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 supports both the simplex projection method from Sugihara & May (1990) and the S-map algorithm in Sugihara (1994). In addition, it implements convergent cross mapping as described in Sugihara et al. (2012) and multiview embedding as described in Ye & Sugihara (2016).

Note Version 1.0.1 is a major rewrite of the rEDM package as a wrapper for the cppEDM library. Function names and signatures have changed. This allows a unified computation engine for EDM algorithms across C++, Python and R implementations. The cppEDM is written in a functional programming paradigm, facilitating clarity and ease of understanding to foster collaboration and extension. Additionally, cppEDM is intended for computationally intensive application, such as supercomputers including the AI Bridging Cloud Infrastructure (ABCI) where it has been applied to massively parallel neuronal timeseries cross mapping. The new package provides improved alignment between observed and forecast data rows, and, strict exclusion of partial data vectors. It does not currently support Gaussian process regression.

License file LICENSE

LazyData true

LazyLoad yes

Imports methods, Rcpp (>= 1.0.1)

LinkingTo Rcpp

Suggests knitr, rmarkdown

VignetteBuilder knitr

R topics documented:

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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\$

CCM 3

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

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

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,
   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.

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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.

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 )</pre>
```

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.

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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 )</pre>
```

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)
```

6 EmbedDimension

Arguments

path to dataFile. path .csv format data file name. The first column must be a time index or time valdataFile ues. 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. Ε embedding dimension. tau positive integer time delay embedding lag specified as number of time column rows. string of whitespace separated column name(s) in the input data to be embedded. columns 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" )</pre>
```

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)
```

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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.

1ib 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.

column name in the input data used for prediction.

logical specifying if the input data are embedded.

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)</pre>
```

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.

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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.

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|--|

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 = "", verbose = FALSE)
```

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	positive 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(TentMap)
embed <- MakeBlock(TentMap, 3, 1, "TentMap")</pre>
```

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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 = "", E = 0, Tp = 1, knn = 0,
    tau = 1, columns = "", target = "", multiview = 0, exclusionRadius = 0,
    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 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 multivariate 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.

verbose logical to produce additional console reporting.

numThreads number of CPU threads to use in multiview processing.

showPlot logical to plot results.

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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. Evaluate E-dimensional variable combinations within the library itself (in-sample).
- 2. Rank in-sample (library) forecasts.
- 3. Compute Multiview averaged out-of-sample prediction

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")
```

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)
```

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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)</pre>
```

PredictNonlinear Test for nonlinear dynamics

Description

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

PredictNonlinear 13

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 $\exp(\text{(-theta D/D_avg)})$ 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.

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Examples

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

rEDM

Empirical dynamic modeling

Description

rEDM 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.

rEDM 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

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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.

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
```

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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)
```

Arg

verbose

const_pred

showPlot

Ę	guments	
	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.
	Тр	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.

logical to produce additional console reporting.

is X(t+1) = X(t).

logical to plot results.

logical to add a constant predictor column to the output. The constant predictor

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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)</pre>
```

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)
```

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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.

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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 corresponding 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)

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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 exacted 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")</pre>
```

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 21

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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