```

Arguments

block	a data.frame or matrix where each column is a time series
columns	list of column names to time delay.
t	Not used
max_lag	the total number of lags to include for each variable. So if max_lag == 3, a variable X is offset with lags X[t], X[t + tau], X[t + 2*tau]
tau	the time delay offset for embedding
lib	not used
restrict_to_lib	not used

Value

A data.frame with time offset columns. If the original block had columns X, Y and max_lag = 3, then the returned data.frame will have columns X(t-0) X(t-1) X(t-2) Y(t-0) Y(t-1) Y(t-2).

Examples

```
data("block_3sp")
make_block(block_3sp[, c(1, 2, 5)])
```

make_surrogate_data *Generate surrogate data for permutation/randomization tests*

Description

This is a wrapper function for generating surrogate time series using several different null models.

Usage

```
make_surrogate_data(ts, method = c("random_shuffle", "ebisuzaki",
  "seasonal"), num_surr = 100, T_period = 1, alpha = 0)
```

Arguments

ts	the original time series
method	which algorithm to use to generate surrogate data
num_surr	the number of null surrogates to generate
T_period	the period of seasonality for seasonal surrogates (ignored for other methods)
alpha	standard deviation of seasonal cycle deviates.

Value

A matrix where each column is a separate surrogate with the same length as 'ts'.

Examples

```
data = make_surrogate_data(block_3sp$x_t)
```

Multiview *Forecasting using multiview embedding*

Description

Multiview applies the method of *Ye & Sugihara* to find optimal combinations of variables that best represent the dynamics.

Usage

```
Multiview(pathIn = "./", dataFile = "", dataFrame = NULL, pathOut = "./",
  predictFile = "", lib = "", pred = "", D = 0, E = 1, Tp = 1, knn = 0,
  tau = -1, columns = "", target = "", multiview = 0, exclusionRadius = 0,
  trainLib = TRUE, verbose = FALSE, numThreads = 4, showPlot = FALSE)
```

Arguments

pathIn	path to dataFile.
dataFile	.csv format data file name. The first column must be a time index or time values. The first row must be column names.
dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named.
pathOut	path for predictFile containing output predictions.
predictFile	prediction output file name.
lib	a 2-column matrix, data.frame, 2-element vector or string of row indice pairs, where each pair specifies the first and last *rows* of the time series to create the library.
pred	(same format as lib), but specifying the sections of the time series to forecast.
D	multivariate dimension.
E	embedding dimension.
Tp	prediction horizon (number of time column rows).
knn	number of nearest neighbors. If knn=0, knn is set to E+1.
tau	lag of time delay embedding specified as number of time column rows.
columns	string of whitespace separated column name(s) in the input data used to create multivariable data sets.
target	column name in the input data used for prediction.
multiview	number of multiview ensembles to average for the final prediction estimate.
exclusionRadius	number of adjacent observation vector rows to exclude as nearest neighbors in prediction.
trainLib	logical to use in-sample (lib=pred) projections for the ranking of column combinations.
verbose	logical to produce additional console reporting.
numThreads	number of CPU threads to use in multiview processing.
showPlot	logical to plot results.

Details

Multiview embedding is a method to identify variables in a multivariate dynamical system that are most likely to contribute to the observed dynamics. It is a multistep algorithm with these general steps:

1. Compute D-dimensional variable combination forecasts.
2. Rank forecasts.
3. Compute predictions of top combinations.
4. Compute multiview averaged prediction.

If E>1, all variables are embedded to dimension E. If trainLib is TRUE initial forecasts and ranking are done in-sample (lib=pred) and predictions using the top ranked combinations use the specified lib and pred. If trainLib is FALSE initial forecasts and ranking use the specified lib and pred, the step of computing predictions of the top combinations is skipped.

Value

Named list with data.frames `[[Combo_rho,Predictions]]`.

data.frame `Combo_rho` columns:

Col_1	column index
...	column index
Col_E	column index
rho	Pearson correlation
MAE	mean absolute error
RMSE	root mean square error

References

Ye H., and G. Sugihara, 2016. Information leverage in interconnected ecosystems: Overcoming the curse of dimensionality. Science 353:922-925.

Examples

```
data(block_3sp)
L = Multiview( dataFrame = block_3sp, lib = "1 99", pred = "105 190",
E = 2, columns = "x_t y_t z_t", target = "x_t" )
```

`multiview` *Perform forecasting using multiview embedding*

Description

`multiview` applies the method described in Ye & Sugihara (2016) for forecasting, where multiple attractor reconstructions are tested, and a single nearest neighbor is selected from each of the top k reconstructions to produce final forecasts.

Usage

```
multiview(block, lib = NULL, pred = NULL, norm = 2, E = 1, tau = -1,
    tp = 1, max_lag = 3, num_neighbors = "e+1", k = "sqrt", na.rm = FALSE,
    target_column = 1, stats_only = TRUE, save_lagged_block = FALSE,
    first_column_time = FALSE, exclusion_radius = NULL, silent = FALSE)
```

Arguments

`block`	either a vector to be used as the time series, or a data.frame or matrix where each column is a time series
`lib`	a 2-column matrix, data.frame, 2-element vector or string of row indice pairs, where each pair specifies the first and last *rows* of the time series to create the library. If not specified, all available rows are used

pred	(same format as lib), but specifying the sections of the time series to forecast. If not specified, set equal to lib
norm	the distance measure to use. see 'Details'
E	the embedding dimensions to use for time delay embedding. The default value of 1 does not embed the data.
tau	the time-delay offset to use for time delay embedding
tp	the prediction horizon (how far ahead to forecast)
max_lag	the maximum number of lags to use for variable combinations. If max_lag == 3, a variable X will be embedded with lags X[t], X[t + tau], X[t + 2*tau]
num_neighbors	the number of nearest neighbors to use. Note that the default value will change depending on the method selected. (any of "e+1", "E+1", "e + 1", "E + 1" will set this parameter to E+1 for each run.)
k	the number of embeddings to use for ensemble averaging. "sqrt" or 0 will use k = sqrt(m) where m is the number of multiview combinations of the set of input variables
na.rm	logical. Should missing values (including 'NaN" be omitted from the calculations?)
target_column	the name (index) of the column to forecast
stats_only	specify whether to output just the forecast statistics or the raw predictions for each run
save_lagged_block	specify whether to output the lagged block that is constructed as part of running multiview
first_column_time	indicates whether the first column of the given block is a time column and excluded when building the library
exclusion_radius	excludes vectors from the search space of nearest neighbors if their *time index* is within exclusion_radius (NULL turns this option off)
silent	prevents warning messages from being printed to the R console

Details

multiview uses multiple time series given as input to generate an attractor reconstruction, and then applies the simplex projection to make forecasts. This method generalizes the simplex routine, and allows for "mixed" embeddings, where multiple time series can be used as different dimensions of an attractor reconstruction.

The default parameters are set so that, given a matrix of time series, forecasts will be produced for the first column. By default, all possible combinations of the columns are used for the attractor construction, the k = sqrt(m) heuristic will be used, forecasts will be one time step ahead. If a time vector is not supplied, 1:NROW will be used. The default lib and pred are to use the first half of the data for the "library" and to predict over the second half of the data. Unless otherwise set, the output will be just the forecast statistics.

norm = 2 (only option currently available) uses the "L2 norm", Euclidean distance:

$$distance(a, b) := \sqrt{\sum_i (a_i - b_i)^2}$$

Value

A named list with items "View" and "Predictions". View is a data.frame with components:

col_i,... col_j	column indices of the embedding
name_i,... nam_j	column names of the embedding
rho	correlation of the projection
MAE	maximum absolute error of the projection
RMSE	root mean square error of the projection

Predictions is a data.frame of the predictions from the best multivew ensemble.

Examples

```
block <- block_3sp[, c(2, 5, 8)]
multiview( block, k=10 )
```

---

paramecium_didinium *Time series for the Paramecium-Didinium laboratory experiment*

Description

Time series of Paramecium and Didinium abundances (#/mL) from an experiment by Veilleux (1979)

Usage

```
paramecium_didinium
```

PredictInterval *Forecast interval accuracy*

Description

PredictInterval uses Simplex to evaluate prediction accuracy as a function of forecast interval Tp.

Usage

```
PredictInterval(pathIn = "./", dataFile = "", dataFrame = NULL, pathOut = "./",
  predictFile = "", lib = "", pred = "", maxTp = 10, E = 1, tau = -1,
  columns = "", target = "", embedded = FALSE, verbose = FALSE,
  numThreads = 4, showPlot = TRUE)
```

Arguments

pathIn	path to dataFile.
dataFile	.csv format data file name. The first column must be a time index or time values. The first row must be column names.
dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named.
pathOut	path for predictFile containing output predictions.
predictFile	output file name.
lib	string with start and stop indices of input data rows used to create the library of observations. A single contiguous range is supported.
pred	string with start and stop indices of input data rows used for predictions. A single contiguous range is supported.
maxTp	maximum value of Tp to evalulate.
E	embedding dimension.
tau	lag of time delay embedding specified as number of time column rows.
columns	string of whitespace separated column name(s) in the input data used to create the library.
target	column name in the input data used for prediction.
embedded	logical specifying if the input data are embedded.
verbose	logical to produce additional console reporting.
numThreads	number of parallel threads for computation.
showPlot	logical to plot results.

Value

A data.frame with columns Tp,rho.

Examples

```
data(TentMap)
Tp.rho <- PredictInterval( dataFrame=TentMap, lib="1 100",
pred="201 500", E=2, columns="TentMap", target="TentMap", showPlot = FALSE)
```

PredictNonlinear *Test for nonlinear dynamics*

Description

PredictNonlinear uses SMap to evaluate prediction accuracy as a function of the localisation parameter theta.

Usage

```
PredictNonlinear(pathIn = "./", dataFile = "", dataFrame = NULL,
  pathOut = "./",  predictFile = "", lib = "", pred = "", theta = "",
  E = 1, Tp = 1, knn = 0, tau = -1, columns = "", target = "",
  embedded = FALSE, verbose = FALSE, numThreads = 4, showPlot = TRUE)
```

Arguments

pathIn	path to dataFile.
dataFile	.csv format data file name. The first column must be a time index or time values. The first row must be column names.
dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named.
pathOut	path for predictFile containing output predictions.
predictFile	output file name.
lib	string with start and stop indices of input data rows used to create the library of observations. A single contiguous range is supported.
pred	string with start and stop indices of input data rows used for predictions. A single contiguous range is supported.
theta	A whitespace delimeted string with values of the S-map localisation parameter. An empty string will use default values of [0.01 0.1 0.3 0.5 0.75 1 1.5 2 3 4 5 6 7 8 9].
E	embedding dimension.
Tp	prediction horizon (number of time column rows).
knn	number of nearest neighbors. If knn=0, knn is set to the library size.
tau	lag of time delay embedding specified as number of time column rows.
columns	string of whitespace separated column name(s) in the input data used to create the library.
target	column name in the input data used for prediction.
embedded	logical specifying if the input data are embedded.
verbose	logical to produce additional console reporting.
numThreads	number of parallel threads for computation.
showPlot	logical to plot results.

Details

The localisation parameter theta weights nearest neighbors according to exp((-theta D / D_avg)) where D is the distance between the observation vector and neighbor, D_avg the mean distance. If theta = 0, weights are uniformally unity corresponding to a global autoregressive model. As theta increases, neighbors in closer proximity to the observation are considered.

Value

A data.frame with columns Theta,rho.

Examples

```
data(TentMapNoise)
theta.rho <- PredictNonlinear( dataFrame=TentMapNoise, E=2,lib="1 100",
pred="201 500", columns="TentMap", target="TentMap", showPlot = FALSE)
```

---

sardine_anchovy_sst *Time series for the California Current Anchovy-Sardine-SST system*

Description

Time series of Pacific sardine landings (CA), Northern anchovy landings (CA), and sea-surface temperature (3-year average) at the SIO pier and Newport pier

Usage

```
sardine_anchovy_sst
```

Format

`year` year of measurement

`anchovy` anchovy landings, scaled to mean = 0, sd = 1

`sardine` sardine landings, scaled to mean = 0, sd = 1

`sio_sst` 3-year running average of sea surface temperature at SIO pier, scaled to mean = 0, sd = 1

`np_sst` 3-year running average of sea surface temperature at Newport pier, scaled to mean = 0, sd = 1

Simplex *Simplex forecasting*

Description

`Simplex` performs time series forecasting based on weighted nearest neighbors projection in the time series phase space as described in *Sugihara and May*.

Usage

```
Simplex(pathIn = "./", dataFile = "", dataFrame = NULL, pathOut = "./",
  predictFile = "", lib = "", pred = "", E = 0, Tp = 1, knn = 0, tau = -1,
  exclusionRadius = 0, columns = "", target = "", embedded = FALSE,
  verbose = FALSE, const_pred = FALSE, showPlot = FALSE)
```

Arguments

pathIn	path to dataFile.
dataFile	.csv format data file name. The first column must be a time index or time values. The first row must be column names.
dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named.
pathOut	path for predictFile containing output predictions.
predictFile	output file name.
lib	string with start and stop indices of input data rows used to create the library of observations. A single contiguous range is supported.
pred	string with start and stop indices of input data rows used for predictions. A single contiguous range is supported.
E	embedding dimension.
Tp	prediction horizon (number of time column rows).
knn	number of nearest neighbors. If knn=0, knn is set to E+1.
tau	lag of time delay embedding specified as number of time column rows.
exclusionRadius	excludes vectors from the search space of nearest neighbors if their relative time index is within exclusionRadius.
columns	string of whitespace separated column name(s) in the input data used to create the library.
target	column name in the input data used for prediction.
embedded	logical specifying if the input data are embedded.
verbose	logical to produce additional console reporting.
const_pred	logical to add a *constant predictor* column to the output. The constant predictor is X(t+1) = X(t).
showPlot	logical to plot results.

Details

If embedded is FALSE, the data column(s) are embedded to dimension E with time lag tau. This embedding forms an E-dimensional phase space for the Simplex projection. If embedded is TRUE, the data are assumed to contain an E-dimensional embedding with E equal to the number of columns. Predictions are made using leave-one-out cross-validation, i.e. observation vectors are excluded from the prediction simplex.

To assess an optimal embedding dimension EmbedDimension can be applied. Accuracy statistics can be estimated by ComputeError.

Value

A data.frame with columns Observations,Predictions. If const_pred is TRUE the column Const_Predictions is added. The first column contains the time values.

References

Sugihara G. and May R. 1990. Nonlinear forecasting as a way of distinguishing chaos from measurement error in time series. Nature, 344:734-741.

Examples

```
data(block_3sp)
smplx <- Simplex( dataFrame=block_3sp, lib="1 99", pred="105 190",
E=3, columns="x_t", target="x_t" )
ComputeError(smplx$Predictions, smplx$Observations)
```

`simplex` *Perform univariate forecasting*

Description

`simplex` uses time delay embedding on a single time series to generate an attractor reconstruction, and then applies the simplex projection algorithm to make forecasts.

`s_map` is similar to `simplex`, but uses the S-map algorithm to make forecasts.

Usage

```
simplex(time_series, lib = NULL, pred = NULL, norm = 2, E = 1:10,
    tau = -1, tp = 1, num_neighbors = "e+1", stats_only = TRUE,
    exclusion_radius = NULL, epsilon = NULL, silent = TRUE)

s_map(time_series, lib = NULL, pred = NULL, norm = 2, E = 1,
    tau = -1, tp = 1, num_neighbors = 0, theta = NULL, stats_only = TRUE,
    exclusion_radius = NULL, epsilon = NULL, silent = TRUE,
    save_smap_coefficients = FALSE)
```

Arguments

`time_series`	either a vector to be used as the time series, or a data.frame or matrix with at least 2 columns (in which case the first column will be used as the time index, and the second column as the time series)
`lib`	a 2-column matrix, data.frame, 2-element vector or string of row indice pairs, where each pair specifies the first and last *rows* of the time series to create the library. If not specified, all available rows are used
`pred`	(same format as lib), but specifying the sections of the time series to forecast. If not specified, set equal to lib
`norm`	the distance measure to use. see 'Details'
`E`	the embedding dimensions to use for time delay embedding
`tau`	the time-delay offset to use for time delay embedding
`tp`	the prediction horizon (how far ahead to forecast)

num_neighbors	the number of nearest neighbors to use. Note that the default value will change depending on the method selected. (any of "e+1", "E+1", "e + 1", "E + 1" will set this parameter to E+1.)
stats_only	specify whether to output just the forecast statistics or the raw predictions for each run
exclusion_radius	excludes vectors from the search space of nearest neighbors if their *time index* is within exclusion_radius (NULL turns this option off)
epsilon	Deprecated.
silent	prevents warning messages from being printed to the R console
theta	the nonlinear tuning parameter (theta is only relevant if method == "s-map")
save_smap_coefficients	specifies whether to include the s_map coefficients with the output

Details

simplex is typically applied, and the embedding dimension varied, to find an optimal embedding dimension for the data. Thus, the default parameters are set so that passing a time series as the only argument will run over E = 1:10 (embedding dimension), using leave-one-out cross-validation over the whole time series, and returning just the forecast statistics.

s_map is typically applied, with fixed embedding dimension, and theta varied, to test for nonlinear dynamics in the data. Thus, the default parameters are set so that passing a time series as the only argument will run over a default list of thetas (0, 0.0001, 0.0003, 0.001, 0.003, 0.01, 0.03, 0.1, 0.3, 0.5, 0.75, 1.0, 1.5, 2, 3, 4, 6, and 8), using E = 1, leave-one-out cross-validation over the whole time series, and returning just the forecast statistics.

norm = 2 (only option currently available) uses the "L2 norm", Euclidean distance:

$$distance(a, b) := \sqrt{\sum_i (a_i - b_i)^2}$$

Value

For simplex, if stats_only = TRUE: a data.frame with components for the parameters and forecast statistics:

E	embedding dimension
tau	embedding time offset
tp	prediction horizon
nn	number of neighbors
num_pred	number of predictions
rho	correlation coefficient between observations and predictions
mae	mean absolute error
rmse	root mean square error
perc	percent correct sign
p_val	p-value that rho is significantly greater than 0 using Fisher's z-transformation
const_pred_rho	same as rho, but for the constant predictor
const_pred_mae	same as mae, but for the constant predictor

const_pred_rmse	same as `rmse`, but for the constant predictor
const_pred_perc	same as `perc`, but for the constant predictor
const_p_val	same as `p_val`, but for the constant predictor

For `simplex`, if stats_only = FALSE: a named list with data.frame "stats" specified above, and named list "model_output":

model_output	named list with data.frames for each model. Columns include the time index, observations, predictions, and es

For `s_map`, if stats_only = TRUE, the same data.frame as for `simplex`, but with additional column:

theta	the nonlinear tuning parameter

For `s_map`, if save_smap_coefficients = TRUE, a named list with data.frame "stats" specified above and the following list items:

smap_coefficients	data.frame with columns for the s-map coefficients
smap_coefficient_covariances	list of covariance matrices for the s-map coefficients

For `s_map`, if stats_only = FALSE, a named list with data.frame "stats" specified above, and named list "model_output":

model_output	named list with data.frames for each model. Columns include the time index, observations, predictions, and es

Examples

```
ts <- block_3sp$x_t
simplex(ts, lib = c(1, 100), pred = c(101, 190))

ts <- block_3sp$x_t
simplex(ts, stats_only = FALSE)

ts <- block_3sp$x_t
s_map(ts, E = 2)

ts <- block_3sp$x_t
s_map(ts, E = 2, theta = 1, save_smap_coefficients = TRUE)
```

`SMap` *SMap forecasting*

Description

SMap performs time series forecasting based on localised (or global) nearest neighbor projection in the time series phase space as described in *Sugihara 1994*.

Usage

```
SMap(pathIn = "./", dataFile = "", dataFrame = NULL, pathOut = "./",
  predictFile = "", lib = "", pred = "", E = 0, Tp = 1, knn = 0, tau = -1,
  theta = 0, exclusionRadius = 0, columns = "", target = "", smapFile = "",
  jacobians = "", embedded = FALSE, const_pred = FALSE, verbose = FALSE,
  showPlot = FALSE)
```

Arguments

pathIn	path to dataFile.
dataFile	.csv format data file name. The first column must be a time index or time values. The first row must be column names.
dataFrame	input data.frame. The first column must be a time index or time values. The columns must be named.
pathOut	path for predictFile containing output predictions.
predictFile	prediction output file name.
lib	string with start and stop indices of input data rows used to create the library of observations. A single contiguous range is supported.
pred	string with start and stop indices of input data rows used for predictions. A single contiguous range is supported.
E	embedding dimension.
Tp	prediction horizon (number of time column rows).
knn	number of nearest neighbors. If knn=0, knn is set to the library size.
tau	lag of time delay embedding specified as number of time column rows.
theta	neighbor localisation exponent.
exclusionRadius	excludes vectors from the search space of nearest neighbors if their relative time index is within exclusionRadius.
columns	string of whitespace separated column name(s) in the input data used to create the library.
target	column name in the input data used for prediction.
smapFile	output file containing SMap coefficients.
jacobians	not used.
embedded	logical specifying if the input data are embedded.
const_pred	logical to add a *constant predictor* column to the output. The constant predictor is X(t+1) = X(t).
verbose	logical to produce additional console reporting.
showPlot	logical to plot results.

Details

If embedded is FALSE, the data column(s) are embedded to dimension E with time lag tau. This embedding forms an E-dimensional phase space for the SMap projection. If embedded is TRUE, the data are assumed to contain an E-dimensional embedding with E equal to the number of columns. Predictions are made using leave-one-out cross-validation, i.e. observation vectors are excluded from the prediction regression.

In contrast to Simplex, SMap uses all available neighbors and then weights them with an exponential decay in phase space distance with exponent theta. theta=0 uses all neighbors corresponding to a global autoregressive model. As theta increases, neighbors closer in vicinity to the observation are considered.

Value

A named list with two data.frames [[predictions,coefficients]]. predictions has columns Observations,Predictions. If const_pred is TRUE the column Const_Predictions is added. The first column contains time values.

coefficients data.frame has time values in the first column. Columns 2 through E+2 (E+1 columns) are the SMap coefficients.

Note

SMap should be called with columns explicity correspondng to dimensions E. This means that if a multivariate data set is used (number of columns > 1) it should Not be an embedding from Embed since Embed will add lagged coordinates for each variable. The added columns will not correspond to the intended dimensions in the matrix inversion and prediction reconstruction. In this case, use the embedded = TRUE flag so that the columns selected and their coefficients correspond to the proper dimension.

References

Sugihara G. 1994. Nonlinear forecasting for the classification of natural time series. Philosophical Transactions: Physical Sciences and Engineering, 348 (1688):477-495.

Examples

```
data(circle)
L = SMap( dataFrame=circle,lib="1 100", pred="110 190", theta=4, E=2,
embedded=TRUE,columns="x y", target="x")
```

SurrogateData *Generate surrogate data for permutation/randomization tests*

Description

SurrogateData generates surrogate data under several different null models.

Usage

```
SurrogateData( ts, method = c("random_shuffle", "ebisuzaki",
"seasonal"), num_surr = 100, T_period = 1, alpha = 0 )
```

Arguments

ts	the original time series
method	which algorithm to use to generate surrogate data
num_surr	the number of null surrogates to generate
T_period	the period of seasonality for seasonal surrogates (ignored for other methods)
alpha	additive noise factor: N(0,alpha)

Details

Method "random_shuffle" creates surrogates by randomly permuting the values of the original time series.

Method "Ebisuzaki" creates surrogates by randomizing the phases of a Fourier transform, preserving the power spectra of the null surrogates.

Method "seasonal" creates surrogates by computing a mean seasonal trend of the specified period and shuffling the residuals. It is presumed that the seasonal trend can be exracted with a smoothing spline. Additive Gaussian noise is included according to N(0,alpha).

Value

A matrix where each column is a separate surrogate with the same length as ts.

Examples

```
data("block_3sp")
ts <- block_3sp$x_t
SurrogateData(ts, method = "ebisuzaki")
```

TentMap *Time series for a tent map with mu = 2.*

Description

First-differenced time series generated from the tent map recurrence relation with mu = 2.

Usage

```
TentMap
```

Format

Data frame with 999 rows and 2 columns

`Time` time index.

`TentMap` tent map values.

`TentMapNoise` *Time series of tent map plus noise.*

Description

First-differenced time series generated from the tent map recurrence relation with mu = 2 and random noise.

Usage

```
TentMapNoise
```

Format

Data frame with 999 rows and 2 columns

`Time` time index.

`TentMap` tent map values.

`Thrips` *Apple-blossom Thrips time series*

Description

Seasonal outbreaks of Thrips imaginis.

References

Davidson and Andrewartha, Annual trends in a natural population of Thrips imaginis *Thysanoptera*, Journal of Animal Ecology, 17, 193-199, 1948.

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