Package 'TSA'

September 13, 2020

Type Package
Title Time Series Analysis
Version 1.3
Date 2020-9-11
Author Kung-Sik Chan, Brian Ripley
Maintainer Kung-Sik Chan <kungsik.chan@gmail.com></kungsik.chan@gmail.com>
Depends R (>= $2.5.1$)
Imports leaps, locfit, mgcv, tseries
Description Contains R functions and datasets detailed in the book "Time Series Analysis with Applications in R (second edition)" by Jonathan Cryer and Kung-Sik Chan.
License GPL (>= 2)
URL https://stat.uiowa.edu/~kchan/TSA.htm RoxygenNote 7.1.1 NeedsCompilation no Repository CRAN Date/Publication 2020-09-13 15:00:11 UTC R topics documented:
it topics documented.
TSA-package 3 acf 4 airmiles 5 airpass 6 ar1.2.s 6 ar1.s 7 ar2.s 8 arima 8 arima.boot 10 arimax 11 arma11.s 13

ARMAspec	. 13
armasubsets	. 14
beersales	. 15
bluebird	. 15
bluebirdlite	. 16
boardings	. 17
BoxCox.ar	. 17
co2	. 18
color	. 19
CREF	. 19
cref.bond	. 20
days	. 20
deere1	. 21
deere2	. 21
deere3	. 22
detectAO	. 22
detectIO	. 24
eacf	. 25
eeg	. 26
electricity	. 27
euph	. 27
explode.s	. 28
fitted.Arima	. 28
flow	. 29
garch.sim	. 29
gBox	. 30
gold	. 31
google	. 32
hare	. 33
harmonic	. 33
hours	
ima22.s	. 35
JJ	. 35
Keenan.test	. 36
kurtosis	
lagplot	
larain	. 39
LB.test	. 39
ma1.1.s	. 40
ma1.2.s	. 41
ma2.s	. 41
McLeod.Li.test	
milk	. 43
oil.price	. 43
oilfilters	
periodogram	
plot.Arima	
plot.armasubsets	. 47

TSA-package 3

48

TSA-p	oackage	Time Series Analysis	
Index			7
	ziag		7
	•		7
	~		7
	veilleux		7
	usd.hkd		7
	units		7
	tuba		7
	U		7
	•		7
			6
	tlrt		6
	tempdub		6
	tbone		6
	tar.skeleton		6.
	tar.sim		6
	tar		6
	summary.armasubse	ts	6
	star		6
	spots1		6
	spots		6
	spec		5
	SP		5
	skewness		5'
	rwalk season		5: 5:
	runs		54 53
	rstandard.Arima		54
	robot		5.
	retail		5
	qar.sim		5
	prey.eq		5
	prewhiten		5
			49
	predict. IAR		4
	1' . 4 TAD		4

Description

Contains R functions and datasets detailed in the book "Time Series Analysis with Applications in R (second edition)" by J.D. Cryer and K.S. Chan

Details

4 acf

Package: TSA
Type: Package
Version: 1.3
Date: 2020-9-9

License: GPL version 2 or newer

Author(s)

Kung-Sik Chan, Brian Ripley

acf

Auto- and Cross- Covariance and -Correlation Function Estimation

Description

This function calls the acf function in the stats package and processes to drop lag-0 of the acf. It only works for univariate time series, so x below should be 1-dimensional.

Usage

```
acf(x, lag.max = NULL, type = c("correlation", "covariance", "partial")[1],
plot = TRUE, na.action = na.fail, demean = TRUE, drop.lag.0 = TRUE, ...)
```

Arguments

X	a univariate or multivariate (not ccf) numeric time series object or a numeric vector or matrix, or an "acf" object.
lag.max	maximum number of lags at which to calculate the acf. Default is $10*log10(N/m)$ where N is the number of observations and m the number of series.
type	character string giving the type of acf to be computed. Allowed values are "correlation" (the default), "covariance" or "partial".
plot	logical. If TRUE (the default) the acf is plotted.
na.action	function to be called to handle missing values. na.pass can be used.
demean	logical. Should the covariances be about the sample means?
drop.lag.0	logical. Should lag 0 be dropped
	further arguments to be passed to plot.acf.

airmiles 5

Value

An object of class "acf", which is a list with the following elements:

lag A three dimensional array containing the lags at which the acf is estimated.

An array with the same dimensions as lag containing the estimated acf.

type The type of correlation (same as the type argument).

n.used The number of observations in the time series.

series The name of the series x.

snames The series names for a multivariate time series.

Author(s)

Original authors of stats:::acf are: Paul Gilbert, Martyn Plummer, B.D. Ripley. This wrapper is written by Kung-Sik Chan

References

~put references to the literature/web site here ~

See Also

plot.acf, ARMAacf for the exact autocorrelations of a given ARMA process.

Examples

```
data(rwalk)
model1=lm(rwalk~time(rwalk))
summary(model1)
acf(rstudent(model1),main='')
```

airmiles

Monthly Airline Passenger-Miles in the US

Description

Monthly U.S. airline passenger-miles: 01/1996 - 05/2005.

Usage

```
data(airmiles)
```

Format

```
The format is: 'ts' int [1:113, 1] 30983174 32147663 38342975 35969113 36474391 38772238 40395657 41738499 33580773 36389842 ... - attr(*, "dimnames")=List of 2 ...$ : NULL ...$ : chr "airmiles" - attr(*, "tsp")= num [1:3] 1996 2005 12
```

6 ar1.2.s

Source

```
www.bts.gov/xml/air_traffic/src/index.xml#MonthlySystem
```

Examples

```
data(airmiles)
## maybe str(airmiles) ; plot(airmiles) ...
```

airpass

Monthly total international airline passengers

Description

Monthly total international airline passengers from 01/1960-12/1971.

Usage

```
data(airpass)
```

Format

The format is: Time-Series [1:144] from 1960 to 1972: 112 118 132 129 121 135 148 148 136 119 ...

Source

Box, G. E. P., Jenkins, G. M. and Reinsel, G. C. (1994) Time Series Analysis, Forecasting and Control. Second Edition. New York: Prentice-Hall.

Examples

```
data(airpass)
## maybe str(airpass) ; plot(airpass) ...
```

ar1.2.s

A simulated AR(1) series

Description

A simulated AR(1) series with the AR coefficient equal to 0.4.

```
data(ar1.2.s)
```

ar1.s 7

Format

The format is: Time-Series [1:60] from 1 to 60: -0.0678 1.4994 0.4888 0.3987 -0.5162 ...

Details

The model is Y(t)=0.4*Y(t-1)+e(t) where the e's are iid standard normal.

Examples

```
data(ar1.2.s)
## maybe str(ar1.2.s); plot(ar1.2.s) ...
```

ar1.s

A simulated AR(1) series

Description

A simulated AR(1) series with the AR coefficient equal to 0.9.

Usage

```
data(ar1.s)
```

Format

The format is: Time-Series [1:60] from 1 to 60: -1.889 -1.691 -1.962 -0.566 -0.627 ...

Details

The model is Y(t)=0.9*Y(t-1)+e(t) where the e's are iid standard normal.

```
data(ar1.s)
## maybe str(ar1.s) ; plot(ar1.s) ...
```

8 arima

ar2.s

Asimulated AR(2) series / time series

Description

Asimulated AR(2) series with AR coefficients being equal to 1.5 and -0.75

Usage

```
data(ar2.s)
```

Format

The format is: Time-Series [1:120] from 1 to 120: -2.064 -1.937 0.406 2.039 2.953 ...

Details

The model is Y(t)=1.5*Y(t-1)-0.75*Y(t-2)+e(t) where the e's are iid standard normal random variables.

Examples

```
data(ar2.s)
## maybe str(ar2.s) ; plot(ar2.s) ...
```

arima

Fitting an ARIMA model with Exogeneous Variables

Description

This function is identical to the arimax function which builds on and extends the capability of the arima function in R stats by allowing the incorporation of transfer functions, and innovative and additive outliers. For backward compatitibility, the function is also named arima. Note in the computation of AIC, the number of parameters excludes the noise variance. This function is heavily based on the arima function of the stats core of R.

```
arima(x, order = c(0, 0, 0), seasonal = list(order = c(0, 0, 0), period = NA), xreg = NULL, include.mean = TRUE, transform.pars = TRUE, fixed = NULL, init = NULL, method = c("CSS-ML", "ML", "CSS"), n.cond, optim.control = list(), kappa = 1e+06, io = NULL, xtransf, transfer = NULL)
```

arima 9

Arguments

x time series response
order regular ARIMA order
seasonal seasonal ARIMA order

xreg a dataframe containing covariates

include.mean if true, an intercept term is incorporated in the model; applicable only to station-

ary models.

transform.pars if true, the AR parameters are transformed to ensure stationarity

fixed a vector indicating which coefficients are fixed or free

init initial values

method estimation method

n.cond number of initial values to be conditioned on in a conditional analysis

optim.control control parameters for the optimization procedure kappa prior variance; used in dealing with initial values

All of the above parameters have the same usage as those in the arima function. Please check the help manual of the arima function. Below are new options.

io a list of time points at which the model may have an innovative outlier. The time

point of the outlier can be given either as absolute time point or as c(a,b), i.e. at the b-th 'month' of the a-th 'year' where each year has frequency(x) months,

assuming x is a time series.

xtransf xtranf is a matrix with each column containing a covariate that affects the time

series response in terms of an ARMA filter of order (p,q), i.e. if Z is one such covariate, its effect on the time series is $(theta_0+theta_1B+...+theta_{q-1}B^{q-1})/(1-phi_1B-...-phi_pB^p)Z_t$ In particular, if p=0 and q=1, this specifies a simple regression relationship, which should be included in xreg and not here. Note that the filter starts with zero initial values. Hence, it is pertinent to mean-delete

each distributed-lag covariate, and this is not done automatically.

transfer a list consisting of the ARMA orders for each transfer (distributed lag) covariate.

Value

An Arimax object contining the model fit.

Author(s)

Original author of the arima function in R stats: Brian Ripley. The arimax function is based on the stats:::arima function, with modifications by Kung-Sik Chan.

See Also

arima

10 arima.boot

Examples

```
data(hare)
arima(sqrt(hare),order=c(3,0,0))
```

arima.boot

Compute the Bootstrap Estimates of an ARIMA Model

Description

This function bootstraps time series according to the fitted ARMA(p,d,q) model supplied by the fitted object arima.fit, and estimate the same model using the arima function. Any bootstrap sample that has problem when fitted with the ARIMA model will be omitted from the final results and all error messages will be suppressed. You can check if there is any fitting problem by running the command geterrmessage().

Usage

```
arima.boot(arima.fit, cond.boot = FALSE, is.normal = TRUE, B = 1000, init, ntrans = 100)
```

Arguments

arima.fit	a fitted object from the arima function (seasonal components not allowed)
cond.boot	whether or not the bootstrap is conditional on the (p+d) initial values; if it is set true. If false (default), the stationary bootstrap is used.
is.normal	if true (default), errors are normally distributed, otherwise errors are drawn randomly and with replacement from the residuals of the fitted model.
В	number of bootstrap replicates (1000, default)
init	initial values for the bootstrap; needed if cond.boot=True default values are the initial values of the time series of the fitted model.
ntrans	number of transient values for the stationary bootstrap. Default=100

Value

a matrix each row of which consists of the coefficient estimates of a bootstrap time-series.

Author(s)

Kung-Sik Chan

arimax 11

Examples

```
data(hare)
arima.hare=arima(sqrt(hare),order=c(3,0,0))
boot.hare=arima.boot(arima.hare,B=50,init=sqrt(hare)[1:3],ntrans=100)
apply(boot.hare,2,quantile, c(.025,.975))
period.boot=apply(boot.hare,1,function(x){
roots=polyroot(c(1,-x[1:3]))
min1=1.e+9
rootc=NA
for (root in roots) {
if( abs(Im(root))<1e-10) next
if (Mod(root)< min1) {min1=Mod(root); rootc=root}</pre>
if(is.na(rootc)) period=NA else period=2*pi/abs(Arg(rootc))
period
})
hist(period.boot)
quantile(period.boot,c(0.025,.975))
```

arimax

Fitting an ARIMA model with Exogeneous Variables

Description

This function builds on and extends the capability of the arima function in R stats by allowing the incorporation of transfer functions, innovative and additive outliers. For backward compatitibility, the function is also named arima. Note in the computation of AIC, the number of parameters excludes the noise variance.

Usage

```
arimax(x, order = c(0, 0, 0), seasonal = list(order = c(0, 0, 0), period = NA), xreg = NULL, include.mean = TRUE, transform.pars = TRUE, fixed = NULL, init = NULL, method = c("CSS-ML", "ML", "CSS"), n.cond, optim.control = list(), kappa = 1e+06, io = NULL, xtransf, transfer = NULL)
```

Arguments

x time series response order regular ARIMA order seasonal seasonal ARIMA order

xreg a dataframe containing covariates

include.mean if true, an intercept term is incorporated in the model; applicable only to station-

ary model.

transform.pars if true, the AR parameters are transformed to ensure stationarity

12 arimax

fixed a vector indicating which coefficients are fixed or free

init initial values
method estimation method

n.cond number of initial values to be conditioned on a conditional analysis

optim.control control parameters for the optimization procedure kappa prior variance; used in dealing with initial values

All of the above parameters have the same usage as those in the arima function. Please check the help manual of the arima function. Below are new options.

io a list of time points at which the model may have an innovative outlier. The time

point of the outlier can be given either as absolute time point or as c(a,b), i.e. at the b-th 'month' of the a-th 'year' where each year has frequency(x) months,

assuming x is a time series.

xtransf xtranf is a matrix with each column containing a covariate that affects the time

series response in terms of an ARMA filter of order (p,q), i.e. if Z is one such covariate, its effect on the time series is $(theta_0+theta_1B+...+theta_{q-1}B^{q-1})/(1-phi_1B-...-phi_pB^p)Z_t$ In particular, if p=0 and q=1, this specifies a simple regression relationship, which should be included in xreg and not here. Note that the filter starts with zero initial values. Hence, it is pertinent to mean-delete

each distributed-lag covariate, which is not done automatically.

transfer a list consisting of the ARMA orders for each transfer (distributed lag) covariate.

Value

An Arimax object containing the model fit.

Author(s)

Original author of the arima function in R stats: Brian Ripley. The arimax function is based on the stats:::arima function, with modifications by Kung-Sik Chan.

See Also

arima

```
data(airmiles)
plot(log(airmiles),ylab='Log(airmiles)',xlab='Year', main='')
acf(diff(diff(window(log(airmiles),end=c(2001,8)),12)),lag.max=48,main='')
air.m1=arimax(log(airmiles),order=c(0,1,1),seasonal=list(order=c(0,1,1),
period=12),xtransf=data.frame(I911=1*(seq(airmiles)==69),
I911=1*(seq(airmiles)==69)),
transfer=list(c(0,0),c(1,0)),xreg=data.frame(Dec96=1*(seq(airmiles)==12),
Jan97=1*(seq(airmiles)==13),Dec02=1*(seq(airmiles)==84)),method='ML')
```

arma11.s

arma11.s

A Simulated ARMA(1,1) Series/ time series

Description

A simulated ARMA(1,1) series with the model given by: $y_t = 0.6 * y_{t-1} + e_t + 0.3 * e_{t-1}$ where the e's are iid standard normal random variables.

Usage

```
data(arma11.s)
```

Format

The format is: Time-Series [1:100] from 1 to 100: -0.765 1.297 0.668 -1.607 -0.626 ...

Examples

```
data(arma11.s)
## maybe str(arma11.s) ; plot(arma11.s) ...
```

ARMAspec

Theoretical spectral density function of a stationary ARMA model

Description

Computes and plots the theoretical spectral density function of a stationary ARMA model

Usage

```
ARMAspec(model, freq = seq(0, 0.5, 0.001), plot = TRUE, ...)
```

Arguments

model	an arma model
freq	vector of frequency over which the spectral density is computed
plot	if true, plot the spectral density function; default is true
	other parameters to be passed to the plot function

Value

a list:

spec spectral density values
freq same as freq in the input
model the arma model

14 armasubsets

Author(s)

Kung-Sik Chan

See Also

spec

Examples

```
theta=.9 # Reset theta for other MA(1) plots
ARMAspec(model=list(ma=-theta))
```

armasubsets

Selection of Subset ARMA Models

Description

This function finds a number of subset ARMA models. A "long" AR model is fitted to the data y to compute the residuals which are taken as a proxy of the error process. Then, an ARMA model is approximated by a regression model with the the covariates being the lags of the time series and the lags of the error process. Subset ARMA models may then be selected using the subset regression technique by leaps and bounds, via the regsubsets function of the leaps package in R.

Usage

```
armasubsets(y, nar, nma, y.name = "Y", ar.method = "ols", ...)
```

Arguments

y time-series data
nar maximum AR order
nma maximum MA order
y.name label of the time series

ar.method method used for fitting the long AR model; default is ols with the AR order

determined by AIC

... arguments passed to the plot.armasubsets function

Value

An object of the armasubsets class to be processed by the plot.armasubsets function.

Author(s)

Kung-Sik Chan

beersales 15

Examples

```
set.seed(92397)
test=arima.sim(model=list(ar=c(rep(0,11),.8),ma=c(rep(0,11),0.7)),n=120)
res=armasubsets(y=test,nar=14,nma=14,y.name='test',ar.method='ols')
plot(res)
```

beersales

Monthly beer sales / time series

Description

Monthly beer sales in millions of barrels, 01/1975 - 12/1990.

Usage

```
data(beersales)
```

Format

The format is: Time-Series [1:192] from 1975 to 1991: 11.12 9.84 11.57 13.01 13.42 ...

Source

Frees, E. W., Data Analysis Using Regression Models, Prentice Hall, 1996.

Examples

```
data(beersales)
## maybe str(beersales); plot(beersales) ...
```

bluebird

Blue Bird Potato Chip Data

Description

Weekly unit sales (log-transformed) of Bluebird standard potato chips (New Zealand) and their price for 104 weeks.

Usage

```
data(bluebird)
```

Format

```
The format is: mts [1:104, 1:2] 11.5 11.5 11.8 11.9 11.3 ... - attr(*, "dimnames")=List of 2 ..$: NULL ..$: chr [1:2] "log.sales" "price" - attr(*, "tsp")= num [1:3] 1 104 1 - attr(*, "class")= chr [1:2] "mts" "ts"
```

16 bluebirdlite

Source

www.stat.auckland.ac.nz/~balemi/Assn3.xls

Examples

```
data(bluebird)
## maybe str(bluebird) ; plot(bluebird) ...
```

bluebirdlite

Bluebird Lite potato chip data

Description

Weekly unit sales (log-transformed) of Bluebird Lite potato chips (New Zealand) and their price for 104 weeks.

Usage

```
data(bluebirdlite)
```

Format

A data frame with 104 observations on the following 2 variables.

```
log.sales a numeric vector price a numeric vector
```

Source

www.stat.auckland.ac.nz/~balemi/Assn3.xls

```
data(bluebirdlite)
## maybe str(bluebirdlite); plot(bluebirdlite) ...
```

boardings 17

boardings	Monthly public transit boardings and gasoline price in Denver

Description

Monthly public transit boardings (mostly buses and light rail) and gasoline price (both log-transformed), Denver, Colorado region, 08/2000 - 03/2006.

Source

Personal communication from Lee Cryer, Project Manager, Regional Transportation District, Denver, Colorado. Denver gasoline prices were obtained from the Energy Information Administration, U.S. Department of Energy, Washington, D.C.at www.eia.doe.gov

Examples

```
data(boardings)
plot(boardings)
## maybe str(boardings) ; plot(boardings) ...
```

BoxCox.ar

Determine the power transformation for serially correlated data

Description

Determine the appropriate power transformation for time-series data. The objective is to estimate the power transformation so that the transformed time series is approximately a Gaussian AR process.

Usage

```
BoxCox.ar(y, order, lambda = seq(-2, 2, 0.01), plotit = TRUE,
method = c("mle", "yule-walker", "burg", "ols", "yw"), ...)
```

Arguments

У	univariate time series (must be positive)
order	AR order for the data; if missing, the order is determined by AIC for the log-transformed data
lambda	a vector of candidate power transformation values; if missing, it is set to be from -2 to 2, with increment .01 $$
plotit	logical value, if true, plot the profile log-likelihood for the power estimator
method	method of AR estimation; default is "mle"
	other parameters to be passed to the ar function

18 co2

Value

A list that contains the following:

lambda candidate power transformation parameter values

loglike profile log-likelihood

mle maximum likelihood estimate of the power transformation value

ci 95% C.I. of the power transformation value

Note

The procedure is very computer intensive. Be patient for the outcome

Author(s)

Kung-Sik Chan

Examples

```
data(hare)
# hare.transf=BoxCox.ar(y=hare)
# hare.transf$ci
```

co2

Levels of Carbon Dioxide at Alert, Canada / Time series

Description

Monthly CO2 level at Alert, Northwest Territories, Canada, near the Artic Circle, 01/1994 - 12/2004.

Usage

```
data(co2)
```

Format

```
The format is: Time-Series [1:132] from 1994 to 2005: 363 364 365 364 364 ...
```

Source

```
http://cdiac.ornl.gov/trends/co2/sio-alt.htm
```

```
data(co2)
## maybe str(co2); plot(co2) ...
```

color 19

color

Color property/time series

Description

Color property from 35 consecutive batches in an industrial process.

Usage

```
data(color)
```

Format

The format is: Time-Series [1:35] from 1 to 35: 67 63 76 66 69 71 72 71 72 72 ...

Source

"The Estimation of Sigma for an X Chart", Journal of Quality Technology, Vol. 22, No. 3 (July 1990), by Jonathan D. Cryer and Thomas P. Ryan.

Examples

```
data(color)
## maybe str(color) ; plot(color) ...
```

CREF

Daily CREF Values

Description

Daily values of one unit of the CREF (College Retirement Equity Fund) Stock fund, 08/26/04 - 08/15/06.

Usage

```
data(CREF)
```

Format

```
The format is: Time-Series [1:501] from 1 to 501: 170 170 169 170 171 ...
```

Source

www.tiaa-cref.org/performance/retirement/data/index.html

```
data(CREF)
## maybe str(CREF) ; plot(CREF) ...
```

20 days

cref.bond

Daily CREF Bond Values

Description

Daily values of one unit of the CREF (College Retirement Equity Fund) Bond fund, 08/26/04 - 08/15/06.

Usage

```
data(CREF)
```

Source

www.tiaa-cref.org/performance/retirement/data/index.html

Examples

```
data(CREF)
## maybe str(CREF) ; plot(CREF) ...
```

days

Number of days between payment to Winegard Corp. / time series

Description

Accounts receivable data. Number of days until a distributor of Winegard Company products pays their account.

Usage

```
data(days)
```

Format

The format is: Time-Series [1:130] from 1 to 130: 39 39 41 26 28 28 25 26 24 38 ...

Source

Personal communication from Mark Selergren, Vice President, Winegard, Inc., Burlington, Iowa.

```
data(days)
## maybe str(days) ; plot(days) ...
```

deere1 21

deere1

Deviations of an industrial process at Deere \& Co. - Series 1

Description

82 consecutive values for the amount of deviation (in 0.000025 inch units) from a specified target value in an industrial machining process at Deere \& Co.

Usage

```
data(deere1)
```

Format

The format is: Time-Series [1:82] from 1 to 82: 3 0 -1 -4 7 3 7 3 3 -1 ...

Source

Personal communication from William F. Fulkerson, Deere \& Co. Technical Center, Moline, Illinois.

Examples

```
data(deere1)
## maybe str(deere1) ; plot(deere1) ...
```

deere2

Deviations of an industrial process at Deere \& Co. − Series 2

Description

102 consecutive values for the deviation (in 0.0000025 inch units) from a specified target value.

Usage

```
data(deere2)
```

Format

```
The format is: Time-Series [1:102] from 1 to 102: -18 -24 -17 -27 -37 -34 -8 14 18 7 ...
```

Source

Personal communication from William F. Fulkerson, Deere \& Co. Technical Center, Moline, Illinois.

22 detectAO

Examples

```
data(deere2)
## maybe str(deere2) ; plot(deere2) ...
```

deere3

Deviations of an industrial process at Deere \& Co. - Series 3

Description

Fifty seven consecutive values for the deviation (in 0.0000025 inch units) from a specified target value.

Usage

```
data(deere3)
```

Format

The format is: Time-Series [1:57] from 1 to 57: -500 -1250 -500 -3000 -2375 ...

Source

Personal communication from William F. Fulkerson, Deere \& Co. Technical Center, Moline, Illinois.

Examples

```
data(deere3)
## maybe str(deere3); plot(deere3) ...
```

detectA0

Additive Outlier Detection

Description

This function serves to detect whether there are any additive outliers (AO). It implements the test statistic $lambda_{2,t}$ proposed by Chang, Chen and Tiao (1988).

```
detectAO(object, alpha = 0.05, robust = TRUE)
```

detectAO 23

Arguments

object a fitted ARIMA model

alpha family significance level (5% is the default) Bonferroni rule is used to control

the family error rate.

robust if true, the noise standard deviation is estimated by mean absolute residuals

times sqrt(pi/2). Otherwise, it is the estimated by sqrt(sigma2) from the arima

fit.

Value

A list containing the following components:

ind the time indices of potential AO lambda2 the corresponding test statistics

Author(s)

Kung-Sik Chan

References

Chang, I.H., Tiao, G.C. and C. Chen (1988). Estimation of Time Series Parameters in the Presence of Outliers. Technometrics, 30, 193-204.

See Also

```
detectI0
```

```
set.seed(12345)
y=arima.sim(model=list(ar=.8,ma=.5),n.start=158,n=100)
y[10]
y[10]=10
y=ts(y,freq=1,start=1)
plot(y,type='o')
acf(y)
pacf(y)
eacf(y)
m1=arima(y,order=c(1,0,0))
m1
detectAO(m1)
detectAO(m1, robust=FALSE)
detectIO(m1)
```

24 detectIO

detectIO	Innovative Outlier Detection

Description

This function serves to detect whether there are any innovative outliers (IO). It implements the test statistic $lambda_{2,t}$ proposed by Chang, Chen and Tiao (1988).

Usage

```
detectIO(object, alpha = 0.05, robust = TRUE)
```

Arguments

object a fitted ARIMA model

alpha family significance level (5% is the default) Bonferroni rule is used to control

the family error rate.

robust if true, the noise standard deviation is estimated by mean absolute residuals

times sqrt(pi/2). Otherwise, it is the estimated by sqrt(sigma2) from the arima

fit.

Value

A list containing the following components:

ind the time indices of potential AO lambda1 the corresponding test statistics

Author(s)

Kung-Sik Chan

References

Chang, I.H., Tiao, G.C. and C. Chen (1988). Estimation of Time Series Parameters in the Presence of Outliers. Technometrics, 30, 193-204.

See Also

detectI0

eacf 25

Examples

```
set.seed(12345)
y=arima.sim(model=list(ar=.8,ma=.5),n.start=158,n=100)
y[10]
y[10]=10
y=ts(y,freq=1,start=1)
plot(y,type='o')
acf(y)
pacf(y)
eacf(y)
m1=arima(y,order=c(1,0,0))
m1
detectAO(m1)
detectAO(m1, robust=FALSE)
detectIO(m1)
```

eacf

Compute the sample extended acf (ESACF)

Description

Computes the sample extended acf (ESACF) for the time series stored in z. The matrix of ESACF with the AR order up to ar.max and the MA order up to ma.max is stored in the matrix EACFM.

Usage

```
eacf(z, ar.max = 7, ma.max = 13)
```

Arguments

z the time series data
ar.max maximum AR order; default=7
ma.max maximum MA order; default=13

Value

A list containing the following two components:

eacf a matrix of sample extended ACF

symbol corresponding matrix of symbols indicating the significance of the ESACF

Side effect of the eacf function: The function prints a coded ESACF table with significant values denoted by * and nosignificant values by 0.

Author(s)

Kung-Sik Chan

26 eeg

References

Tsay, R. and Tiao, G. (1984). "Consistent Estimates of Autoregressive Parameters and Extended Sample Autocorrelation Function for Stationary and Nonstationary ARMA Models." Journal of the American Statistical Association, 79 (385), pp. 84-96.

Examples

```
data(arma11.s)
eacf(arma11.s)
```

eeg

EEG Data

Description

An electroencephalogram (EEG) is a noninvasive test used to detect and record the electrical activity generated in the brain. These data were measured at a frequency of 256 per second and came from a patient suffering a seizure. This a portion of a series on the website of Professor Richard Smith, University of North Carolina. His source: Professors Mike West and Andrew Krystal, Duke University.

Usage

```
data(eeg)
```

Format

```
The format is: ts [1:13000, 1] -3.08 -20.15 -45.05 -69.95 -94.57 ... - attr(*, "dimnames")=List of 2 ...$ : NULL ...$ : chr "eeg" - attr(*, "tsp")= num [1:3] 2001 15000 1
```

Source

```
http://www.stat.unc.edu/faculty/rs/s133/Data/datadoc.html
```

```
data(eeg)
## maybe str(eeg) ; plot(eeg) ...
```

electricity 27

electricity

Monthly US electricity production / time series

Description

Monthly U.S. electricity generation (in millions of kilowatt hours) of all types: coal, natural gas, nuclear, petroleum, and wind, 01/1973 - 12/2005.

Usage

```
data(electricity)
```

Format

The format is: 'ts' int [1:396, 1] 160218 143539 148158 139589 147395 161244 173733 177365 156875 154197 ... - attr(*, "dimnames")=List of 2 ..\$: NULL ..\$: chr "electricity" - attr(*, "tsp")= num [1:3] 1973 2006 12

Source

Source: www.eia.doe.gov/emeu/mer/elect.html

Examples

```
data(electricity)
## maybe str(electricity) ; plot(electricity) ...
```

euph

A digitized sound file of a B flat played on a euphonium

Description

A digitized sound file of about 0.4 seconds of a B flat just below middle C played on a euphonium by one of the authors (JDC), a member of the group Tempered Brass.

Usage

```
data(euph)
```

Format

The format is: Time-Series [1:1105] from 1 to 1105: 0.244 0.635 0.712 0.608 0.317 ...

```
data(euph)
## maybe str(euph) ; plot(euph) ...
```

28 fitted.Arima

explode.s

A simulated explosive AR(1) series

Description

A simulated AR(1) series with the AR(1) coefficient being 3.

Usage

```
data(explode.s)
```

Format

The format is: Time-Series [1:8] from 1 to 8: 0.63 0.64 3.72 12.67 39.57 ...

Examples

```
data(explode.s)
## maybe str(explode.s); plot(explode.s) ...
```

fitted.Arima

Fitted values of an arima model.

Description

Computes the fitted values of an arima model.

Usage

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

Arguments

object a fitted model from the arima function.

... other arguments; not used here but kept to be consistent with the generic method

Value

fitted values

Author(s)

Kung-Sik Chan

flow 29

See Also

```
arima
```

Examples

```
data(hare)
hare.m1=arima(sqrt(hare),order=c(3,0,0))
fitted(hare.m1)
```

flow

Monthly River Flow for the Iowa River

Description

Flow data (in cubic feet per second) for the Iowa river measured at Wapello, Iowa for the period 09/1958 - 08/2006.

Usage

```
data(flow)
```

Source

```
http://waterdata.usgs.gov/ia/nwis/sw
```

Examples

```
data(flow)
## maybe str(flow) ; plot(flow) ...
```

garch.sim

Simulate a GARCH process

Description

Simulate a GARCH process.

```
garch.sim(alpha, beta, n = 100, rnd = rnorm, ntrans = 100,...)
```

30 gBox

Arguments

alpha	The vector of ARCH coefficients including the intercept term as the first element
beta	The vector of GARCH coefficients
n	sample size
rnd	random number generator for the noise; default is normal
ntrans	burn-in size, i.e. number of initial simulated data to be discarded
	parameters to be passed to the random number generator

Details

Simulate data from the GARCH(p,q) model: $x_t = \sigma_{t|t-1}e_t$ where $\{e_t\}$ is iid, e_t independent of past $x_{t-s}, s=1,2,\ldots$, and

$$\sigma_{t|t-1} = \sum_{j=1}^{p} \beta_j \sigma_{t-j|t-j-1} + \alpha_0 + \sum_{j=1}^{q} \alpha_j x_{t-i}^2$$

Value

simulated GARCH time series of size n.

Author(s)

Kung-Sik Chan

Examples

```
set.seed(1235678)
garch01.sim=garch.sim(alpha=c(.01,.9),n=500)
plot(garch01.sim,type='l', main='',ylab=expression(r[t]),xlab='t')
```

gBox

Generalized Portmanteau Tests for GARCH Models

Description

Perform a goodness-of-fit test for the GARCH model by checking whether the standardized residuals are iid based on the ACF of the absolute residuals or squared residuals.

```
gBox(model, lags = 1:20, x, method = c("squared", "absolute")[1], plot = TRUE)
```

gold 31

Arguments

model fitted model from the garch function of the tseries library lags a vector of maximum ACF lags to be used in the test x time series data to which the GARCH model is fitted

method "squared": test is based on squared residuals; "absolute": test is based on abso-

lute residuals

plot logical variable, if TRUE, the p-values of the tests are plotted

Value

lags in the input

pvalue a vector of p-values of the tests

method method used

X X

Author(s)

Kung-Sik Chan

References

"Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

Examples

```
require(tseries) # need to uncomment this line when running the example
data(CREF)
r.cref=diff(log(CREF))*100
m1=tseries::garch(x=r.cref,order=c(1,1))
summary(m1)
gBox(m1,x=r.cref,method='squared')
```

gold

Gold Price / time series

Description

Daily price of gold (in \\$ per troy ounce) for the 252 trading days of 2005

```
data(gold)
```

32 google

Format

The format is: Time-Series [1:252] from 1 to 252: 427 426 426 423 421 ...

Source

www.lbma.org.uk/2005dailygold.htm

Examples

```
data(gold)
## maybe str(gold) ; plot(gold) ...
```

google

Daily returns of the google stock

Description

Daily returns of the google stock from 08/20/04 - 09/13/06.

Usage

```
data(google)
```

Format

The format is: Time-Series [1:521] from 1 to 521: 0.0764 0.0100 -0.0423 0.0107 0.0179 ...

Source

```
http://finance.yahoo.com/q/hp?s=GOOG
```

```
data(google)
## maybe str(google) ; plot(google) ...
```

hare 33

hare

Canadian hare data/ time series

Description

Annual number of hare data.

Usage

data(hare)

Format

The format is: Time-Series [1:31] from 1905 to 1935: 50 20 20 22 27 50 55 78 70 59 ...

Details

These are yearly hare abundances for the main drainage of the Hudson Bay, based on trapper questionnaires.

Source

MacLulich, D. A. (1937) Fluctuations in the Number of the Varying Hare (Lepus americanus) (Univ. of Toronto Press, Toronto)

References

Stenseth, N. C., Falck, W., Bjornstad, O. N. and Krebs. C. J. (1997) Population regulation in snowshoe hare and Canadian lynx: Asymmetric food web configurations between hare and lynx. Proc. Natl. Acad. Sci., 94, 5147-5152.

Examples

data(hare)

harmonic

Construct harmonic functions for fitting harmonic trend model

Description

The function creates a matrix of the first m pairs of harmonic functions for fitting a harmonic trend (cosine-sine trend, Fourier regresssion) models with the response being x, a time series.

```
harmonic(x, m = 1)
```

34 hours

Arguments

x a time series

m the number of pairs of harmonic functions to be created; 2m must be less than

or equal to the frequency of x

Value

```
a matrix consisting of \cos(2k\pi t), \sin(2k\pi t), k=1,2,...,m, excluding any zero functions.
```

Author(s)

Kung-Sik Chan

See Also

season

Examples

```
data(tempdub)
# first creates the first pair of harmonic functions and then fit the model
har.=harmonic(tempdub,1)
model4=lm(tempdub~har.)
summary(model4)
```

hours

Average hours worked in US manufacturing sector / time series

Description

Average hours worked (times 10) in U.S. manufacturing sector, from 07/1982 - 06/1987

Usage

```
data(hours)
```

Format

```
The format is: Time-Series [1:60] from 1983 to 1987: 389 390 389 390 393 397 392 388 396 398 ...
```

Source

Cryer, J. D. Time Series Analysis, Duxbury Press, 1986.

```
data(hours)
## maybe str(hours) ; plot(hours) ...
```

ima22.s

ima22.s

Simulated IMA(2,2) series / time series

Description

A simulated IMA(2,2) series with theta1=1 and theta2=-0.6

Usage

```
data(ima22.s)
```

Format

The format is: Time-Series [1:62] from 1 to 62: 0.00000 0.00000 -0.00569 2.12404 2.15337 ...

Examples

```
data(ima22.s)
## maybe str(ima22.s) ; plot(ima22.s) ...
```

JJ

Quarterly earnings per share for the Johnson \& Johnson Company

Description

Quarterly earnings per share for 1960Q1 to 1980Q4 of the U.S. company, Johnson \& Johnson, Inc.

Usage

```
data(JJ)
```

Format

The format is: Time-Series [1:84] from 1960 to 1981: $0.71\ 0.63\ 0.85\ 0.44\ 0.61\ 0.69\ 0.92\ 0.55\ 0.72\ 0.77\ ...$

Source

```
http://www.stat.pitt.edu/stoffer/tsa2/
```

```
data(JJ)
## maybe str(JJ) ; plot(JJ) ...
```

36 Keenan.test

Keenan.test

Keenan's one-degree test for nonlinearity

Description

Carry out Keenan's 1-degree test for nonlinearity against the null hypothesis that the time series follows some AR process.

Usage

```
Keenan.test(x, order, ...)
```

Arguments

x time series

order working AR order; if missing, it is estimated by minimizing AIC via the ar

function.

... user-supplied options to the ar function.

Details

The test is designed to have optimal local power against depature from the linear autoregressive function in the direction of the square of the linear autoregressive function.

Value

A list containing the following components

test.stat The observed test statistic
p.value p-value of the test
order working AR order

Author(s)

Kung-Sik Chan

References

Keenan, D. M. (1985), A Tukey nonadditivity-type test for time series Nonlinearity, Biometrika, 72, 39-44.

See Also

```
Tsay.test,tlrt
```

```
data(spots)
Keenan.test(sqrt(spots))
```

kurtosis 37

kurtosis

Kurtosis

Description

Computes the Kurtosis.

Usage

```
kurtosis(x, na.rm = FALSE)
```

data

Arguments

X

na.rm

logical variable, if true, missing values are excluded from analysis

Details

Given data x_1, x_2, \dots, x_n , the sample kurtosis is defined by the formula:

$$\frac{\sum_{i=1}^{n} (x_i - \bar{x})^4 / n}{(\sum_{i=1}^{n} (x_i - \bar{x})^2 / n)^2} - 3.$$

Value

The function returns the kurtosis of the data.

Author(s)

Kung-Sik Chan

```
data(CREF)
r.cref=diff(log(CREF))*100
kurtosis(r.cref)
```

38 lagplot

_		_	
- 1	ag	n	~ t
- 1	a٧	L) I	LOI

Lagged Regression Plot

Description

Computes and plots the nonparametric regression function of a time series against its various lags.

Usage

```
lagplot(x, lag.max = 6, deg = 1, nn = 0.7, method = c("locfit", "gam", "both")[1])
```

Arguments

x	time series
lag.max	maximum lag
deg	degree of local polynomial, needed only for the locfit method
nn	fraction of nearest data contained in a window, needed only for the locfit method
method	Two methods for nonparametric estimation: "locfit" is the default which uses

Two methods for nonparametric estimation: "locfit" is the default which uses the local polynomial approach via the locfit library to estimate the conditional mean function of $E(X_t|X_{t-k}=x)$ for $1 \le k \le lag.max$; Another method is

GAM, via the mgcv library.

Value

Side effects: The nonparametric lagged regression functions are plotted lag by lag, with the raw data superimposed on the plots.

Author(s)

Kung-Sik Chan

References

"Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

```
set.seed(2534567)
par(mfrow=c(3,2))
y=arima.sim(n=61,model=list(ar=c(1.6,-0.94),ma=-0.64))
# lagplot(y)
```

larain 39

larain

Annual rainfall in Los Angeles / time series

Description

Annual precipitation (in inches) in Los Angeles, 1878-1992.

Usage

```
data(larain)
```

Format

The format is: Time-Series [1:115] from 1778 to 1892: 20.86 17.41 18.65 5.53 10.74 ...

Source

Personal communication from Professor Donald Bentley, Pomona College, Claremont, California. For more data see http://www.wrh.noaa.gov/lox/climate/cvc.php

Examples

```
data(larain)
## maybe str(larain) ; plot(larain) ...
```

LB.test

Portmanteau Tests for Fitted ARIMA models

Description

This function modifies the Box.test function in the stats package, and it computes the Ljung-Box or Box-Pierce tests checking whether or not the residuals appear to be white noise.

Usage

```
LB.test(model, lag = 12, type = c("Ljung-Box", "Box-Pierce"), no.error = FALSE, omit.initial = TRUE)
```

Arguments

model model fit from the arima function

lag number of lags of the autocorrelation of the residuals to be included in the test

statistic. (default=12)

type either Ljung-Box or Box-Pierce

no.error a system variable; normally it is not changed

omit.initial if true, (d+Ds) initial residuals are omitted from the test

40 ma1.1.s

Value

a list:

statistics test statistic p.value p-value

parameter d.f. of the Chi-square test

lag no of lags

Author(s)

Kung-Sik Chan, based on A. Trapletti's work on the Box.test function in the stats package

References

Box, G. E. P. and Pierce, D. A. (1970), Distribution of residual correlations in autoregressive-integrated moving average time series models. Journal of the American Statistical Association, 65, 15091526.

Ljung, G. M. and Box, G. E. P. (1978), On a measure of lack of fit in time series models. Biometrika 65, 553564.

Examples

```
data(color)
m1.color=arima(color,order=c(1,0,0))
LB.test(m1.color)
```

ma1.1.s

A simulated MA(1) series / time series

Description

A simulated MA(1) series with the MA(1) coefficient equal to 0.9.

Usage

```
data(ma1.1.s)
```

Format

The format is: Time-Series [1:120] from 1 to 120: 0.182 -0.748 -0.355 1.014 -2.363 ...

Details

The model is Y(t) = e(t) - 0.9e(t-1) where the e's are iid standard normal.

```
data(ma1.1.s)
## maybe str(ma1.1.s); plot(ma1.1.s) ...
```

ma1.2.s 41

ma1.2.s

A simulated MA(1) series / time series

Description

A simulated MA(1) series with the MA(1) coefficient equal to -0.9.

Usage

```
data(ma1.2.s)
```

Format

The format is: Time-Series [1:120] from 1 to 120: 1.511 1.821 0.957 -1.538 -2.888 ...

Details

The model is Y(t) = e(t) + 0.9e(t-1) where the e's are iid standard normal.

Examples

```
data(ma1.2.s)
## maybe str(ma1.2.s) ; plot(ma1.2.s) ...
```

ma2.s

A simulated MA(2) series

Description

A simulated MA(2) series with MA coefficients being 1 and -0.6.

Usage

```
data(ma2.s)
```

Format

The format is: Time-Series [1:120] from 1 to 120: -0.4675 0.0815 0.9938 -2.6959 2.8116 ...

Details

The model is Y(t) = e(t) - e(t-1) + 0.6 * e(t-2) where the e's are iid standard normal random variables.

```
data(ma2.s)
## maybe str(ma2.s); plot(ma2) ...
```

42 McLeod.Li.test

McLeod.Li.test	McLeod-Li test
TICLEOU.LI.LEST	MCDeou-Di iesi

Description

Perform the McLeod-Li test for conditional heteroscedascity (ARCH).

Usage

```
McLeod.Li.test(object, y, gof.lag, col = "red", omit.initial = TRUE,
plot = TRUE, ...)
```

Arguments

object	a fitted Arima model, ususally the output from the arima function. If supplied, then the Mcleod-Li test is applied to the residuals of the model, and the yargument is ignored.
у	time series data with which one wants to test for the presence of conditional heteroscedascity
gof.lag	maximum number of lags for which the test is carried out.
col	color of the reference line
omit.initial	suppress the initial (d+Ds) residuals if set to be TRUE
plot	suppress plotting if set to be FALSE
	other arguments to be passed to the plot function

Details

The test checks for the presence of conditional heteroscedascity by computing the Ljung-Box (port-manteau) test with the squared data (if y is supplied and object suppressed) or with the squared residuals from an arima model (if an arima model is passed to the function via the object argument.)

Value

pvlaues the vector of p-values for the Ljung-Box test statistics computed using the first

m lags of the ACF of the squared data or residuals, for m ranging from 1 to

gof.lag.

Author(s)

Kung-Sik Chan

References

McLeod, A. I. and W. K. Li (1983). Diagnostic checking ARMA time series models using squared residual autocorrelations. Journal of Time Series Analysis, 4, 269273.

milk 43

Examples

```
data(CREF)
r.cref=diff(log(CREF))*100
McLeod.Li.test(y=r.cref)
```

milk

Monthly Milk Production

Description

Average monthly milk production per cow in the US, 01/1994 - 12/2005

Usage

```
data(milk)
```

Format

```
The format is: 'ts' int [1:144, 1] 1343 1236 1401 1396 1457 1388 1389 1369 1318 1354 ... - attr(*, "dimnames")=List of 2 ... : NULL ... : chr "milk" - attr(*, "tsp")= num [1:3] 1994 2006 12
```

Examples

```
data(milk)
str(milk)
plot(milk)
```

oil.price

Monthly Oil Price / time series

Description

Monthly spot price for crude oil, Cushing, OK (in U.S. dollars per barrel), 01/1986 - 01/2006.

Usage

```
data(oil.price)
```

Format

The format is: Time-Series [1:241] from 1986 to 2006: 22.9 15.4 12.6 12.8 15.4 ...

Source

tonto.eia.doe.gov/dnav/pet/hist/rwtcM.htm

44 periodogram

Examples

```
data(oil.price)
## maybe str(oil.price) ; plot(oil.price) ...
```

oilfilters

Monthly sales to dealers of a specialty oil filter/time series

Description

Monthly wholesale specialty oil filters sales, Deere & Co, 07/1983 - 06/1987.

Usage

```
data(oilfilters)
```

Format

```
The format is: Time-Series [1:48] from 1984 to 1987: 2385 3302 3958 3302 2441 3107 5862 4536 4625 4492 ... - attr(*, "freq")= num 12 - attr(*, "start")= num [1:2] 1987 7
```

Source

Data courtesy of William F. Fulkerson, Deere & Company, Technical Center, Moline, Illinois.

Examples

```
data(oilfilters)
## maybe str(oilfilters) ; plot(oilfilters) ...
```

periodogram

Computing the periodogram

Description

This is a wrapper that computes the periodogram

```
periodogram(y,log='no',plot=TRUE,ylab="Periodogram", xlab="Frequency",lwd=2,...)
```

plot.Arima 45

Arguments

у	A univariate time series
log	if set to "yes", the periodogram is plotted on the log-scale; default="no"
plot	The periodogram is plotted if it is set to be TRUE which is the default
ylab	label on the y-axis
xlab	label on the x-axis
lwd	thickness of the periodogram lines
	other arguments to be passed to the plot function

Value

A list that contains the following elements:

freq Vector of frequencies at which the spectral density is estimated. (Possibly ap-

proximate Fourier frequencies.

spec Vector of estimates of the periodogram at frequencies corresponding to freq.

References

```
Bloomfield, P. (1976) Fourier Analysis of Time Series: An Introduction. Wiley.
Brockwell, P. J. and Davis, R. A. (1991) Time Series: Theory and Methods. Second edition.
```

Examples

Springer.

```
data(star)
plot(star,xlab='Day',ylab='Brightness')
periodogram(star,ylab='Variable Star Periodogram'); abline(h=0)
```

plot.Arima

Compute and Plot the Forecasts Based on a Fitted Time Series Model

Description

Plots the time series data and its predictions with 95% prediction bounds.

```
## S3 method for class 'Arima'
plot(x, n.ahead = 12, col = "black", ylab = object$series,
lty = 2, n1, newxreg, transform, Plot=TRUE, ...)
```

46 plot.Arima

Arguments

x a fitted arima model

n. ahead number of prediction steps ahead (default=12)

col color of the prediction bounds

ylab label of the y-axis

line type of the point predictor; default=dashed lines
n1 starting time point of the plot (default=earliest time point)

newxreg a matrix of covariate(s) over the period of prediction

transform function used to transform the forecasts and their prediction bounds; if missing,

no transformation will be carried out. This option is useful if the model was fitted to the transformed data and it is desirable to obtain the forecasts on the original scale. For example, if the model was fitted with the logarithm of the data, then transform = exp will plot the forecasts and their prediction bounds on

the original scale.

Plot Plotting will be suppressed if Plot is set to be FALSE; default is TRUE

... additional parameters passed to the plot function

Value

Side effects of the function: plot the forecasts and their 95% prediction bounds, unless Plot is set to be FALSE. The part of the observed series is plotted with all data plotted as open circles and linked by a smooth line. By default the predicted values are plotted as open circles joined up by a dashed line. The plotting style of the predicted values can be altered by supplying relevant plotting options, e.g specifying the options type='o', pch=19 and lty=1 will plot the predicted values as solid circles that are overlaid on the connecting smooth solid line. The prediction limits are plotted as dotted lines, with default color being black. However, the prediction limits can be drawn in other colors. For example, setting col='red' paints the prediction limits in red. An interesting use of the col argument is setting col=NULL which has the effect of not drawing the prediction limits.

The function returns an invisible list containing the following components.

pred the time series of predicted values

the corresponding lower 95% prediction limits the corresponding upper 95% prediction limits

Author(s)

Kung-Sik Chan

```
data(oil.price)
oil.IMA11alt=arima(log(oil.price),order=c(0,1,1),
# create the design matrix of the covariate for prediction
xreg=data.frame (constant=seq(oil.price)))
n=length(oil.price)
n.ahead=24
```

plot.armasubsets 47

```
newxreg=data.frame(constant=(n+1):(n+n.ahead))
# do the prediction and plot the results
plot(oil.IMA11alt,n.ahead=n.ahead,newxreg=newxreg,
ylab='Log(Oil Price)',xlab='Year',n1=c(2000,1))
# do the same thing but on the orginal scale
plot(oil.IMA11alt,n.ahead=n.ahead,newxreg=newxreg,
ylab='Oil Price',xlab='Year',n1=c(2000,1),transform=exp,pch=19, lty=1,type='o')
# Setting pch=19 plots the predicted values as solid circles.
res=plot(oil.IMA11alt,n.ahead=n.ahead,newxreg=newxreg,
ylab='Oil Price',xlab='Year',n1=c(2000,1),transform=exp,pch=19,col=NULL)
# Setting col=NULL will make the prediction bands invisible. Try col='red'.
res
# prints the predicted values and their 95% prediction limits.
```

plot.armasubsets

Plot the Best Subset ARMA models

Description

This function is adapted from the plot.regsubsets function of the leaps package, and its main use is to plot the output from the armasubsets function.

Usage

```
## S3 method for class 'armasubsets'
plot(x, labels = obj$xnames, main = NULL,
scale = c("BIC", "AICc", "AIC", "Cp", "adjR2", "R2"),
col = gray(c(seq(0.4, 0.7, length = 10), 0.9)), draw.grid = TRUE,
axis.at.3 = TRUE, ...)
```

Arguments

x	an object of class armasubsets
labels	variable names
main	title for plot
scale	which summary statistic to use for ordering plots
col	the last color should be close to but distinct from white
draw.grid	a logical argument; if it is true (default), gray grid lines are superimposed on the graph.
axis.at.3	a logical argument; if if it is true (default), the x-labels are drawn on the upper horizontal axis.
	other arguments

Value

Plot the few best subset ARMA models.

48 predict.TAR

Author(s)

Kung-Sik Chan, based on previoud work by Thomas Lumley and Merlise Clyde

See Also

armasubsets

Examples

```
set.seed(53331) test=arima.sim(model=list(ar=c(rep(0,11),.8),ma=c(rep(0,11),0.7)),n=120) res=armasubsets(y=test,nar=14,nma=14,y.name='test',ar.method='ols') plot(res)
```

plot1.acf

Plot1

Description

A workhorse function for the acf function in the TSA pacakage.

Author(s)

Kung-Sik Chan

predict.TAR

Prediction based on a fitted TAR model

Description

Predictions based on a fitted TAR model. The errors are assumed to be normally distributed. The predictive distributions are approximated by simulation.

Usage

```
## S3 method for class 'TAR'
predict(object, n.ahead = 1, n.sim = 1000,...)
```

Arguments

object a fitted TAR model from the tar function

n.ahead number of prediction steps ahead

n.sim simulation size

... other arguments; not used here but kept for consistency with the generic method

prescrip 49

Value

fit a vector of medians of the 1-step to n.ahead-step predictive distributions

pred.interval a matrix whose i-th row consists of the 2.5 and 97.5 precentiles of the i-step

predictive distribution

pred.matrix a matrix whose j-th column consists of all simulated values from the j-step pre-

dictive distribution

Author(s)

Kung-Sik Chan

References

"Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

See Also

tar

Examples

```
data(prey.eq)
prey.tar.1=tar(y=log(prey.eq),p1=4,p2=4,d=3,a=.1,b=.9,print=TRUE)
set.seed(2357125)
pred.prey=predict(prey.tar.1,n.ahead=60,n.sim=1000)
yy=ts(c(log(prey.eq),pred.prey$fit),frequency=1,start=1)
plot(yy,type='n',ylim=range(c(yy,pred.prey$pred.interval)),ylab='Log Prey',
xlab=expression(t))
lines(log(prey.eq))
lines(window(yy, start=end(prey.eq)[1]+1),lty=2)
lines(ts(pred.prey$pred.interval[2,],start=end(prey.eq)[1]+1),lty=2)
lines(ts(pred.prey$pred.interval[1,],start=end(prey.eq)[1]+1),lty=2)
```

prescrip

Cost per prescription / time series

Description

Monthly U.S. average prescription costs for the months 08/1986 - 03/1992.

Usage

```
data(prescrip)
```

Format

The format is: Time-Series [1:68] from 1987 to 1992: 14.5 14.7 14.8 14.6 14.3 ...

50 prewhiten

Source

Frees, E. W., Data Analysis Using Regression Models, Prentice Hall, 1996.

Examples

```
data(prescrip)
## maybe str(prescrip) ; plot(prescrip) ...
```

prewhiten

Prewhiten a Bivariate Time Series, and Compute and Plot Their Sample Cross-Correlation Function

Description

The bivariate time series are prewhitened according to an AR model fitted to the x-component of the bivariate series. Alternatively, if an ARIMA model is provided, it will be used to prewhiten both series. The CCF of the prewhitened bivariate series is then computed and plotted.

Usage

```
prewhiten(x, y, x.model = ar.res,ylab="CCF", ...)
```

Arguments

х	first component series
у	second component series
x.model	an ARIMA model; if provided, it is used to prewhiten both series. Otherwise, an AR model is fitted to the x-series and used to prewhiten both series. The AR order is chosen by minimizing the AIC and the fit carried out by the ar.ols function.
ylab	label of y-axis; default is "CCF"
	additional parameters to be passed to the ar.ols and the ccf function.

Value

A list containing the following components:

ccf	Output from the ccf function on the prewhitened data.
ar	The AR model fit to the x-series, or x.model if it is provided.

Author(s)

Kung-Sik Chan

prey.eq 51

Examples

```
data(milk)
data(electricity)
milk.electricity=ts.intersect(milk,log(electricity))
plot(milk.electricity,yax.flip=TRUE,main='')
ccf(as.numeric(milk.electricity[,1]),as.numeric(milk.electricity[,2]),
main='milk & electricity',ylab='CCF')
me.dif=ts.intersect(diff(diff(milk,12)),diff(diff(log(electricity),12)))
prewhiten(as.numeric(me.dif[,1]),as.numeric(me.dif[,2]),
,ylab='CCF')
```

prey.eq

Prey series / time series

Description

The stationary part of the Didinium series in the veilleux data frame.

Usage

```
data(prey.eq)
```

Format

The format is: Time-Series [1:57] from 7 to 35: 26.9 53.2 65.6 81.2 143.9 ...

See Also

veilleux

Examples

```
data(prey.eq)
## maybe str(prey.eq) ; plot(prey.eq) ...
```

qar.sim

Simulate a first-order quadratic AR model

Description

Simulates a first-order quadratic AR model with normally distributed noise.

```
qar.sim(const = 0, phi0 = 0, phi1 = 0.5, sigma = 1, n = 20, init = 0)
```

52 retail

Arguments

const	intercept
phi0	coefficient of the lag 1
phi1	coefficient of the squared lag 1
sigma	noise standard deviation
n	sample size
init	number of burn-in values

Details

The quadratic AR(1) model specifies that

$$Y_t = \text{const} + \phi_0 Y_{t-1} + \phi_1 Y_{t-1}^2 + e_t$$

where e_t are iid normally distributed with zero mean and standard deviation σ . If $\sigma = 0$, the model is deterministic.

Value

A simulated series from the quadratic AR(1) model, as a vector

Author(s)

Kung-Sik Chan

See Also

```
tar.sim
```

Examples

```
set.seed(1234567) \\ plot(y=qar.sim(n=15,phi1=.5,sigma=1),x=1:15,type='l',ylab=expression(Y[t]),xlab='t') \\ y=qar.sim(n=100,const=0.0,phi0=3.97,phi1=-3.97,sigma=0,init=.377) \\ plot(y,x=1:100,type='l',ylab=expression(Y[t]),xlab='t') \\ acf(y,main='') \\ \end{cases}
```

retail

U.K. retail sales / time series

Description

Monthly total UK (United Kingdom) retail sales (non-food stores in billions of pounds), 01/1983 - 12/1987.

```
data(retail)
```

robot 53

Format

The format is: Time-Series [1:60] from 1983 to 1988: 81.3 78.9 93.8 94 97.8 1.6 99.6 1.2 98 1.7 ...

Source

www.statistics.gov.uk/statbase/TSDdownload1.asp

Examples

```
data(retail)
## maybe str(retail) ; plot(retail) ...
```

robot

The distance of a robot from a desired position / time series

Description

Final position in the x direction of an industrial robot put through a series of planned exercises many times.

Usage

```
data(robot)
```

Format

The format is: Time-Series [1:324] from 1 to 324: 0.0011 0.0011 0.0024 0 -0.0018 0.0055 0.0055 -0.0015 0.0047 -1e-04 ...

Source

Personal communication from William F. Fulkerson, Deere \& Co. Technical Center, Moline, Illinois.

```
data(robot)
## maybe str(robot) ; plot(robot) ...
```

54 runs

rstandard.Arima

Compute the Standardized Residuals from a Fitted ARIMA Model

Description

Computes the internally standardized residuals from a fitted ARIMA model.

Usage

```
## S3 method for class 'Arima'
rstandard(model,...)
```

Arguments

model model fitted by the arima function
... not used; kept here for consistency with the generic method

Details

```
residuals/(error std. dev.)
```

Value

time series of standarized residuals

Examples

```
data(oil.price)
m1.oil=arima(log(oil.price),order=c(0,1,1))
plot(rstandard(m1.oil),ylab='Standardized residuals',type='l')
abline(h=0)
```

runs

Runs test

Description

Test the independence of a sequence of random variables by checking whether there are too many or too few runs above (or below) the median.

```
runs(x,k=0)
```

rwalk 55

Arguments

x time series

k the value above or below which runs are counted; default is zero, so data is

assumed to have zero median

Details

The runs test examines the data in sequence to look for patterns that would give evidence against independence. Runs above or below k are counted. A small number of runs would indicate that neighboring values are positively dependent and tend to hang together over time. On the other hand, too many runs would indicate that the data oscillate back and forth across their median of zero. Then neighboring residuals are negatively dependent. So either too few or too many runs lead us to reject independence. When applied to residuals, the runs test is useful for model diagnostics.

Value

pvalue p-value of the test

observed.runs observed number of runs expected.runs expected number of runs

n1 number of data less than or equal to k

n2 number of data above k

Author(s)

Kung-Sik Chan

Examples

```
data(tempdub)
month.=season(tempdub) # the period sign is included to make the printout from
# the following command clearer.
model3=lm(tempdub~month.) # intercept is automatically included so one month (Jan) is dropped
summary(model3)
runs(rstudent(model3))
```

rwalk

A simulated random walk / Time series

Description

A simulated random walk with standard normal increments

```
data(rwalk)
```

56 season

Examples

```
data(rwalk)
## maybe str(rwalk) ; plot(rwalk) ...
```

season

Extract the season info from a time series

Description

Extract the season info from a equally spaced time series and create a vector of the season info. For example for monthly data, the function outputs a vector containing the months of the data.

Usage

```
season(x, labels)
```

Arguments

x a time series

labels the user supplied labels for the seasons

Details

The time series must have frequency greater than 1, otherwise the function will stop and issue an error message. If labels is missing, labels will be set as follows: It is set to be c("1Q","2Q","3Q","4Q) if the frequency of x equals 4, c("January",...,"December") if the frequency equals 12, and c("Monday",...,"Sunday") if frequency equals 7. Otherwise, it is set to be c("S1",...)

Value

An invisible vector containing the seasons of the data

Author(s)

Kung-Sik Chan

See Also

harmonic

```
data(tempdub)
month.=season(tempdub) # the period sign is included to make the printout from
# the commands two line below clearer; ditto below.
model2=lm(tempdub~month.-1) # -1 removes the intercept term
summary(model2)
```

skewness 57

skewness

Skewness

Description

Computes the skewness of the data

Usage

```
skewness(x, na.rm = FALSE)
```

Arguments

Χ

data

na.rm

logical variable, if true, missing values are excluded from analysis

Details

Given data x_1, x_2, \dots, x_n , the sample skewness is defined by the formula:

$$\frac{\sum_{i=1}^{n} (x_i - \bar{x})^3 / n}{(\sum_{i=1}^{n} (x_i - \bar{x})^2 / n)^{3/2}}.$$

Value

The function returns the skewness of the data.

Author(s)

Kung-Sik Chan

```
data(CREF)
r.cref=diff(log(CREF))*100
skewness(r.cref)
```

58 spec

SP

Quarterly Standard \& Poor's Composite Index of stock price values / time series

Description

Quarterly S&P Composite Index, 1936Q1 - 1977Q4.

Usage

```
data(SP)
```

Format

The format is: Time-Series [1:168] from 1936 to 1978: 149 148 160 172 179 ...

Source

Frees, E. W., Data Analysis Using Regression Models, Prentice Hall, 1996.

Examples

```
data(SP)
## maybe str(SP) ; plot(SP) ...
```

spec

Computing the spectrum

Description

This is a wrapper that allows the user to invoke either the spec.pgram function or the spec.ar function in the stats pacakge. Note that the seasonal attribute of the data, if it exists, will be removed, for our preferred way of presenting the output.

```
spec(x, taper = 0, detrend = FALSE, demean = TRUE, method = c("pgram",
   "ar")[1], ci.plot = FALSE, ylim = range(c(lower.conf.band, upper.conf.band)),
   ...)
```

spec 59

Arguments

A list that contains the following:

A univariate or multivariate time series

xaper amount of taper; 0 is the default

detrend logical; if True, the data are detrended; default is False demean logical; if True, the data are centered; default is True

method String specifying the method used to estimate the spectral density. Allowed

methods are "pgram" (the default) and "ar".

ci.plot logical; if True, the 95% confidence band will be plotted.

ylim Plotting parameter vector specifying the minimum and maximum of the y-axis.

... other arguments

Value

The output is from the spec.pgram function or spec.ar function, and the following description of the output is taken from the help manual of the spec function in the stats package. An object of class "spec", which is a list containing at least the following components:

freq Vector of frequencies at which the spectral density is estimated. (Possibly ap-

proximate Fourier frequencies.) The units are the reciprocal of cycles per unit

time (and not per observation spacing): see Details below.

spec Vector (for univariate series) or matrix (for multivariate series) of estimates of

the spectral density at frequencies corresponding to freq. coh NULL for univariate series. For multivariate time series, a matrix containing the squared coherency between different series. Column i + (j - 1) * (j - 2)/2 of coh contains

the squared coherency between columns i and j of x, where i < j.

phase NULL for univariate series. For multivariate time series a matrix containing the

cross-spectrum phase between different series. The format is the same as coh.

series The name of the time series.

snames For multivariate input, the names of the component series.

method The method used to calculate the spectrum.

The result is returned invisibly if plot is true.

References

Bloomfield, P. (1976) Fourier Analysis of Time Series: An Introduction. Wiley.

Brockwell, P. J. and Davis, R. A. (1991) Time Series: Theory and Methods. Second edition. Springer.

Venables, W. N. and Ripley, B. D. (2002) Modern Applied Statistics with S-PLUS. Fourth edition. Springer. (Especially pages 3927.)

60 spots

Examples

```
set.seed(271435); n=200; phi=-0.6
y=arima.sim(model=list(ar=phi),n=n)
k=kernel('daniell',m=15)
sp=spec(y,kernel=k,main='',sub='',xlab='Frequency',
ylab='Log(Smoothed Sample Spectrum)',ci.plot=TRUE,ci.col='black')
lines(sp$freq,ARMAspec(model=list(ar=phi),sp$freq,plot=FALSE)$spec,lty=4)
abline(h=0)
```

spots

Relative annual sunspot number / time series

Description

Annual American (relative) sunspot numbers collected from 1945 to 2007. The annual (relative) sunspot number is a weighted average of solar activities measured from a network of observatories.

Usage

```
data(spots)
```

Format

The format is: Time-Series [1:61] from 1945 to 2005: 32.3 99.9 170.9 166.6 174.1 ...

Source

http://www.ngdc.noaa.gov/stp/SOLAR/ftpsunspotnumber.html#american

References

"Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

```
data(spots)
## maybe str(spots) ; plot(spots) ...
```

spots1 61

spots1

Annual international sunspot numbers

Description

Annual international sunspot numbers, NOAA National Geophysical Data Center, 1700 - 2005.

Usage

```
data(spots1)
```

Format

```
The format is: ts [1:306, 1] 5 11 16 23 36 58 29 20 10 8 ... - attr(*, "dimnames")=List of 2 ..$: NULL ..$: chr "spots" - attr(*, "tsp")= num [1:3] 1700 2005 1
```

Source

```
ftp://ftp.ngdc.noaa.gov/STP/SOLAR_DATA/SUNSPOT_NUMBERS/YEARLY.PLT
```

Examples

```
data(spots1)
## maybe str(spots1) ; plot(spots1) ...
```

star

Star Brightness

Description

Brightness (magnitude) of a particular star at midnight on 600 consecutive nights.

Usage

```
data(star)
```

Source

Whittaker, E. T. and Robinson, G., (1924). The Calculus of Observations. London: Blackie and Son.

```
data(star)
## maybe str(star) ; plot(star) ...
data(star)
plot(star,xlab='Day',ylab='Brightness')
```

62 tar

summary.armasubsets

Summary of output from the armasubsets function

Description

Add the calculation of AIC and AICc. See the help manual of regsubsets function of the leaps package

Usage

```
## S3 method for class 'armasubsets'
summary(object, all.best = TRUE, matrix = TRUE, matrix.logical = FALSE,
df = NULL, ...)
```

Arguments

object	armasubsets object
all.best	Show all the best subsets or just one of each size
matrix	Show a matrix of the variables in each model or just summary statistics
matrix.logical	With matrix=TRUE, the matrix is logical TRUE/FALSE or string "*"/code" "
df	Specify a number of degrees of freedom for the summary statistics. The default is $n\text{-}1$
	Other arguments for future methods

Author(s)

Kung-Sik Chan, based on previous work of Thomas Lumley

tar

Estimation of a TAR model

Description

Estimation of a two-regime TAR model.

```
tar(y, p1, p2, d, is.constant1 = TRUE, is.constant2 = TRUE, transform = "no",
  center = FALSE, standard = FALSE, estimate.thd = TRUE, threshold,
method = c("MAIC", "CLS")[1], a = 0.05, b = 0.95, order.select = TRUE, print = FALSE)
```

tar 63

Arguments

У	time series
p1	AR order of the lower regime
p2	AR order of the upper regime
d	delay parameter
is.constant1	if True, intercept included in the lower regime, otherwise the intercept is fixed at zero
is.constant2	similar to is.constant1 but for the upper regime
transform	available transformations: "no" (i.e. use raw data), "log", "log10" and "sqrt"
center	if set to be True, data are centered before analysis
standard	if set to be True, data are standardized before analysis
estimate.thd	if True, threshold parameter is estimated, otherwise it is fixed at the value supplied by threshold
threshold	known threshold value, only needed to be supplied if estimate.thd is set to be False.
method	"MAIC": estimate the TAR model by minimizing the AIC; "CLS": estimate the TAR model by the method of Conditional Least Squares.
a	lower percent; the threshold is searched over the interval defined by the $a*100$ percentile to the $b*100$ percentile of the time-series variable
b	upper percent
order.select	If method is "MAIC", setting order.select to True will enable the function to further select the AR order in each regime by minimizing AIC
print	if True, the estimated model will be printed

Details

The two-regime Threshold Autoregressive (TAR) model is given by the following formula:

$$Y_t = \phi_{1,0} + \phi_{1,1}Y_{t-1} + \dots + \phi_{1,p}Y_{t-p_1} + \sigma_1 e_t, \text{ if } Y_{t-d} \le r$$

$$Y_t = \phi_{2,0} + \phi_{2,1}Y_{t-1} + \dots + \phi_{2,p_2}Y_{t-p} + \sigma_2 e_t, \text{ if } Y_{t-d} > r.$$

where r is the threshold and d the delay.

Value

A list of class "TAR" which can be further processed by the by the predict and tsdiag functions.

Author(s)

Kung-Sik Chan

References

Tong, H. (1990) "Non-linear Time Series, a Dynamical System Approach," Clarendon Press Oxford "Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

64 tar.sim

See Also

```
predict.TAR, tsdiag.TAR, tar.sim, tar.skeleton
```

Examples

```
data(prey.eq)
prey.tar.1=tar(y=log(prey.eq),p1=4,p2=4,d=3,a=.1,b=.9,print=TRUE)
```

tar.sim

Simulate a two-regime TAR model

Description

Simulate a two-regime TAR model.

Usage

```
tar.sim(object, ntransient = 500, n = 500, Phi1, Phi2, thd, d, p, sigma1,
sigma2, xstart = rep(0, max(p,d)), e)
```

Arguments

object	a TAR model fitted by the tar function; if it is supplied, the model parameters and initial values are extracted from it
ntransient	the burn-in size
n	sample size of the simulated series
Phi1	the coefficient vector of the lower-regime model
Phi2	the coefficient vector of the upper-regime model
thd	threshold
d	delay
р	maximum autoregressive order
sigma1	noise std. dev. in the lower regime
sigma2	noise std. dev. in the upper regime
xstart	initial values for the simulation
е	standardized noise series of size equal to length(xstart)+ntransient+n; if missing, it will be generated as some normally distributed errors

Details

The two-regime Threshold Autoregressive (TAR) model is given by the following formula:

$$Y_t = \phi_{1,0} + \phi_{1,1} Y_{t-1} + \ldots + \phi_{1,p} Y_{t-p_1} + \sigma_1 e_t, \text{ if } Y_{t-d} \le r$$

$$Y_t = \phi_{2,0} + \phi_{2,1} Y_{t-1} + \ldots + \phi_{2,p_2} Y_{t-p} + \sigma_2 e_t, \text{ if } Y_{t-d} > r.$$

where r is the threshold and d the delay.

tar.skeleton 65

Value

A list containing the following components:

```
y simulated TAR series
```

e the standardized errors

...

Author(s)

Kung-Sik Chan

References

Tong, H. (1990) "Non-linear Time Series, a Dynamical System Approach," Clarendon Press Oxford "Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

See Also

tar

Examples

```
set.seed(1234579)
y=tar.sim(n=100,Phi1=c(0,0.5),
Phi2=c(0,-1.8),p=1,d=1,sigma1=1,thd=-1,
sigma2=2)$y
plot(y=y,x=1:100,type='b',xlab="t",ylab=expression(Y[t]))
```

tar.skeleton

Find the asympotite behavior of the skeleton of a TAR model

Description

The skeleton of a TAR model is obtained by suppressing the noise term from the TAR model.

```
tar.skeleton(object, Phi1, Phi2, thd, d, p, ntransient = 500, n = 500,
xstart, plot = TRUE,n.skeleton = 50)
```

66 tar.skeleton

Arguments

object a TAR model fitted by the tar function; if it is supplied, the model parameters

and initial values are extracted from it

ntransient the burn-in size

n sample size of the skeleton trajectory

Phi1 the coefficient vector of the lower-regime model
Phi2 the coefficient vector of the upper-regime model

thd threshold d delay

p maximum autoregressive order

xstart initial values for the iteration of the skeleton

plot if True, the time series plot of the skeleton is drawn

n. skeleton number of last n.skeleton points of the skeleton to be plotted

Details

The two-regime Threshold Autoregressive (TAR) model is given by the following formula:

$$Y_t = \phi_{1,0} + \phi_{1,1}Y_{t-1} + \ldots + \phi_{1,p}Y_{t-p_1} + \sigma_1 e_t$$
, if $Y_{t-d} \le r$

$$Y_t = \phi_{2,0} + \phi_{2,1}Y_{t-1} + \ldots + \phi_{2,p_2}Y_{t-p} + \sigma_2 e_t$$
, if $Y_{t-d} > r$.

where r is the threshold and d the delay.

Value

A vector that contains the trajectory of the skeleton, with the burn-in discarded.

Author(s)

Kung-Sik Chan

References

Tong, H. (1990) "Non-linear Time Series, a Dynamical System Approach," Clarendon Press Oxford. "Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

See Also

tar

```
data(prey.eq)
prey.tar.1=tar(y=log(prey.eq),p1=4,p2=4,d=3,a=.1,b=.9,print=TRUE)
tar.skeleton(prey.tar.1)
```

tbone 67

tbone

A digitized sound file of a B flat played on a tenor trombone

Description

A digitized sound file of about 0.4 seconds of a B flat just below middle C played on a tenor trombone by Chuck Kreeb, a member of Tempered Brass and a friend of one of the authors.

Usage

```
data(tbone)
```

Format

The format is: Time-Series [1:17689] from 1 to 17689: 0.0769 0.0862 0.0961 0.1050 0.1129 ...

Examples

```
data(tbone)
## maybe str(tbone) ; plot(tbone) ...
```

tempdub

Monthly average temperature in Dubuque/time series

Description

Monthly average temperature (in degrees Fahrenheit) recorded in Dubuque 1/1964 - 12/1975.

Usage

```
data(tempdub)
```

Format

The format is: Time-Series [1:144] from 1964 to 1976: 24.7 25.7 30.6 47.5 62.9 68.5 73.7 67.9 61.1 48.5 ...

Source

```
http://mesonet.agron.iastate.edu/climodat/index.phtml?station=ia2364&report=16
```

```
data(tempdub)
## maybe str(tempdub) ; plot(tempdub) ...
```

68 tlrt

tlrt

Likelihood ratio test for threshold nonlinearity

Description

Carry out the likelihood ratio test for threshold nonlinearity, with the null hypothesis being a normal AR process and the alternative hypothesis being a TAR model with homogeneous, normally distributed errors.

Usage

```
tlrt(y, p, d = 1, transform = "no", a = 0.25, b = 0.75,...)
```

Arguments

у	time series
р	working AR order
d	delay
transform	available transformations: "no" (i.e. use raw data), "log", "log10" and "sqrt"
a	lower percent; the threshold is searched over the interval defined by the a*100 percentile to the b*100 percentile of the time-series variable
b	upper percent
• • •	other arguments to be passed to the ar function which determines the Ar order, if p is missing

Details

The search for the threshold parameter may be narrower than that defined by the user as the function attempts to ensure adequate sample size in each regime of the TAR model. The p-value of the test is based on large-sample approximation and also is more reliable for small p-values.

Value

p.value	p-value of the test
test.statistic	likelihood ratio test statistic
a	the actual lower fraction that defines the interval of search for the threshold; it may differ from the a specified by the user
b	the actual upper fraction that defines the interval of search for the threshold

Author(s)

Kung-Sik Chan

Tsay.test 69

References

Chan, K.S. (1990). Percentage points of likelihood ratio tests for threshold autoregression. Journal of Royal Statistical Society, B 53, 3, 691-696.

See Also

```
Keenan.test, Tsay.test
```

Examples

```
data(spots)
pvaluem=NULL
for (d in 1:5){
  res=tlrt(sqrt(spots),p=5,d=d,a=0.25,b=0.75)
  pvaluem= cbind( pvaluem, round(c(d,signif(c(res$test.statistic, res$p.value))),3))
}
rownames(pvaluem)=c('d','test statistic','p-value')
pvaluem
```

Tsay.test

Tsay's Test for nonlinearity

Description

Carry out Tsay's test for quadratic nonlinearity in a time series.

Usage

```
Tsay.test(x, order, ...)
```

Arguments

```
    x time series
    order working linear AR order; if missing, it will be estimated via the ar function by minimizing AIC
    ... options to be passed to the ar function
```

Details

The null hypothesis is that the true model is an AR process. The AR order, if missing, is estimated by minimizing AIC via the ar function, i.e. fitting autoregressive model to the data. The default fitting method of the ar function is "yule-walker."

70 tsdiag.Arimax

Value

A list containing the following components

test.stat The observed test statistic

p.value p-value of the test order working AR order

Author(s)

Kung-Sik Chan

References

Tsay, R. S. (1986), Nonlinearity test for time series, Biometrika, 73, 461-466.

See Also

```
Tsay.test,tlrt
```

Examples

```
data(spots)
Tsay.test(sqrt(spots))
```

tsdiag.Arimax

Model Diagnostics for a Fitted ARIMAX Model

Description

This function is modified from the tsdiag function of the stats package.

Usage

```
## S3 method for class 'Arimax'
tsdiag(object, gof.lag, tol = 0.1, col = "red", omit.initial = TRUE,...)
```

Arguments

object	a fitted ARIMAX model
gof.lag	maximum lag used in ACF and Ljung-Box tests for the residuals
tol	tolerance (default=0.1); see below
col	color of some warning lines in the figures (default=red)
omit.initial	suppress the initial (d+Ds) residuals if true
	other arguments to be passed to the acf function

tsdiag.TAR 71

Value

Side effects: Plot the time plot of the standardized residuals. Red dashed line at +/-qnorm(0.025/no of data) are added to the plot. Data beyond these lines are deemed outliers, based on the Bonferronni rule. The ACF of the standardized residuals is plotted. The p-values of the Ljung-Box test are plotted using a variety of the first K residuals. K starts at the lag on and beyond which the moving-average weights (in the MA(infinity) representation) are less than tol.

Author(s)

Kung-Sik Chan, based on the tsdiag function of the stats pacakage

Examples

```
data(color)
m1.color=arima(color,order=c(1,0,0))
tsdiag(m1.color,gof=15,omit.initial=FALSE)
```

tsdiag.TAR

Model diagnostics for a fitted TAR model

Description

The time series plot and the sample ACF of the standardized residuals are plotted. Also, a portmanteau test for detecting residual correlations in the standardized residuals are carried out.

Usage

```
## S3 method for class 'TAR'
tsdiag(object, gof.lag, col = "red",xlab = "t", ...)
```

Arguments

object a fitted TAR model output from the tar function
gof.lag number of lags of ACF to be examined
col color of the lines flagging outliers, etc.
xlab x labels for the plots
... any additional user-supplied options to be passed to the acf function

Value

Side effects: plot the time-series plot of the standardized residuals, their sample ACF and portmanteau test for residual autocorrelations in the standardized errors.

Author(s)

Kung-Sik Chan

72 tuba

References

"Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

See Also

tar

Examples

```
data(prey.eq)
prey.tar.1=tar(y=log(prey.eq),p1=4,p2=4,d=3,a=.1,b=.9,print=TRUE)
tsdiag(prey.tar.1)
```

tuba

A digitized sound file of a B flat played on a BB flat tuba

Description

A digitized sound file of about 0.4 seconds of a B flat an octave and one whole step below middle C played on a BB flat tuba by Linda Fisher, a member of Tempered Brass and a friend one of the authors.

Usage

```
data(tuba)
```

Format

The format is: Time-Series [1:4402] from 1 to 4402: 0.217 0.209 0.200 0.195 0.196 ...

```
data(tuba)
## maybe str(tuba) ; plot(tuba) ...
```

units 73

units

Annual sales of certain large equipment

Description

Annual sales of certain large equipment, 1983 - 2005.

Usage

```
data(units)
```

Format

```
The format is: ts [1:24, 1] 71.7 78.6 111.1 125.6 133.0 ... - attr(*, "tsp")= num [1:3] 1982 2005 1 - attr(*, "dimnames")=List of 2 ..$ : NULL ..$ : chr "Units"
```

Source

Proprietary sales data from a large international company

Examples

```
data(units)
## maybe str(units); plot(units) ...
```

usd.hkd

Daily US Dollar to Hong Kong Dollar Exchange Rates

Description

Daily USD/HKD (US dollar to Hong Kong dollar) exchange rate from January 1, 2005 to March 7, 2006

Usage

```
data(usd.hkd)
```

Format

A data frame with 431 observations on the following 6 variables.

```
r daily returns of USD/HKD exchange rates
```

v estimated conditional variances based on an AR(1)+GARCH(3,1) model

hkrate daily USD/HKD exchange rates

outlier1 dummy variable of day 203, corresponding to July 22, 2005

outlier2 dummy variable of day 290, another possible outlier

day calendar day

74 veilleux

Source

```
http://www.oanda.com/convert/fxhistory
```

References

"Time Series Analysis, with Applications in R" by J.D. Cryer and K.S. Chan

Examples

```
data(usd.hkd)
## maybe str(usd.hkd) ; plot(usd.hkd) ...
```

veilleux

An experimental prey-predator time series

Description

A data frame consisting of bivariate time series from an experiment for studying prey-predator dynamics. The first time series consists of the numbers of prey individuals (Didinium natsutum) per ml measured every twelve hours over a period of 35 days; the second time series consists of the corresponding number of predators (Paramecium aurelia) per ml.

Usage

```
data(veilleux)
```

Format

```
The format is: mts [1:71, 1:2] 15.7 53.6 73.3 93.9 115.4 ... - attr(*, "dimnames")=List of 2 ..$: NULL ..$: chr [1:2] "Didinium" "Paramecium" - attr(*, "tsp")= num [1:3] 0 35 2 - attr(*, "class")= chr [1:2] "mts" "ts"
```

Source

```
http://www-personal.buseco.monash.edu.au/~hyndman/TSDL/
```

References

Veilleux (1976) "The analysis of a predatory interaction between Didinium and Paramecium", Masters thesis, University of Alberta.

Jost & Ellner (2000) "Testing for predator dependence in predator-prey dynamics: a non-parametric approach", Proceedings of the Royal Society of London B, 267, 1611-1620.

```
data(veilleux)
## maybe str(veilleux) ; plot(veilleux) ...
```

wages 75

wages

Average hourly wages in the apparel industry / time series

Description

Average hourly wages in the apparel industry, from 07/1981 - 06/1987.

Usage

```
data(wages)
```

Format

The format is: Time-Series [1:72] from 1982 to 1987: $4.92\ 4.96\ 5.04\ 5.05\ 5.04\ 5.04\ 5.18\ 5.13\ 5.15\ 5.18\ ...$

Source

Cryer, J. D. Time Series Analysis, Duxbury Press, 1986.

Examples

```
data(wages)
## maybe str(wages) ; plot(wages) ...
```

winnebago

Monthly unit sales of recreational vehicles / time series

Description

Monthly unit sales of recreational vehicles from Winnebago, Inc., Forrest City, Iowa, from 11/1966 - 02/1972.

Usage

```
data(winnebago)
```

Format

The format is: Time-Series [1:64] from 1967 to 1972: 61 48 53 78 75 58 146 193 124 120 ...

Source

Roberts, H. V., Data Analysis for Managers with Minitab, second edition, The Scientific Press, 1991.

76 zlag

Examples

```
data(winnebago)
## maybe str(winnebago) ; plot(winnebago) ...
```

zlag

Compute the lag of a vector.

Description

Computes the lag of a vector, with missing elements replaced by NA

Usage

```
zlag(x, d= 1)
```

Arguments

x vector

d compute the lag d of x

Value

A vector whose k-th element equals x[k-d] with x[t]=NA for t<=0

Author(s)

Kung-Sik Chan

```
##--- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.
x=1:5
zlag(x,2)
```

Index

* datasets	retail, 52
airmiles, 5	robot, 53
airpass, 6	rwalk, 55
ar1.2.s, 6	SP, 58
ar1.s, 7	spots, 60
ar2.s, 8	spots1, 61
arma11.s, 13	star, 61
beersales, 15	tbone, 67
bluebird, 15	tempdub, 67
bluebirdlite, 16	tuba, 72
boardings, 17	units, 73
co2, 18	usd.hkd, 73
color, 19	veilleux, 74
CREF, 19	wages, 75
cref.bond, 20	winnebago, 75
days, 20	* methods
deere1, 21	acf, 4
deere2, 21	arima, 8
deere3, 22	arima.boot,10
eeg, <u>26</u>	arimax, 11
electricity, 27	ARMAspec, 13
euph, <u>27</u>	armasubsets, 14
explode.s, 28	BoxCox.ar, 17
flow, 29	detectAO, 22
gold, 31	detectIO, 24
google, 32	eacf, 25
hare, 33	fitted.Arima, 28
hours, 34	garch.sim,29
ima22.s, <u>35</u>	gBox, 30
JJ, 35	harmonic, 33
larain, 39	Keenan.test, 36
ma1.1.s, 40	kurtosis, 37
ma1.2.s,41	lagplot, 38
ma2.s,41	LB.test, 39
milk, 43	McLeod.Li.test,42
oil.price,43	periodogram, 44
oilfilters,44	plot.Arima,45
prescrip, 49	plot.armasubsets,47
prey.eq, 51	predict.TAR,48

78 INDEX

prewhiten, 50	deere3, 22
qar.sim, 51	detectAO, 22
rstandard.Arima, 54	detectIO, 23, 24, 24
runs, 54	eacf, 25
season, 56	eeg, 26
skewness, 57	electricity, 27
spec, 58	euph, 27
summary.armasubsets, 62	explode.s, 28
tar, 62	CAP100C. 3, 20
tar.sim,64	fitted.Arima, 28
tar.skeleton, 65	flow, 29
tlrt, 68	,
Tsay.test, 69	garch.sim,29
tsdiag.Arimax,70	gBox, 30
tsdiag.TAR,71	gold, 31
zlag, 76	google, 32
* misc	
plot1.acf, 48	hare, 33
* package	harmonic, 33, 56
TSA-package, 3	hours, 34
acf, 4	ima22.s,35
airmiles, 5	
airpass, 6	JJ, 35
ar1.2.s, 6	Voonan tost 26 60
ar1.s, 7	Keenan. test, 36, 69
ar2.s,8	kurtosis, 37
arima, 8, 9, 12, 29	lagplot, 38
arima.boot, 10	larain, 39
arimax, 11	LB. test, 39
arma11.s, 13	25. 0000, 09
ARMAacf, 5	ma1.1.s, 40
ARMAspec, 13	ma1.2.s, 41
armasubsets, 14	ma2.s, 41
	McLeod.Li.test, 42
beersales, 15	milk, 43
bluebird, 15	
bluebirdlite, 16	oil.price,43
boardings, 17	oilfilters, 44
BoxCox.ar, 17	
	periodogram, 44
co2, 18	plot.acf, 5
color, 19	plot.Arima,45
CREF, 19	plot.armasubsets, 47
cref.bond, 20	plot1.acf, 48
	predict.TAR, 48, <i>64</i>
days, 20	prescrip, 49
deere1, 21	prewhiten, 50
deere2, 21	prey.eq, 51

INDEX 79

```
qar.sim, 51
retail, 52
robot, 53
rstandard.Arima, 54
runs, 54
rwalk, 55
season, 34, 56
skewness, 57
SP, 58
spec, 14, 58
spots, 60
spots1, 61
star, 61
summary.armasubsets, 62
tar, 49, 62, 65, 66, 72
tar.sim, 52, 64, 64
tar.skeleton, 64, 65
tbone, 67
tempdub, 67
tlrt, 36, 68, 70
TSA (TSA-package), 3
TSA-package, 3
Tsay.test, 36, 69, 69, 70
tsdiag.Arimax, 70
tsdiag.TAR, 64, 71
tuba, 72
units, 73
usd.hkd, 73
veilleux, 51, 74
wages, 75
winnebago, 75
zlag, 76
```