

# Package ‘tseriesChaos’

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**Title** Analysis of nonlinear time series

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**Depends** R (>= 2.2.0), deSolve

**Suggests** scatterplot3d

**LazyData** yes

**LazyLoad** yes

**Description** Routines for the analysis of nonlinear time series. This work is largely inspired by the TISEAN project, by Rainer Hegger, Holger Kantz and Thomas Schreiber: <http://www.mpiyks-dresden.mpg.de/~tisean/>

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C2	<i>Sample correlation integral</i>
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### Description

Sample correlation integral for the specified length scale

### Usage

C2(series, m, d, t, eps)

### Arguments

series	time series
m	embedding dimension
d	time delay
t	Theiler window
eps	length scale

### Details

Computes the sample correlation integral on the provided time series for the specified length scale, and considering a time window  $t$  (see references). It uses a naive algorithm: simply returns the fraction of points pairs nearer than  $\text{eps}$ . Normally, you would use `d2`, which takes roughly the same time, but computes the correlation sum for multiple length scales and embedding dimensions at once.

### Value

The sample correlation integral at  $\text{eps}$  length scale.

### Author(s)

Antonio, Fabio Di Narzo

**References**

Hegger, R., Kantz, H., Schreiber, T., Practical implementation of nonlinear time series methods: The TISEAN package; CHAOS 9, 413-435 (1999)

**See Also**

[d2](#)

---

d2

*Sample correlation integral (at multiple length scales)*

---

**Description**

Computes the sample correlation integral over a grid of neps length scales starting from eps.min, and for multiple embedding dimensions

**Usage**

```
d2(series, m, d, t, eps.min, neps=100)
```

**Arguments**

series	time series
m	max embedding dimension
d	time delay
t	Theiler window
eps.min	min length scale
neps	number of length scales to evaluate

**Details**

Computes the sample correlation integral over neps length scales starting from eps.min, for embedding dimension  $1, \dots, m$ , considering a t time window (see references). The slope of the linear segment in the log-log plot gives an estimate of the correlation dimension (see the example).

**Value**

Matrix. Column 1: length scales. Column  $i=2, \dots, m+1$ : sample correlation integral for embedding dimension  $i-1$ .

**Author(s)**

Antonio, Fabio Di Narzo

**References**

Hegger, R., Kantz, H., Schreiber, T., Practical implementation of nonlinear time series methods: The TISEAN package; CHAOS 9, 413-435 (1999)

**Examples**

```
d2(lorenz.ts, m=6, d=2, t=4, eps.min=2)
```

---

```
duffing.syst
```

```
Duffing oscillator
```

---

**Description**

Duffing oscillator system, to be used with [sim.cont](#)

**Details**

To be used with [sim.cont](#)

**Author(s)**

Antonio, Fabio Di Narzo

---

```
embedd
```

```
Embedding of a time series
```

---

**Description**

Embedding of a time series with provided time delay and embedding dimension parameters.

**Usage**

```
embedd(x, m, d, lags)
```

**Arguments**

x	time series
m	embedding dimension (if lags missed)
d	time delay (if lags missed)
lags	vector of lags (if m and d are missed)

**Details**

Embedding of a time series with provided delay and dimension parameters.

**Value**

Matrix with columns corresponding to lagged time series.

**Author(s)**

Antonio, Fabio Di Narzo. Multivariate time series patch by Jonathan Shore.

**Examples**

```
library(scatterplot3d)
x <- window(rossler.ts, start=90)
xyz <- embedd(x, m=3, d=8)
scatterplot3d(xyz, type="l")

## embedding multivariate time series
series <- cbind(seq(1,50),seq(101,150))
head(embedd(series, m=6, d=1))
```

---

false.nearest	<i>Method of false nearest neighbours</i>
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**Description**

Method of false nearest neighbours to help deciding the optimal embedding dimension

**Usage**

```
false.nearest(series, m, d, t, rt=10, eps=sd(series)/10)
```

**Arguments**

series	time series
m	maximum embedding dimension
d	delay parameter
t	Theiler window
rt	escape factor
eps	neighborhood diameter

**Details**

Method of false nearest neighbours to help deciding the optimal embedding dimension.

**Value**

Fraction of false neighbors (first row) and total number of neighbors (second row) for each specified embedding dimension (columns)

**Author(s)**

Antonio, Fabio Di Narzo

**References**

Hegger, R., Kantz, H., Schreiber, T., Practical implementation of nonlinear time series methods: The TISEAN package; CHAOS 9, 413-435 (1999)

Kennel M. B., Brown R. and Abarbanel H. D. I., Determining embedding dimension for phase-space reconstruction using a geometrical construction, Phys. Rev. A, Volume 45, 3403 (1992).

**Examples**

```
(fn.out <- false.nearest(rossler.ts, m=6, d=8, t=180, eps=1, rt=3))
plot(fn.out)
```

---

lorenz.syst	<i>Lorenz system</i>
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**Description**

Lorenz system, to be used with [sim.cont](#)

**Details**

To be used with [sim.cont](#)

**Author(s)**

Antonio, Fabio Di Narzo

---

lorenz.ts	<i>Lorenz simulated time series, without noise</i>
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---

**Description**

Lorenz simulated time series, without noise. Of each state of the system, we observe the euclidean norm.

**Details**

Lorenz simulated time series, without noise, obtained with the call: `lorenz.ts <- sim.cont(lorenz.syst, 0, 100, 0.05)`

**Author(s)**

Antonio, Fabio Di Narzo

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Lyapunov exponent      *Tools to evaluate the maximal Lyapunov exponent of a dynamic system*

---

**Description**

Tools to evaluate the maximal Lyapunov exponent of a dynamic system from a univariate time series

**Usage**

```
lyap_k(series, m, d, t, k=1, ref, s, eps)
lyap(dsts, start, end)
```

**Arguments**

series	time series
m	embedding dimension
d	time delay
k	number of considered neighbours
eps	radius where to find nearest neighbours
s	iterations along which follow the neighbours of each point
ref	number of points to take into account
t	Theiler window
dsts	Should be the output of a call to lyap_k (see the example)
start	Starting time of the linear bite of dsts
end	Ending time of the linear bite of dsts

**Details**

The function `lyap_k` estimates the largest Lyapunov exponent of a given scalar time series using the algorithm of Kantz.

The function `lyap` computes the regression coefficients of a user specified segment of the sequence given as input.

**Value**

`lyap_k` gives the logarithm of the stretching factor in time.

`lyap` gives the regression coefficients of the specified input sequence.

**Author(s)**

Antonio, Fabio Di Narzo

## References

Hegger, R., Kantz, H., Schreiber, T., Practical implementation of nonlinear time series methods: The TISEAN package; CHAOS 9, 413-435 (1999)

M. T. Rosenstein, J. J. Collins, C. J. De Luca, A practical method for calculating largest Lyapunov exponents from small data sets, Physica D 65, 117 (1993)

## See Also

[mutual](#), [false.nearest](#) for the choice of optimal embedding parameters. [embedd](#) to perform embedding.

## Examples

```
output <-lyap_k(lorenz.ts, m=3, d=2, s=200, t=40, ref=1700, k=2, eps=4)
plot(output)
lyap(output, 0.73, 2.47)
```

---

mutual	<i>Average Mutual Information</i>
--------	-----------------------------------

---

## Description

Estimates the average mutual information index (ami) of a given time series for a specified number of lags

## Usage

```
mutual(series, partitions = 16, lag.max = 20, plot=TRUE, ...)
```

## Arguments

series	time series
partitions	number of bins
lag.max	largest lag
plot	logical. If 'TRUE' (the default) the ami is plotted
...	further arguments to be passed to the plot method

## Details

Estimates the mutual information index for a specified number of lags. The joint probability distribution function is estimated with a simple bi-dimensional density histogram.

## Value

An object of class "ami", which is a vector containing the estimated mutual information index for each lag between 0 and lag.max.



**Author(s)**

Antonio, Fabio Di Narzo

**References**

Hegger, R., Kantz, H., Schreiber, T., Practical implementation of nonlinear time series methods: The TISEAN package; CHAOS 9, 413-435 (1999)

**Examples**

```
mutual(lorenz.ts)
```

---

plot.ami

*Plotting average mutual information index*

---

**Description**

Plotting method for objects inheriting from class "ami".

**Usage**

```
## S3 method for class 'ami'  
plot(x, main = NULL, ...)
```

**Arguments**

x                    "ami" object  
main, ...            additional graphical arguments

**Details**

Plots the ami for each lag in x.

**Author(s)**

Antonio, Fabio Di Narzo

**See Also**

[mutual](#)

---

plot.d2                      *Plotting sample correlation integrals*

---

**Description**

Plotting method for objects inheriting from class "d2".

**Usage**

```
## S3 method for class 'd2'  
plot(x, ...)
```

**Arguments**

x                      "d2" object  
...                    additional graphical arguments

**Details**

Plots the sample correlation integrals in x in log-log scale, as a line for each considered embedding dimension.

**Author(s)**

Antonio, Fabio Di Narzo

**See Also**

[d2](#)

---

plot.false.nearest            *Plotting false nearest neighbours results*

---

**Description**

Plotting method for objects inheriting from class "false.nearest".

**Usage**

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

**Arguments**

x                      "false.nearest" object  
...                    additional graphical arguments

**Details**

Plots the results of [false.nearest](#).

**Author(s)**

Antonio, Fabio Di Narzo

**See Also**

[false.nearest](#)

---

print.d2

*Printing sample correlation integrals*

---

**Description**

Printing method for objects inheriting from class `"d2"`.

**Usage**

```
## S3 method for class 'd2'  
print(x, ...)
```

**Arguments**

x	"d2" object
...	additional arguments to 'print'

**Details**

Simply calls [plot.d2](#).

**Author(s)**

Antonio, Fabio Di Narzo

**See Also**

[plot.d2](#), [d2](#)

---

`print.false.nearest`     *Printing false nearest neighbours results*

---

### Description

Printing method for objects inheriting from class `"false.nearest"`.

### Usage

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

### Arguments

<code>x</code>	<code>"false.nearest"</code> object
<code>...</code>	additional arguments to <code>'print'</code>

### Details

Prints the table of results of [false.nearest](#).

### Author(s)

Antonio, Fabio Di Narzo

### See Also

[plot.false.nearest](#), [false.nearest](#)

---

`recurr`     *Recurrence plot*

---

### Description

Recurrence plot

### Usage

```
recurr(series, m, d, start.time=start(series), end.time=end(series), ...)
```

**Arguments**

series	time series
m	embedding dimension
d	time delay
start.time	starting time window (in time units)
end.time	ending time window (in time units)
...	further parameters to be passed to filled.contour

**Details**

Produces the recurrence plot, as proposed by Eckmann et al. (1987). White is maximum distance, black is minimum.

**warning**

Be aware that number of distances to store goes as  $n^2$ , where  $n = \text{length}(\text{window}(\text{series}, \text{start}=\text{start.time}, \text{end}=\text{end.time}))$ .

**Author(s)**

Antonio, Fabio Di Narzo

**References**

Eckmann J.P., Oliffson Kamphorst S. and Ruelle D., Recurrence plots of dynamical systems, Europhys. Lett., volume 4, 973 (1987)

**Examples**

```
recurr(lorenz.ts, m=3, d=2, start.time=15, end.time=20)
```

---

rossler.syst

*Roessler system of equations*


---

**Description**

Roessler system of equations

**Details**

To be used with [sim.cont.](#)

**Author(s)**

Antonio, Fabio Di Narzo

---

rossler.ts	<i>Roessler simulated time series, without noise</i>
------------	--

---

**Description**

Roessler simulated time series, without noise. Of each state of the system, we observe the first component.

**Details**

Roessler simulated time series, without noise, obtained with the call:

```
rossler.ts <- sim.cont(rossler.syst, start=0, end=650, dt=0.1, start.x=c(0,0,0), parms=c(0.15, 0.2, 1
```

**Author(s)**

Antonio, Fabio Di Narzo

---

sim.cont	<i>Simulates a continuous dynamic system</i>
----------	--

---

**Description**

Simulates a dynamic system of provided ODEs

**Usage**

```
sim.cont(syst, start.time, end.time, dt, start.x, parms=NULL, obs.fun=function(x) x[1])
```

**Arguments**

syst	ODE system
start.time	starting time
end.time	ending time
dt	time between observations
start.x	initial conditions
parms	parameters for the system
obs.fun	observed function of the state

**Details**

Simulates a dynamic system of provided ODEs. Uses lsoda in odesolve for numerical integration of the system.

**Value**

The time series of the observed function of the system's state

**Author(s)**

Antonio, Fabio Di Narzo

**See Also**

[lorenz.syst](#), [rossler.syst](#), [duffing.syst](#)

**Examples**

```
rossler.ts <- sim.cont(rossler.syst, start=0, end=650, dt=0.1, start.x=c(0,0,0), parms=c(0.15, 0.2, 10))
```

---

stplot

*Space-time separation plot*


---

**Description**

Space-time separation plot

**Usage**

```
stplot(series, m, d, idt=1, mdt)
```

**Arguments**

series	time series
m	embedding dimension
d	time delay
idt	observation steps in each iteration
mdt	number of iterations

**Details**

Produces the space-time separation plot, as introduced by Provenzale et al. (1992), which can be used to decide the Theiler time window  $t$ , which is required in many other algorithms in this package.

It plots the probability that two points in the reconstructed phase-space have distance smaller than epsilon in function of epsilon and of the time  $t$  between the points, as iso-lines at levels 10%, 20%, ..., 100%.

**Value**

lines of constant probability at 10%, 20%, ..., 100%.

**Author(s)**

Antonio, Fabio Di Narzo

**References**

Kantz H., Schreiber T., Nonlinear time series analysis. Cambridge University Press, (1997)

Provenzale A., Smith L. A., Vio R. and Murante G., Distinguishing between low-dimensional dynamics and randomness in measured time series. Physica D., volume 58, 31 (1992)

**See Also**

[false.nearest, d2, lyap\\_k](#)

**Examples**

```
stplot(rossler.ts, m=3, d=8, idt=1, mdt=250)
```



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