

Package ‘dse1’

July 11, 2007

Title Dynamic Systems Estimation (time series package)

Description Multivariate Time Series - ARMA and State Space models. See ?dse.Intro for more details.

Depends R (>= 2.5.0), tframe (>= 2007.5-3), setRNG (>= 2004.4-1)

Version 2007.7-1

Date 2007-07-11

LazyLoad yes

License Free. See the LICENCE file for details.

Author Paul Gilbert <pgilbert@bank-banque-canada.ca>

Maintainer Paul Gilbert <pgilbert@bank-banque-canada.ca>

URL <http://www.bank-banque-canada.ca/pgilbert>

R topics documented:

00.dse.Intro	3
acf	3
addPlotRoots	4
ARMA	5
balanceMittnik	6
bestTSEstModel	8
checkBalanceMittnik	9
checkBalance	10
checkConsistentDimensions	11
checkResiduals	12
coef.TSmodel	13
combine	14
combine.TSdata	14
DSEflags	15
dse-package	16
DSEutilities	18
DSEversion	18
eg1.DSE.data	19
egJofF.1dec93.data	20

estBlackBox1	21
estBlackBox2	22
estBlackBox3	23
estBlackBox4	24
estBlackBox	25
estMaxLik	26
estSSfromVARX	27
estSSMittnik	28
estVARXar	29
estVARXls	31
estWtVariables	32
fixConstants	33
fixF	34
gmap	34
informationTestsCalculations	35
informationTests	36
inputData	37
l.ARMA	38
l	39
l.SS	40
makeTSnoise	42
markovParms	43
McMillanDegree	44
MittnikReducedModels	45
MittnikReduction	45
nseriesInput	47
nstates	48
observability	48
percentChange.TSdata	49
periodsInput	50
periods.TSdata	52
plot.roots	52
Polynomials	53
Portmanteau	54
print.TSdata	54
print.TSestModel	55
reachability	56
residualStats	57
residuals.TSestModel	58
Riccati	58
roots	59
scale.TSdata	61
seriesNamesInput	62
seriesNames.TSdata	63
setArrays	64
setTSmodelParameters	64
simulate	65
smoother	67
SS	69
stability	71
state	72
summary.TSdata	73

sumSqerror	73
testEqual.ARMA	74
tfplot.TSdata	75
tframed.TSdata	76
toARMA	77
toSSChol	78
toSSinnov	79
toSSOform	79
toSS	80
TSdata.object	81
TSdata	82
TSestModel	83
TSmodel	84

Index**86**

00.dse.Intro

*Dynamic Systems Estimation - Multivariate Time Series Package***Description**

Functions for multivariate time series modeling

Details

See [dse-package](#) (in the help system use package?dse or ?"dse-package") for an overview.

acf*Calculate the acf for an object***Description**

Calculate the acf for an object.

Usage

```
## S3 method for class 'TSdata':
acf(x, ...)
## S3 method for class 'TSestModel':
acf(x, ...)
```

Arguments

x an object.
... additional arguments passed to stats::acf.

Details

The default method uses stats::acf. The method for TSdata extracts the output data and then uses acf. The method for TSestModel calculates the residuals (prediction minus data) and then uses acf.

Value

see stats::acf.

Author(s)

Paul Gilbert

See Also

`stats::acf`,

`addPlotRoots` *Add Model Roots to a plot*

Description

Calculate and plot roots of a model.

Usage

```
addPlotRoots(v, pch='*', fuzz=0)
```

Arguments

- | | |
|-------------------|---|
| <code>v</code> | An object containing a TSmodel. |
| <code>pch</code> | Character to use for plotting. |
| <code>fuzz</code> | If non-zero then roots within fuzz distance are considered equal. |

Value

The eigenvalues of the state transition matrix or the inverse of the roots of the determinant of the AR polynomial are returned invisibly.

Side Effects

The roots are added to an existing plot.

See Also

`plot.roots`

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
plot(roots(model))
addPlotRoots(toSS(model))
```

ARMA*ARMA Model Constructor*

Description

Constructs an ARMA TSmodel object as used by the DSE package.

Usage

```
ARMA(A=NULL, B=NULL, C=NULL, TREND=NULL,
      constants=NULL,
      description=NULL, names=NULL, input.names=NULL, output.names=NULL)
is.ARMA(obj)
```

Arguments

A	The auto-regressive polynomial, an axpxp array.
B	The moving-average polynomial, an bpxxp array.
C	The input polynomial, an cxpxm array. C should be NULL if there is no input
TREND	A matrix, p-vector, or NULL.
constants	NULL or a list of logical arrays with the same names as arrays above, indicating which elements should be considered constants.
description	An arbitrary string.
names	A list with elements input and output, each a vector of strings. Arguments input.names and output.names should not be used if argument names is used.
input.names	A vector of strings.
output.names	A vector of strings.
obj	Any object.

Details

The ARMA model is defined by:

$$A(L)y(t) = B(L)w(t) + C(L)u(t) + TREND(t)$$

where

A (axpxp) is the auto-regressive polynomial array.

B (bpxxp) is the moving-average polynomial array.

C (cxpxm) is the input polynomial array. C should be NULL if there is no input

y is the p dimensional output data.

u is the m dimensional control (input) data.

TREND is a matrix the same dimension as y, a p-vector (which gets replicated for each time period), or NULL.

This is sometime called a vector ARMA (VARMA) model, but the univariate case is also handled by this structure. VAR models are a special case where $B(L) = I$. ARIMA models are also special

cases where the polynomial arrays have unit roots, but these are not distinguished in a separate term as is sometimes done in other programs.

The name of last term, TREND, is misleading. If it is NULL it is treated as zero. If it is a p-vector, then this constant vector is added to the p-vector $y(t)$ at each period. For a stable model this would give the none zero mean, and might more appropriately be called the constant or intercept rather than trend. If the model is for differenced data, then this mean is the trend of the undifferenced model. The more general case is when TREND is a time series matrix of the same dimension as y . In this case it is added to y . This allows for a very general deterministic component, which may or may not be a traditional trend.

By default, elements in parameter arrays are treated as constants if they are exactly 1.0 or 0.0, and as parameters otherwise. A value of 1.001 would be treated as a parameter, and this is the easiest way to initialize an element which is not to be treated as a constant of value 1.0. Any array elements can be fixed to constants by specifying the list `constants`. Arrays which are not specified in the list will be treated in the default way. An alternative for fixing constants is the function `fixConstants`.

The function `ARMA` sets up a model but does not estimate it. See [estVARXls](#) for one possibility for estimating VAR models and [estMaxLik](#) for one possibility for estimating ARMA models.

Value

An ARMA TSmodel

See Also

`TSmodel`, [simulate.ARMA](#), `fixConstants`, [estVARXls](#), [estMaxLik](#)

Examples

```
mod1 <- ARMA(A=array(c(1,-.25,-.05), c(3,1,1)), B=array(1,c(1,1,1)))
AR   <- array(c(1, .5, .3, 0, .2, .1, 0, .2, .05, 1, .5, .3) ,c(3,2,2))
VAR  <- ARMA(A=AR, B=diag(1,2))
C    <- array(c(0.5,0,0,0.2),c(1,2,2))
VARX <- ARMA(A=AR, B=diag(1,2), C=C)
MA   <- array(c(1, .2, 0, .1, 0, 0, 1, .3), c(2,2,2))
ARMA <- ARMA(A=AR, B=MA, C=NULL)
ARMAX <- ARMA(A=AR, B=MA, C=C)
```

Description

Balance a state space model a la Mittnik.

Usage

```
balanceMittnik(model, n=NULL)
SVDbalanceMittnik(M, m, n=NULL)
```

Arguments

model	An TSmodel object.
M	a matrix. See details in MittnikReduction .
m	an integer indicating the number of input series in the model.
n	see details

Details

`balanceMittnik` calculate a state space model balance a la Mittnik. n is intended primarily for producing a state space model from the markov parameters of an ARMA model, but if it is supplied with an SS model the result will be a model with state dimension n based on the n largest singular values of the svd of a Hankel matrix of markov parameters generated by the original model. If n is not supplied then the singular values are printed and the program prompts for n. `balanceMittnik` calls `SVDbalanceMittnik`

`SVDbalanceMittnik` calculates a nested-balanced state space model by svd a la Mittnik. If state dim n is supplied then svd criteria are not calculated and the given n is used. Otherwise, the singular values are printed and the program prompts for n. M is a matrix with p x (m+p) blocks giving the markov parameters, that is, the first row of the Hankel matrix. It can be generated from the model as in the function `markovParms`, or from the data, as in the function `estSSMittnik`. m is the dimension of input series, which is needed to decompose M. The output dimension p is taken from `nrow(M)`.

See also [MittnikReduction](#) and references.

Value

A state space model in a `TSSestModel` object.

References

See references for [MittnikReduction](#).

See Also

[estVARXls](#), [estVARXar](#) [MittnikReduction](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(TSmodel(estVARXls(eg1.DSE.data.diff)))
## Not run: newmodel <-balanceMittnik(model)
# this prints information about singular values and prompts with
#Enter the number of singular values to use for balanced model:
# 18 might be a good choice in this example.
newmodel <-balanceMittnik(model, n=18)
```

bestTSEstModel *Select Best Model*

Description

Select the best model.

Usage

```
bestTSEstModel(models, sample.start=10, sample.end=NULL,
criterion='aic', verbose=TRUE)
```

Arguments

- models** a list of TSestModels.
- sample.start** the starting point to use for calculating information criteria.
- sample.end** the end point to use for calculating information criteria.
- criterion** Criterion to be used for model selection. see **informationTestsCalculations**.
'taic' would be a better default but this is not available for VAR and ARMA models.
- verbose** if TRUE then additional information is printed.

Details

Information criteria are calculated and return the best model from ... according to criterion models should be a list of TSestModel's. `models[[i]]$estimates$pred` is not recalculated but a sub-sample identified by `sample.start` and `sample.end` is used and the likelihood is recalculated. If `sample.end=NULL` data is used to the end of the sample. `taic` might be a better default selection criteria but it is not available for ARMA models.

Value

A TSestModel

See Also

`estBlackBox1`, `estBlackBox2` `estBlackBox3` `estBlackBox4` `informationTestsCalculations`

Examples

```
data("eg1.DSE.data.diff", package="dse1")
models <- list(estVARXls(eg1.DSE.data.diff), estVARXar(eg1.DSE.data.diff))
z <- bestTSEstModel(models)
```

```
checkBalanceMittnik
```

Check Balance of a TSmodel

Description

Calculate the difference between observability and reachability gramians of the model transformed to Mittnik's form.

Usage

```
checkBalanceMittnik(model)
## S3 method for class 'ARMA':
checkBalanceMittnik(model)
## S3 method for class 'SS':
checkBalanceMittnik(model)
## S3 method for class 'TSestModel':
checkBalanceMittnik(model)
```

Arguments

`model` An object of class `TSmodel`.

Details

Balanced models should have equal observability and reachability gramians.

Value

No value is returned.

Side Effects

Differences between the observability and reachability gramians are printed.

See Also

[checkBalance MittnikReduction](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
checkBalanceMittnik(model)
```

checkBalance *Check Balance of a TSmodel*

Description

Calculate the difference between observability and reachability gramians.

Usage

```
checkBalance(model)
## S3 method for class 'SS':
checkBalance(model)
## S3 method for class 'ARMA':
checkBalance(model)
## S3 method for class 'TSestModel':
checkBalance(model)
```

Arguments

model A TSmodel object.

Details

Balanced models should have equal observability and reachability gramians.

Value

No value is returned.

Side Effects

Differences between the observability and reachability gramians are printed.

See Also

[checkBalance](#) [Mittnik](#) [MittnikReduction](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
checkBalance(model)
```

```
checkConsistentDimensions  
Check Consistent Dimensions
```

Description

Check that dimensions of a model and data agree.

Usage

```
checkConsistentDimensions(obj1, obj2=NULL)  
## Default S3 method:  
checkConsistentDimensions(obj1, obj2=NULL)  
## S3 method for class 'ARMA':  
checkConsistentDimensions(obj1, obj2=NULL)  
## S3 method for class 'SS':  
checkConsistentDimensions(obj1, obj2=NULL)  
## S3 method for class 'TSdata':  
checkConsistentDimensions(obj1, obj2=NULL)  
## S3 method for class 'TSEstModel':  
checkConsistentDimensions(obj1, obj2=NULL)
```

Arguments

`obj1` An object containing a TSmodel, TSdata, or TSEstModel, depending on the method

`obj2` Another object containing TSdata corresponding to the TSmodel in `obj1`, or a TSmodel corresponding to the TSdata in `obj1`.

Details

Check that dimensions of a model and data agree. If `obj1` is a TSEstModel then if `obj2` is NULL, TSdata is taken from `obj1`.

Value

`logical`

Examples

```
data("eg1.DSE.data.diff", package="dse1")  
model <- estVARXls(eg1.DSE.data.diff)  
checkConsistentDimensions(model)
```

checkResiduals Autocorrelations Diagnostics

Description

Calculate autocorrelation diagnostics of a time series matrix or TSdata or residuals of a TSestModel

Usage

```
checkResiduals(obj, ...)
## Default S3 method:
checkResiduals(obj, ac=TRUE, pac=TRUE, select=seq(nseries(obj)),
               drop=NULL, plot.=TRUE, graphs.per.page=5, verbose=FALSE, ...)
## S3 method for class 'TSdata':
checkResiduals(obj, ...)
## S3 method for class 'TSestModel':
checkResiduals(obj, ...)
```

Arguments

obj	An TSestModel or TSdata object.
ac	If TRUE the auto-correlation function is plotted.
pac	If TRUE the partial auto-correlation function is plotted.
select	Is used to indicate a subset of the residual series. By default all residuals are used.
drop	Is used to indicate a subset of the residual time periods to drop. All residuals are used with the default (NULL). Typically this can be used to get rid of bad initial conditions (eg. drop=seq(10)) or outliers.
plot.	If FALSE then plots are not produced.
graphs.per.page	Integer indicating number of graphs to place on a page.
verbose	If TRUE then the auto-correlations and partial auto-correlations are printed if they are calculated.
...	arguments passed to other methods.

Details

This is a generic function. The default method works for a time series matrix which is treated as if it were a matrix of residuals. However, in a Box-Jenkins type of analysis the matrix may be data which is being evaluated to determine a model. The method for a TSestModel evaluates the residuals calculated by subtracting the output data from the model predictions.

Value

A list with residual diagnostic information: residuals, mean, cov, acf= autocorrelations, pacf= partial autocorrelations.

Side Effects

Diagnostic information is printed and plotted if a device is available. Output graphics can be paused between pages by setting par(ask=TRUE).

See Also

[informationTests](#), [Portmanteau](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
checkResiduals(model)
```

coef.TSmodel

Extract or set Model Parameters

Description

Set or extract coefficients (parameter values) of model objects.

Usage

```
## S3 method for class 'TSmodel':
coef(object, ...)
## S3 method for class 'TSEstModel':
coef(object, ...)
coef(object) <- value
## S3 replacement method for class 'default':
coef(object) <- value
```

Arguments

- object An object of class TSmodel or TSEstModel.
- value value to be assigned to object.
- ... (further arguments, currently disregarded).

Value

A vector of parameter values.

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
coef(model)
coef(model) <- 0.1 + coef(model)
```

combine	<i>Combine two objects.</i>
---------	-----------------------------

Description

This is a generic method to combine two objects of the same class to make a single object of that class.

Usage

```
combine(e1, e2)
## Default S3 method:
combine(e1, e2)
```

Arguments

e1, e2 TSdata objects.

Value

An object of the same class as the argument but containing both e1 and e2.

See Also

tbinding, combine.TSdata, combine.forecastCov

Examples

```
data("eg1.DSE.data.diff", package="dse1")
data("eg1.DSE.data", package="dse1")
new.data.set <- combine(eg1.DSE.data.diff, eg1.DSE.data)
```

combine.TSdata	<i>Combine series from two TSdata objects.</i>
----------------	--

Description

Combine series from two TSdata objects.

Usage

```
## S3 method for class 'TSdata':
combine(e1, e2)
```

Arguments

e1, e2 TSdata objects.

Value

An object of class TSdata which includes series from both e1 and e2.

See Also

tbind

Examples

```
data("egl.DSE.data.diff", package="dse1")
data("egl.DSE.data", package="dse1")
new.data.set <- combine(egl.DSE.data.diff, egl.DSE.data)
```

DSEflags

Flags to Indicate Use of Compiled Code

Description

Determines if compiled code should be used or not.

Usage

```
.DSEflags(new)
```

Arguments

new A list which must have elements COMPILED and DUP.

Details

Setting flags with this function is primarily for debugging. It should not normally be needed by users. If called with no arguments, `.DSEflags()` returns the current setting. Several **dse** functions which call compiled fortran or C code will use the equivalent S/R version if the `.DSEflags()$COMPILED` returns FALSE. `.DSEflags()$DUP` is passed to fortran calls and controls whether R duplicates arguments passed to the fortran code. The safe setting is TRUE. Setting FALSE saves some memory but does not seem to give much speed gain.

Side Effects

The flag settings affect whether and how compiled fortran or C code is called.

Examples

```
.DSEflags(list(COMPILED=TRUE, DUP=TRUE))
.DSEflags()$COMPILED
```

Description

Functions for time series modeling, including multi-variate state-space and ARMA (VAR, ARIMA, ARIMAX) models.

Details

A *Brief User's Guide* is distributed with the **dse** bundle in the directory ‘dse/dse1/inst/doc/dse-guide.pdf’. The package implements an R/S style object oriented approach to time series modeling. This means that different model and data representations can be implemented with fairly simple extensions to the package.

The package includes methods for simulating, estimating, and converting among different model representations. These are mainly in **dse1**. Package **dse2** has methods for studying estimation techniques and for examining the forecasting properties of models. There are also functions for forecasting and for evaluating the performance of forecasting models, as well as functions for evaluating model estimation techniques.

Bundle:	dse
Contains:	tframe dse1 dse2
Depends:	R, setRNG, tframe
License:	free, see LICENSE file for details.
URL:	http://www.bank-banque-canada.ca/pgilbert

The main objects are:

TSdata time series input and output data structure

TSmodel a DSE model structure

TSestModel model, data and some estimation information

The main general methods are:

TSdata create, extract a DSE data structure

TSmodel create, extract a DSE model structure

simulate simulate a model to produce artifical data

toSS convert to a state-space model

toARMA convert to an ARMA model

ARMA construct an ARMA model

SS construct a state-space model

I evaluate a model with data

smoother calculate the smoothed state estimate

The main estimation methods are:

estVARXls estimate an ARMA model with least squares

estVARXar estimate an ARMA model with ar

estSSfromVARX calculate a state-space model from an estimated VAR model

bft a (usually) good “black-box” estimated model

estMaxLik estimate a model using maximum likelihood

The main diagnostic methods are:

checkResiduals autocorrelation diagnostics

informationTests calculate several information tests for a model

McMillanDegree calculate the McMillanDegree of a model

stability calculate the stability of a model

roots calculate the roots of a model

The methods for producing and evaluating forecasts are:

l evaluate a model with data (and simple forecasts)

forecast calculate forecasts

featherForecasts calculate forecasts starting at different periods

horizonForecasts calculate forecasts at different horizons

forecastCov calculate the covariance of forecasts

MonteCarloSimulations multiple simulations

The methods for evaluating estimation methods are:

EstEval evaluate estimation methods

The functions described in the *Brief User’s Guide* and examples in the help pages should work fairly reliably (since they are tested regularly), however, the code is distributed on an “as-is” basis. This is a compromise which allows me to make the software available with minimum effort. This software is not a commercial product. It is the by-product of ongoing research. Error reports, constructive suggestions, and comments are welcomed.

Usage

```
library("dse1")
library("dse2")
```

Author(s)

Paul Gilbert <pgilbert@bank-banque-canada.ca>

Maintainer: Paul Gilbert <pgilbert@bank-banque-canada.ca>

References

Anderson, B. D. O. and Moore, J. B. (1979) *Optimal Filtering*. Prentice-Hall.

Gilbert, P. D. (1993) State space and ARMA models: An overview of the equivalence. Working paper 93-4, Bank of Canada. Available at www.bank-banque-canada.ca/pgilbert

Gilbert, P. D. (1995) Combining VAR Estimation and State Space Model Reduction for Simple Good Predictions. *J. of Forecasting: Special Issue on VAR Modelling*. **14**:229–250.

Gilbert, P.D. (2000) A note on the computation of time series model roots. *Applied Economics Letters*, **7**, 423–424

Jazwinski, A. H. (1970) *Stochastic Processes and Filtering Theory*. Academic Press.

See Also

[TSdata](#), [TSmodel](#), [TSEstModel.object](#)

DSEutilities

DSE Utilities

Description

These functions are used by other functions in DSE and are not typically called directly by the user.

Usage

```
estVARXmean.correction(x, y, bbar, fuzz=sqrt(.Machine$double.eps), warn=TRUE)
fake.TSEstModel.missing.data(model,data, residual, max.lag, warn=TRUE)
printTestValue(x, digits=16)
svd.criteria(sv)
criteria.table.heading()
criteria.table.legend()
criteria.table.nheading()
DSE.ar(data, ...)
```

DSEversion

Print Version Information

Description

Print version information.

Usage

```
DSEversion()
```

Examples

```
DSEversion()
```

eg1.DSE.data *Four Time Series used in Gilbert (1993)*

Description

Data is for Canada. The series start in March 1961 (April 1961 for eg1.DSE.data.diff) and end in June 1991, giving 364 observations on each variable (363 for eg1.DSE.data.diff).

The input series is 90-day interest rates (R90) in both eg1.DSE.data and eg1.DSE.data.diff.

The output series are M1, GDP lagged two months, and CPI. M1, GDP and CPI were all seasonally adjusted data. These are not transformed in eg1.DSE.data and are first difference of logs in eg1.DSE.data.diff.

GDP is lagged because it is not available on as timely a basis. (The data was used in an example where the intent was to build a model for timely monitoring.)

The Statistics Canada series identifiers are B14017, B1627, I37026, and B820200.

The data for M1 (B1627) were taken prior to revisions made in December 1993.

The file eg1.dat contains the same data as eg1.DSE.data in a simple ASCII file.

Usage

```
data(eg1.DSE.data)
data(eg1.DSE.data.diff)
```

Format

The objects eg1.DSE.data and eg1.DSE.data.diff are TSdata objects. The file eg1.dat is an ASCII file with 5 columns, the first enumerating the observations, the second giving the input series, and the third to fifth giving the output series. The input series name is "R90" and the output series names are "M1", "GDPI2" and "CPI". GDPI2 is GDP lagged two months

Source

Statistics Canada, Bank of Canada.

References

Gilbert, P.D. (1993) State Space and ARMA Models: An Overview of the Equivalence. Bank of Canada Working Paper 93-4. Also available at www.bank-banque-canada.ca/pgilbert.

See Also

[TSdata](#)

`egJoff.1dec93.data` *Eleven Time Series used in Gilbert (1995)*

Description

Data is for Canada unless otherwise indicated. The series start in February 1974 and end in September 1993 (236 observations on each variable).

The input series is 90 day interest rates (R90) and the ten output variables are CPI, GDP, M1, long run interest rates (RL), the Toronto stock exchange 300 index (TSE300), employment, the Canada/US exchange rate (PFX), a commodity price index in US dollars, US industrial production, and US CPI.

R90, RL and TSE are differenced. All other variables are in terms of percent change.

R90 is the 3 month prime corporate paper rate. While it is not set directly by the Bank of Canada, Bank policy influences it directly and it is often thought of as a proxy "policy variable."

The Statistics Canada identifiers are B14017 (R90), P484549 (CPI), I37026 (GDP), B1627 (M1), B14013 (RL), B4237 (TSE300), D767608 (employment), B3400 (PFX).

M.BCPI (commodity price index) is published by the Bank of Canada. JQIND (US industrial production), and CUSA0 (US CPI) are DRI identifiers.

The data for M1 (B1627) were taken prior to revisions made in December 1993.

Usage

```
data(egJoff.1dec93.data)
```

Format

This data is a TSdata object. The input series name is "R90" and the output series names are "CPI", "GDP", "M1", "RL", "TSE300", "employment", "PFX", "commod.price index", "US ind.prod." and "US CPI"

Source

Statistics Canada, Bank of Canada, DRI.

References

Gilbert, P.D. 1995 "Combining VAR Estimation and State Space Model Reduction for Simple Good Predictions" *J. of Forecasting: Special Issue on VAR Modelling.* 14:229-250

See Also

[TSdata](#)

<code>estBlackBox1</code>	<i>Estimate a TSmodel</i>
---------------------------	---------------------------

Description

Estimate a TSmodel.

Usage

```
estBlackBox1(data, estimation="estVARXls",
            reduction="MittnikReduction",
            criterion="taic", trend=FALSE, subtract.means=FALSE,
            verbose=TRUE, max.lag=6)
```

Arguments

<code>data</code>	Data in an object of class TSdata.
<code>estimation</code>	Initial estimation method to be used.
<code>reduction</code>	Reduction method to be used.
<code>criterion</code>	Criterion to be used for model selection. see informationTestsCalculations .
<code>trend</code>	logical indicating if a trend should be estimated.
<code>subtract.means</code>	logical indicating if the mean should be subtracted from data before estimation.
<code>verbose</code>	logical indicating if information should be printed during estimation.
<code>max.lag</code>	integer indicating the maximum number of lags to consider.

Value

A state space model in an object of class TSestModel.

Side Effects

If `verbose` is TRUE then estimation information is printed and `checkResiduals` is run, which gives plots of information about the residuals.

See Also

[informationTestsCalculations](#)

Examples

```
data("egJoffF.1dec93.data", package="dse1")
goodmodel <- estBlackBox1(egJoffF.1dec93.data)
```

<code>estBlackBox2</code>	<i>Estimate a TSmodel</i>
---------------------------	---------------------------

Description

Estimate a TSmodel.

Usage

```
estBlackBox2(data, estimation='estVARXls',
             lag.weight=.9,
             reduction='MittnikReduction',
             criterion='taic',
             trend=FALSE,
             subtract.means=FALSE, re.add.means=TRUE,
             standardize=FALSE, verbose=TRUE, max.lag=12)
```

Arguments

<code>data</code>	a TSdata object.
<code>estimation</code>	a character string indicating the estimation method to use.
<code>lag.weight</code>	weighting to apply to lagged observations.
<code>reduction</code>	character string indicating reduction procedure to use.
<code>criterion</code>	criterion to be used for model selection. see <code>informationTestsCalculations</code> .
<code>trend</code>	if TRUE include a trend in the model.
<code>subtract.means</code>	if TRUE the mean is subtracted from the data before estimation.
<code>re.add.means</code>	if subtract.means is TRUE then if re.add.means is TRUE the estimated model is converted back to a model for data without the mean subtracted.
<code>standardize</code>	if TRUE the data is transformed so that all variables have the same variance.
<code>verbose</code>	if TRUE then additional information from the estimation and reduction procedures is printed.
<code>max.lag</code>	The number of lags to include in the VAR estimation.

Details

A model is estimated and then a reduction procedure applied. The default estimation procedure is least squares estimation of a VAR model with lagged values weighted. This procedure is discussed in Gilbert (1995).

Value

A TSestModel.

References

Gilbert, P.D. (1995) Combining VAR Estimation and State Space Model Reduction for Simple Good Predictions *J. of Forecasting: Special Issue on VAR Modelling*, **14**, 229–250.

See Also

[estBlackBox1](#), [estBlackBox3](#) [estBlackBox4](#) [informationTestsCalculations](#)

Examples

```
data("egl.DSE.data.diff", package="dse1")
z <- estBlackBox2(egl.DSE.data.diff)
```

estBlackBox3

Estimate a TSmodel

Description

Estimate a TSmodel.

Usage

```
estBlackBox3(data, estimation='estVARXls',
             lag.weight=1.0,
             reduction='MittnikReduction',
             criterion='aic',
             trend=FALSE,
             subtract.means=FALSE, re.add.means=TRUE,
             standardize=FALSE, verbose=TRUE, max.lag=12, sample.start=10)
```

Arguments

data	A TSdata object.
estimation	A character string indicating the estimation method to use.
lag.weight	Weighting to apply to lagged observations.
reduction	Character string indicating reduction procedure to use.
criterion	Criterion to be used for model selection. see informationTestsCalculations . taic might be a better default selection criteria but it is not available for ARMA models.
trend	If TRUE include a trend in the model.
subtract.means	If TRUE the mean is subtracted from the data before estimation.
re.add.means	If subtract.means is TRUE then if re.add.means is T the estimated model is converted back to a model for data without the mean subtracted.
standardize	If TRUE the data is transformed so that all variables have the same variance.
verbose	If TRUE then additional information from the estimation and reduction procedures is printed.
max.lag	The number of lags to include in the VAR estimation.
sample.start	The starting point to use for calculating information criteria.

Details

VAR models are estimated for each lag up to the specified max.lag. From these the best is selected according to the specified criteria. The reduction procedure is then applied to this best model and the best reduced model selected. The default estimation procedure is least squares estimation of a VAR model.

Value

A TSestModel.

See Also

`estBlackBox1`, `estBlackBox2` `estBlackBox4` `informationTestsCalculations`

Examples

```
data("eg1.DSE.data.diff", package="dse1")
z <- estBlackBox3(eg1.DSE.data.diff)
```

`estBlackBox4` *Estimate a TSmodel*

Description

Estimate a TSmodel with Brute Force Technique.

Usage

```
estBlackBox4(data, estimation="estVARXls",
             lag.weight=1.0, variable.weights=1,
             reduction="MittnikReduction",
             criterion="taic",
             trend=FALSE, subtract.means=FALSE, re.add.means=TRUE,
             standardize=FALSE, verbose=TRUE, max.lag=12, sample.start=10,
             waft=10, bft(data, ... )
```

Arguments

<code>data</code>	A TSdata object.
<code>estimation</code>	a character string indicating the estimation method to use.
<code>lag.weight</code>	weighting to apply to lagged observations.
<code>variable.weights</code>	weighting to apply to series if estimation method is <code>estWtVariables</code> .
<code>reduction</code>	character string indicating reduction procedure to use.
<code>criterion</code>	criterion to be used for model selection. see <code>informationTestsCalculations</code> .
<code>trend</code>	if TRUE include a trend in the model.
<code>subtract.means</code>	if TRUE the mean is subtracted from the data before estimation.
<code>re.add.means</code>	if <code>subtract.means</code> is TRUE then if <code>re.add.means</code> is T the estimated model is converted back to a model for data without the mean subtracted.

standardize if TRUE the data is transformed so that all variables have the same variance.
verbose if TRUE then additional information from the estimation and reduction procedures is printed.
max.lag VAR estimation is done for each lag up to max.lag.
sample.start the starting point to use for calculating information criteria in the final selection.
warn logical indicating if warning messages should be suppressed.
... arguments passed to estBlackBox4.

Details

For each lag up to max.lag a VAR model is estimated and then a reduction procedure applied to select the best reduced model. Finally the best of the best reduced models is selected. The default estimation procedure is least squares estimation of the VAR models. This procedure is described as the brute force technique (bft) in *Gilbert (1995)*.

Value

A TSestModel.

References

Gilbert, P.D. (1995) Combining VAR Estimation and State Space Model Reduction for Simple Good Predictions *J. of Forecasting: Special Issue on VAR Modelling*, **14**, 229–250.

See Also

[estBlackBox1](#), [estBlackBox2](#) [estBlackBox3](#) [informationTestsCalculations](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
z <- bft(eg1.DSE.data.diff)
```

estBlackBox *Estimate a TSmodel*

Description

Estimate a TSmodel.

Usage

```
estBlackBox(data, ...)
```

Arguments

data Data in an object of class TSdata.
... Optional arguments depend on the function which is eventually called.

Details

The function makes a call the most promising procedure currently available. These tend to have names like estBlackBox1, estBlackBox2, estBlackBox4. This is an active area of ongoing research and so the actual routine called will probably change with new versions.

Value

A state space model in an object of class TSestModel.

Examples

```
data("egJoff.1dec93.data", package="dse1")
goodmodel <- estBlackBox(egJoff.1dec93.data)
```

estMaxLik

Maximum Likelihood Estimation

Description

Maximum likelihood estimation.

Usage

```
estMaxLik(obj1, obj2=NULL, ...)
## S3 method for class 'TSmodel':
estMaxLik(obj1, obj2, algorithm="optim",
           algorithm.args=list(method="BFGS", upper=Inf, lower=-Inf, hessian=TRUE),
           ...)
## S3 method for class 'TSestModel':
estMaxLik(obj1, obj2=TSdata(obj1), ...)
## S3 method for class 'TSdata':
estMaxLik(obj1, obj2, ...)
```

Arguments

- obj1 an object of class TSmodel, TSdata or TSestModel
- obj2 TSdata or a TSmodel to be fitted with obj1.
- algorithm the algorithm ('optim', or 'nlm') to use for maximization.
- algorithm.args arguments for the optimization algorithm.
- ... arguments passed on to other methods.

Details

One of obj1 or obj2 should specify a TSmodel and the other TSdata. If obj1 is a TSestModel and obj2 is NULL, then the data is extracted from obj1. The TSmodel object is used to specify both the initial parameter values and the model structure (the placement of the parameters in the various arrays of the TSmodel). Estimation attempts to minimize the negative log likelihood (as returned by 1) of the given model structure by adjusting the parameter values. A TSmodel can also have constant values in some array elements, and these are not changed. (See SS, ARMA and fixConstants regarding setting of constants.)

With the number of parameter typically used in multivariate time series models, the default maximum number of iterations may not be enough. Be sure to check for convergence (a warning is printed at the end, or use `summary` on the result). The maximum iterations is passed to the estimation algorithm with `algorithm.args`, but the elements of that list will depend on the specified optimization algorithm (so see the help for the algorithm). The example below is for the default `optim` algorithm.

Value

The value returned is an object of class `TSestModel` with additional elements `est$converged`, which is TRUE or FALSE indicating convergence, `est$convergenceCode`, which is the code returned by the estimation algorithm, and `est$results`, which are detailed results returned by the estimation algorithm. The hessian and gradient in results could potentially be used for restarting in the case of non-convergence, but that has not yet been implemented.

Warning

Maximum likelihood estimation of multivariate time series models tends to be problematic, even when a good structure and good starting parameter values are known. This is especially true for state space models. Also, it seems that in-sample fit is often obtained at the expense of out-of-sample forecasting ability. If a prior model structure is not important then the `bft` estimation method may be preferable.

See Also

`optim`, `nlm`, `estVARXls`, `bft`, `TSmodel`, `l`, `SS`, `ARMA`, `fixConstants`

Examples

```
true.model <- ARMA(A=c(1, 0.5), B=1)
est.model <- estMaxLik(true.model, simulate=true.model))
summary(est.model)
est.model
tfplot(est.model)
est.model <- estMaxLik(true.model, simulate=true.model),
algorithm.args=list(method="BFGS", upper=Inf, lower=-Inf, hessian=TRUE,
control=list(maxit=10000)))
```

Description

Estimate a VAR `TSmodel` with (optionally) an exogenous input and convert to state space.

Usage

```
estSSfromVARX(data, warn=TRUE, ...)
```

Arguments

- `data` An object with the structure of an object of class TSdata (see TSdata).
`warn` Logical indicating if warnings should be printed (TRUE) or suppressed (FALSE).
`...` See arguments to estVARXls

Details

This function uses the functions estVARXls and toSS.

Value

A state space model in an object of class TSestModel.

References

- Gilbert, P. D. (1993) State space and ARMA models: An overview of the equivalence. Working paper 93-4, Bank of Canada. Available at www.bank-banque-canada.ca/pgilbert.
 Gilbert, P. D. (1995) Combining VAR Estimation and State Space Model Reduction for Simple Good Predictions. *J. of Forecasting: Special Issue on VAR Modelling*. 14:229-250.

See Also

`toSS` `estSSMittnik` `bft` `estVARXls` `estMaxLik`

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estSSfromVARX(eg1.DSE.data.diff)
```

`estSSMittnik`

Estimate a State Space Model

Description

Estimate a state space model using Mittnik's markov parameter estimation.

Usage

```
estSSMittnik(data, max.lag=6, n=NULL, subtract.means=FALSE, normalize=FALSE)
```

Arguments

- `data` A TSdata object.
`max.lag` The number of markov parameters to estimate.
`n` The state dimension.
`subtract.means` If TRUE subtract the means from the data before estimation.
`normalize` If TRUE normalize the data before estimation.

Details

Estimate a nested-balanced state space model by svd from least squares estimate of markov parameters a la *Mitnik* (1989, p1195). The quality of the estimate seems to be quite sensitive to `max.lag`, and this is not properly resolved yet. If `n` is not supplied the svd criteria will be printed and `n` prompted for. If `subtract.means=T` then the sample mean is subtracted. If `normalize` is `T` the lsfit estimation is done with outputs normalize to `cov=I` (There still seems to be something wrong here!!). The model is then re-transformed to the original scale.

See [MitnikReduction](#) and references cited there. If the state dimension is not specified then the singular values of the Hankel matrix are printed and the user is prompted for the state dimension.

Value

A state space model in an object of class `TSestModel`.

References

See references for [MitnikReduction](#).

See Also

[MitnikReduction](#) [estVARXls](#) [bft](#)

Examples

```
data("egJoff.1dec93.data", package="dse1")
## Not run: model <- estSSMitnik(egJoff.1dec93.data)
# this prints information about singular values and prompts with
#Enter the number of singular values to use for balanced model:
# the choice is difficult in this example.
model <- estSSMitnik(egJoff.1dec93.data, n=3)
```

Description

Estimate a VAR TSmodel with (optionally) an exogenous input.

Usage

```
estVARXar(data, subtract.means=FALSE, re.add.means=TRUE, standardize=FALSE,
unstandardize=TRUE, aic=TRUE, max.lag=NULL, method="yule-walker", warn=
```

Arguments

`data` A `TSdata` object.

`subtract.means` If TRUE subtract the means from the data before estimation.

`re.add.means` If TRUE the model is adjusted for the non-zero mean data when returned. If `subtract.means` is also TRUE then the mean is added back to the data.

standardize	Note that the mean is not subtracted unless subtract.means is TRUE. A VAR model in an object of class TSestModel.
unstandardize	If TRUE and standardize is TRUE then the returned model is adjusted to correspond to the original data.
aic	Passed to function ar.
max.lag	The maximum number of lags that should be considered.
method	Passed to function ar.
warn	If TRUE certain warning message are suppressed.

Details

This function estimates a VAR model with exogenous variable using ar(). Residuals,etc, are calculated by evaluating the estimated model with ARMA. The procedure ar is used by combine exogenous variables and endogenous variable and estimating as if all variables were endogenous. The estVARXar method does not support trend estimation (as in estVARXls).

If aic=TRUE the number of lags is determined by an AIC statistic (see ar). If an exogenous (input) variable is supplied the input and output are combined (i.e.- both treated as outputs) for estimation, and the resulting model is converted back by transposing the exogenous variable part of the polynomial and discarding inappropriate blocks. Residuals,etc, are calculated by evaluating the estimated model as a TSmodel/ARMA with the data (ie. residuals are not the residuals from the regression).

Note: ar uses a Yule-Walker approach (uses autocorrelations) so effectively the model is for data with means removed. Thus subtract.means does not make much difference and re.add.means must be TRUE to get back to a model for the original data.

The convention for AR(0) and sign are changed to ARMA format. Data should be of class TSdata. The exog. variable is shifted so contemporaneous effects enter. the model for the exog. variable (as estimated by ar()) is discarded.

Value

A TSestModel object containing an ARMA TSmodel object. The model has no MA portion so it is a VAR model.

References

- Gilbert, P. D. (1993) State space and ARMA models: An overview of the equivalence. Working paper 93-4, Bank of Canada. Available at www.bank-banque-canada.ca/pgilbert
- Gilbert, P. D. (1995) Combining VAR Estimation and State Space Model Reduction for Simple Good Predictions. *J. of Forecasting: Special Issue on VAR Modelling*. **14**:229–250.

See Also

`estSSfromVARX` `estSSMittnikBFT` `estVARXls` `estMaxLik` `ar` `DSE.ar`

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXar(eg1.DSE.data.diff)
```

<code>estVARXls</code>	<i>Estimate a VAR TSmodel</i>
------------------------	-------------------------------

Description

Estimate a VAR TSmodel with (optionally) an exogenous input and (optionally) a trend.

Usage

```
estVARXls(data, subtract.means=FALSE, re.add.means=TRUE, standardize=FALSE,
           unstandardize=TRUE, max.lag=NULL, trend=FALSE, lag.weight=1.0, warn=TRUE)
```

Arguments

- `data` A TSdata object.
- `subtract.means` If TRUE subtract the means from the data before estimation.
- `re.add.means` If TRUE and subtract.means is TRUE then the mean is added back to the data and the model is adjusted for the non-zero mean data when returned.
- `standardize` If TRUE divide each series by its sample standard deviation before estimation. Note that the mean is not subtracted unless subtract.means is TRUE.
- `unstandardize` If TRUE and standardize is TRUE then the returned model is adjusted to correspond to the original data.
- `trend` If TRUE a trend is estimated.
- `max.lag` Number of lags to be used.
- `lag.weight` Weight between 0 and 1 to be applied to lagged data. Lower weights mean lagged data is less important (more noisy).
- `warn` If TRUE a warning message is issued when missing data (NA) is detected and the model predictions are reconstructed from the lsfit residuals.

Details

A VAR model is fitted by least squares regression using `lsfit`. The argument `max.lag` determines the number of lags. If it is not specified then six lags are used. This is an exceedingly naive approach, so the `max.lag` argument really should be specified (or see `bft` for a more complete approach to model selection.) If a trend is not estimated the function `estVARXar` may be preferred. Missing data is allowed in `lsfit`, but not (yet) by `ARMA` which generates the model predictions, etc., based on the estimated model and the data. (This is done to ensure the result is consistent with other estimation techniques.) In the case of missing data `ARMA` is not used and the model predictions, etc., are generated by adding the data and the `lsfit` residual. This is slightly different from using `ARMA`, especially with respect to initial conditions.

Value

A `TStestModel` object containing a `TSmodel` object which is a VAR model.

References

- Gilbert, P. D. (1993) State space and ARMA models: An overview of the equivalence. Working paper 93-4, Bank of Canada. Available at www.bank-banque-canada.ca/pgilbert
- Gilbert, P. D. (1995) Combining VAR Estimation and State Space Model Reduction for Simple Good Predictions. *J. of Forecasting: Special Issue on VAR Modelling.* **14**:229–250.

See Also

[estSSfromVARX](#) [estSSMittnik](#) [bft](#) [estVARXar](#) [estMaxLik](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
```

estWtVariables *Weighted Estimation*

Description

estWtVariables

Usage

```
estWtVariables(data, variable.weights,
                estimation="estVARXls", estimation.args=NULL)
```

Arguments

data	A TSdata object.
variable.weights	weights to use for each output series.
estimation	An estimation method.
estimation.args	An arguments for the estimation method.

Details

Weight series so that some series residuals are more important than others. Each output variable is scaled according to **variable.weights**, estimate is done, and then the estimated model unscaled. Estimation is done the method specified by **estimate** and any arguments specified by **estimation.args**. **estimation.args** should be **NULL** if no args are needed.

Value

A TSestModel.

See Also

[estVARXls](#) [estBlackBox](#) [bft](#) [estMaxLik](#)

fixConstants*Fix TSmodel Coefficients (Parameters) to Constants*

Description

Fix specified coefficients to constant values or any coefficients within fuzz of 0.0 or 1.0 to exactly 0.0 or 1.0. This will not change the model much but will affect some estimation techniques and information criteria results, as these are considered to be constants rather than coefficients.

Usage

```
fixConstants(model, fuzz=1e-5, constants=NULL)
```

Arguments

- | | |
|------------------------|--|
| <code>model</code> | an object of class TSmodel. |
| <code>fuzz</code> | absolute difference to be considered equivalent. |
| <code>constants</code> | NULL or a list of logical arrays. |

Details

If constants is not NULL then parameters within fuzz of 0.0 or 1.0 are set as constants 0.0 or 1.0. If constants is not NULL then it should be a list with logical arrays named F, G ..., with TRUE corresponding to any array elements which are to be treated as constant.

Value

An object of class 'SS' 'TSmodel' with some array entries set to constants 0.0 or 1.0.

See Also

[fixF](#)

Examples

```
f <- array(c(.5,.3,.2,.4),c(2,2))
h <- array(c(1,0,0,1),c(2,2))
k <- array(c(.5,.3,.2,.4),c(2,2))
ss <- SS(F=f,G=NULL,H=h,K=k)
ss
coef(ss)
ss <- fixConstants(ss, constants=list(
    F = matrix(c(TRUE, FALSE, FALSE, FALSE), 2,2)))
ss
coef(ss)
data("eg1.DSE.data.diff", package="dse1")
model <- toARMA(toSS(estVARXls(eg1.DSE.data.diff)))
model <- fixConstants(model)
```

fixF*Set SS Model F Matrix to Constants***Description**

Set any parameters of the F matrix to constants. The same values are retained but they are considered to be constants rather than parameters. This will not change the model but will affect some estimation techniques and information criteria results.

Usage

```
fixF(model)
```

Arguments

<code>model</code>	An object of class TSmodel.
--------------------	-----------------------------

Value

An SS TSmodel object.

See Also

[fixConstants](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
model <- fixF(model)
```

gmap*Basis Transformation of a Model.***Description**

Transform the basis for the state by a given invertible matrix.

Usage

```
gmap(g, model)
```

Arguments

<code>g</code>	An invertible matrix
<code>model</code>	An object of class TSmodel.

Details

If the input model is in state space form g is a change of basis for the state. If the input model is in ARMA form then the polynomials are premultiplied by g. If g is a scalar it is treated as a diagonal matrix.

Value

An equivalent model transformed using g.

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
gmap(2, model)
```

informationTestsCalculations
Calculate selection criteria

Description

Calculates several model selection criteria.

Usage

```
informationTestsCalculations(lst, sample.start=1, sample.end=NULL, warn=TRUE)
```

Arguments

- lst One or more objects of class TSestModel.
- sample.start The start of the period to use for criteria calculations.
- sample.end The end of the period to use for criteria calculations. If omitted the end of the sample is used.
- warn If FALSE then some warning messages are suppressed.

Value

The calculated values are returned in a vector with names: port, like, aic, bic, gvc, rice, fpe, taic, tbic, tgvc, trice and tfpe. These correspond to values for the Portmanteau test, likelihood, Akaike Information Criterion, Bayes Information Criterion, Generalized Cross Validation, Rice Criterion, and Final Prediction Error. The preceding 't' indicates that the theoretical parameter space dimension has been used, rather than the number of coefficient (parameter) values. Methods which select a model based on some information criterion calculated by informationTestsCalculations should use the name of the vector element to specify the test value which is to be used.

See Also

[informationTests](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
informationTestsCalculations(model)
```

informationTests *Tabulates selection criteria*

Description

Tabulates several model selection criteria.

Usage

```
informationTests(..., sample.start=1, sample.end=NULL, Print=TRUE, warn=TRUE)
```

Arguments

- ... At least one object of class TSestModel.
- sample.start The start of the period to use for criteria calculations.
- sample.end The end of the period to use for criteria calculations. If omitted the end of the sample is used.
- Print If FALSE then printing suppressed.
- warn If FALSE then some warning messages are suppressed.

Value

A matrix of the value for each model on each test returned invisibly.

Side Effects

Criteria are tabulated for all models in the list.

See Also

[informationTestsCalculations](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model1 <- estVARXls(eg1.DSE.data.diff)
model2 <- estVARXar(eg1.DSE.data.diff)
informationTests(model1, model2)
```

<code>inputData</code>	<i>TSdata Series</i>
------------------------	----------------------

Description

Extract or set input or output series in a TSdata object.

Usage

```
inputData(x, series=seqN(nseriesInput(x)))
## Default S3 method:
inputData(x, series=seqN(nseriesInput(x)))
## S3 method for class 'TSdata':
inputData(x, series=seqN(nseriesInput(x)))
## S3 method for class 'TSEstModel':
inputData(x, series=seqN(nseriesInput(x)))

outputData(x, series=seqN(nseriesOutput(x)))
## Default S3 method:
outputData(x, series=seqN(nseriesOutput(x)))
## S3 method for class 'TSdata':
outputData(x, series=seqN(nseriesOutput(x)))
## S3 method for class 'TSEstModel':
outputData(x, series=seqN(nseriesOutput(x)))

inputData(x) <- value
outputData(x) <- value
```

Arguments

- `x` object of class TSdata.
- `value` a time series matrix.
- `series` vector of strings or integers indicating the series to select.

Value

The first usages returns the input or output series. The second usages assigns the input or output series.

See Also

[TSdata](#) [selectSeries](#)

Examples

```
data("eg1.DSE.data", package="dsel")
outputData(eg1.DSE.data)
```

1 . ARMA*Evaluate an ARMA TSmodel*

Description

Evaluate an ARMA TSmodel.

Usage

```
## S3 method for class 'ARMA':
l(obj1, obj2, sampleT=NULL, predictT=NULL, result=NULL,
  error.weights=0, compiled=.DSEflags()$COMPILED, warn=TRUE,
  return.debug.info=FALSE, ...)
```

Arguments

obj1	an 'ARMA' 'TSmodel' object.
obj2	a TSdata object.
sampleT	an integer indicating the number of periods of data to use.
predictT	an integer to what period forecasts should be extrapolated.
result	if non-NULL then the returned value is only the sub-element indicated by result. result can be a character string or integer.
error.weights	a vector of weights to be applied to the squared prediction errors.
compiled	indicates if a call should be made to the compiled code for computation. A FALSE value is mainly for testing purposes.
warn	if FALSE then certain warning messages are turned off.
return.debug.info	logical indicating if additional debugging information should be returned.
...	(further arguments, currently disregarded).

Details

This function is called by the function l() when the argument to l is an ARMA model (see [ARMA](#)). Using l() is usually preferable to calling I.ARMA directly. I.ARMA calls a compiled program unless compiled=FALSE. The compiled version is much faster.

sampleT is the length of data which should be used to calculate the one-step ahead predictions, and likelihood value for the model: Output data must be at least as long as sampleT. If sampleT is not supplied it is taken to be periods(data).

Input data must be at least as long as predictT. predictT must be at least as large as sampleT. If predictT is not supplied it is taken to be sampleT.

If error.weights is greater than zero then weighted prediction errors are calculated up to the horizon indicated by the length of error.weights. The weights are applied to the squared error at each period ahead.

Value

An object of class TSestModel (see TSestModel) containing the calculated likelihood, prediction, etc. for ARMA model.

See Also

[ARMA](#) [l](#) [l.SS](#) [TSmodel](#) [TSestModel](#).[object](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- TSmodel(estVARXls(eg1.DSE.data.diff))
evaluated.model <- l(model, eg1.DSE.data.diff)
```

l

*Evaluate a TSmodel***Description**

Evaluate a model with data.

Usage

```
l(obj1, obj2, ...)
## S3 method for class 'TSdata':
l(obj1, obj2, ...)
## S3 method for class 'TSestModel':
l(obj1, obj2, ...)
```

Arguments

obj1	a TSmodel, TSdata, or TSestModel object.
obj2	a TSmodel or TSdata object.
...	arguments to be passed to other methods.

Details

For state space models [l.SS](#) is called and for ARMA models [l.ARMA](#) is called.

Value

Usually a TSestModel object is returned. Most methods allow an argument `result` which specifies that a certain part of the object is returned. (This is passed in ... to another method in most cases.) The likelihood can be returned by specifying `result="like"`, which is useful for optimization routines.

See Also

[l.SS](#), [l.ARMA](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(TSmodel(estVARXls(eg1.DSE.data.diff)))
evaluated.model <- l(model, eg1.DSE.data.diff)
```

l.SS*Evaluate a state space TSmodel*

Description

Evaluate a state space TSmodel.

Usage

```
## S3 method for class 'SS':
l(obj1, obj2, sampleT=NULL, predictT=NULL, error.weights=0,
  return.state=FALSE, return.track=FALSE, result=NULL,
  compiled=.DSEflags()$COMPILED,
  warn=TRUE, return.debug.info=FALSE, ...)
```

Arguments

obj1	An 'SS' 'TSmodel' object.
obj2	A TSdata object.
sampleT	an integer indicating the last data point to use for one step ahead filter estimation. If NULL all available data is used.
predictT	an integer indicating how far past the end of the sample predictions should be made. For models with an input, input data must be provided up to predictT. Output data is necessary only to sampleT. If NULL predictT is set to sampleT.
error.weights	a vector of weights to be applied to the squared prediction errors.
return.state	if TRUE the element filter\$state containing $E[z(t) y(t-1), u(t)]$ is returned as part of the result. This can be a fairly large matrix.
return.track	if TRUE the element filter\$track containing the expectation of the tracking error given $y(t-1)$ and $u(t)$ is returned as part of the result. This can be an very large array.
result	if result is not specified an object of class TSEstModel is returned. Otherwise, the specified element of TSEstModel\$estimates is returned.
compiled	if TRUE the compiled version of the code is used. Otherwise the S/R version is used.
warn	if FALSE then certain warning messages are turned off.
return.debug.info	logical indicating if additional debugging information should be returned.
...	(further arguments, currently disregarded).

Details

This function is called by the function l() when the argument to l is a state space model. Using l() is usually preferable to calling l.SS directly. l.SS calls a compiled program unless compiled=FALSE. The compiled version is much faster than the S version.

Output data must be at least as long as sampleT. If sampleT is not supplied it is taken to be periods(data).

Input data must be at least as long as predictT. predictT must be at least as large as sampleT. If predictT is not supplied it is taken to be sampleT.

If error.weights is greater than zero then weighted prediction errors are calculated up to the horizon indicated by the length of error.weights. The weights are applied to the squared error at each period ahead.

sampleT is the length of data which should be used for calculating one step ahead predictions. y must be at least as long as sampleT. If predictT is large than sampleT then the model is simulated to predictT. y is used if it is long enough. u must be at least as long as predictT. The default result=0 returns a list of all the results. Otherwise only the indicated list element is returned (eg. result=1 return the likelihood and result=3 returns the one step ahead predictions).

If z0 is supplied in the model object it is used as the estimate of the state at time 0. If not supplied it is set to zero.

If rootP0 is supplied in the model object then t(rootP0) %*% rootP0 is used as P0. If P0 is supplied or calculated from rootP0 in the model object, it is used as the initial tracking error P(t=1|t=0). If not supplied it is set to the identity matrix.

Additional objects in the result are Om is the estimated output cov matrix. pred is the time series of the one-step ahead predictions, E[y(t)|y(t-1),u(t)]. The series of prediction error is given by y - pred. If error.weights is greater than zero then weighted prediction errors are calculated up to the horizon indicated by the length of error.weights. The weights are applied to the squared error at each period ahead. trackError is the time series of P, the one step ahead estimate of the state tracking error matrix at each period, Covz(t)-E[z(t)|t-1]. The tracking error can only be calculated if Q and R are provided (i.e. non innovations form models). Using the Kalman Innov K directly these are not necessary for the likelihood calculation, but the tracking error cannot be calculated.

Value

Usually an object of class TSestModel (see TSestModel), but see result above.

References

Anderson, B. D. O. and Moore, J. B. (1979) *Optimal Filtering*. Prentice-Hall. (note p.39,44.)

See Also

[SS 1 1.ARMA](#) [TSmodel](#) [TSestModel](#) [TSestModel.object](#) [state](#) [smoother](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(TSmodel(estVARXls(eg1.DSE.data.diff)))
lmodel <- l(model, eg1.DSE.data.diff)
summary(lmodel)
tfplot(lmodel)
lmodel <- l(model, eg1.DSE.data.diff, return.state=TRUE)
tfplot(state(lmodel, filter=TRUE))
```

makeTSnoise *Generate a random time series*

Description

Generate a random time series (matrix). This is a utility typically used in a time series model simulate method and not called directly by the user.

Usage

```
makeTSnoise(sampleT,p,lags,noise=NULL, rng=NULL, Cov=NULL, sd=1,
            noise.model=NULL, noise.baseline=0,
            tf=NULL, start=NULL,frequency=NULL)
```

Arguments

<code>sampleT</code>	an integer indicating the number of periods.
<code>p</code>	an integer indicating the number of series.
<code>lags</code>	an integer indicating the number of periods prior to the sample (initial data w0) for which random numbers should be generated. This is useful in ARMA models.
<code>noise</code>	Noise can be supplied. Otherwise it will be generated. If supplied it should be a list as described below under returned value.
<code>Cov</code>	The covariance of the noise process. If this is specified then <code>sd</code> is ignored. A vector or scalar is treated as a diagonal matrix. For an object of class <code>TSEstModel</code> , if neither <code>Cov</code> nor <code>sd</code> are specified, then <code>Cov</code> is set to the estimated covariance (<code>model\$estimates\$cov</code>).
<code>sd</code>	The standard deviation of the noise. This can be a vector.
<code>noise.model</code>	A <code>TSmodel</code> to be used for generating noise (not yet supported by SS methods).
<code>noise.baseline</code>	a constant or matrix to be added to noise. Alternately this can be a vector of length <code>p</code> , each value of which is treated as a constant to add to the corresponding noise series.
<code>rng</code>	The random number generator information needed to regenerate a simulation.
<code>tf</code>	a time frame to use for the generated matrix. (alternately use <code>start</code> and <code>frequency</code>)
<code>start</code>	a time start date to use for the generated matrix.
<code>frequency</code>	a time frequency to use for the generated matrix.

Value

A time series matrix.

markovParms	<i>Markov Parameters</i>
-------------	--------------------------

Description

Construct a Matrix of the Markov Parameters

Usage

```
markovParms(model, blocks=NULL)
```

Arguments

- | | |
|--------|--------------------------------|
| model | An ARMA or SS TSmodel. |
| blocks | Number of blocks to calculate. |

Details

Construct a matrix with partitions $[M_0|...|M_i]$ giving the Markov parameters M_i , $i+1 = \text{blocks}$ where each M_i is a p by $(m+p)$ matrix, (m is the dimension of the exogeneous series and p is the dimension of endogeneous series) ie. $y(t) = e(t) + M [u'(t)|y'(t-1) | u'(t-1)|y'(t-2)]'$ This requires that models be transformed so that lagged endogeneous variables are inputs. See Mitnik p1190. If $\text{blocks}=NULL$ (the default) then at least 3 blocks are generated, and up to $n+1$, but the series is truncated if the blocks are effectively zero. This will affect the size of the Hankel matrix.

Value

A matrix

References

See references for [MitnikReduction](#).

See Also

[SVDbalance](#)[Mitnik](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
markovParms(model)
```

McMillanDegree *Calculate McMillan Degree*

Description

Calculate the McMillan degree of an ARMA TSmodel.

Usage

```
McMillanDegree(model, ...)
## S3 method for class 'ARMA':
McMillanDegree(model, fuzz=1e-4, verbose=TRUE, warn=TRUE, ...)
## S3 method for class 'SS':
McMillanDegree(model, fuzz=1e-4, ...)
## S3 method for class 'TSEstModel':
McMillanDegree(model, ...)
```

Arguments

- | | |
|---------|--|
| model | An object of class TSmodel. |
| fuzz | Roots within fuzz distance are counted as equivalent. |
| verbose | If TRUE roots are printed. |
| warn | If FALSE then warnings about unit roots added for TREND are not printed. |
| ... | arguments to be passed to other methods. |

Value

A list with elements gross and distinct containing all roots and distinct roots.

Side Effects

The number of roots and distinct roots is printed if verbose is TRUE.

See Also

[stability](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
McMillanDegree(model)
```

MittnikReducedModels
Reduced Models via Mittnik SVD balancing

Description

Reduced Models via Mittnik SVD balancing.

Usage

```
MittnikReducedModels(largeModel)
```

Arguments

`largeModel` An SS TSmodel.

Details

The `largeModel` is balanced by the SVD method promoted by Mittnik (see `MittnikReduction`) and then models for every state dimension up to the state dimension of the `largeModel` are return. Note that this procedure does not result in smaller models which are balanced.

Value

A list of state space TSmodels with smaller state dimensions.

See Also

[MittnikReduction](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
z <- MittnikReducedModels(toSS(estVARXls(eg1.DSE.data.diff)))
```

MittnikReduction *Balance and Reduce a Model*

Description

Balance and reduce the state dimension of a state space model a la Mittnik.

Usage

```
MittnikReduction(model, data=NULL, criterion=NULL, verbose=TRUE, warn=TRUE)
MittnikReduction.from.Hankel(M, data=NULL, nMax=NULL,
                           criterion=NULL, verbose=TRUE, warn=TRUE)
```

Arguments

model	An object of class TSmodel or TSestModel.
data	If the supplied model is of class TSestModel and data is not supplied then it is taken from the model. If the model is of class TSmodel then data must be supplied.
criterion	Criterion to be used for model selection. see informationTestsCalculations.
verbose	logical indicating if information should be printed during estimation.
warn	logical indicating if some warning messages should be suppressed.
M	a matrix. See details.
nMax	integer indicating the state dimension of the largest model considered.

Details

`MittnikReduction` gives nested-balanced state space model using reduction by svd of the Hankel matrix generated from a model. If a state space model is supplied the max. state dimension for the result is taken from the model. If an ARMA model is supplied then singular values will be printed and the program prompts for the max. state dimension. criterion should be the name of one of the values returned by informationTests, that is, one of ("port", "like", "aic", "bic", "gvc", "rice", "fpe", "taic", "tbic", "tgvc", "trice", "tfpe"). If criteria is not specified then the program prompts for the state dimension (n) to use for the returned model. The program requires data to calculate selection criteria. (The program balanceMittnik calculates svd criteria only and can be used for reduction without data.)

The function `MittnikReduction.from.Hankel` is called by `MittnikReduction` and typically not by the user, but there are situations when the former might be called directly. It selects a reduced state space model by svd a la Mittnik. Models and several criteria for all state dimensions up to the max. state dim. specified are calculated. (If nMax is not supplied then svd criteria are printed and the program prompts for nMax). The output dimension p is taken from nrow(M). M is a matrix with p x (m+p) blocks giving the markov parameters, that is, the first row of the Hankel matrix. It can be generated from the model as in the function markovParms, or from the data, as in the function estSSMittnik.

`data` is necessary only if criteria (AIC,etc) are to be calculated.

Value

A state space model balance a la Mittnik in an object of class TSestModel.

References

- Gilbert, P. D. (1993) State space and ARMA models: An overview of the equivalence. Working paper 93-4, Bank of Canada. Available at www.bank-banque-canada.ca/pgilbert.
- Gilbert, P. D. (1995) Combining VAR Estimation and State Space Model Reduction for Simple Good Predictions. *J. of Forecasting: Special Issue on VAR Modelling*, **14**, 229-250.
- Mitnik, S. (1989), Multivariate Time Series Analysis With State Space Models. *Computers Math Appl.* **17**, 1189–1201.
- Mitnik, S. (1990), Macroeconomic Forecasting Experience With Balance State Space Models. *International Journal Of Forecasting*, **6**, 337–348.
- Mitnik, S. (1990), Forecasting With Balanced State Space Representations of Multivariate Distributed Lag Models. *J. of Forecasting*, **9**, 207–218.

See Also

[estVARXls](#) [bft](#) [balance](#) [Mittnik](#) [informationTests](#) [informationTestsCalculations](#)

Examples

```
data("egJoff.1dec93.data", package="dse1")
model <- toSS(estVARXls(egJoff.1dec93.data))
newmodel <- MittnikReduction(model, criterion="taic")
```

nseriesInput	<i>Number of Series in Input or Output</i>
--------------	--

Description

Number of input or output series in a TSdata object.

Usage

```
nseriesInput(x)
## Default S3 method:
nseriesInput(x)
## S3 method for class 'TSdata':
nseriesInput(x)
## S3 method for class 'SS':
nseriesInput(x)
## S3 method for class 'ARMA':
nseriesInput(x)
## S3 method for class 'TSestModel':
nseriesInput(x)

nseriesOutput(x)
## Default S3 method:
nseriesOutput(x)
## S3 method for class 'TSdata':
nseriesOutput(x)
## S3 method for class 'SS':
nseriesOutput(x)
## S3 method for class 'ARMA':
nseriesOutput(x)
## S3 method for class 'TSestModel':
nseriesOutput(x)
```

Arguments

x Object of class TSdata, TSmodel or TSestModel.

Value

An integer indicating the number of series.

See Also

[seriesNamesInput](#) [seriesNamesOutput](#)

Examples

```
data("eg1.DSE.data", package="dse1")
nseriesOutput(eg1.DSE.data)
```

nstates

State Dimension of a State Space Model

Description

Extract the state dimension of a state space model object.

Usage

```
nstates(x)
## S3 method for class 'SS':
nstates(x)
## S3 method for class 'ARMA':
nstates(x)
## S3 method for class 'TSestModel':
nstates(x)
```

Arguments

x Object of class TSmodel or TSestModel.

Value

An integer indicating the state dimension.

See Also

[nseriesInput](#)

observability

Calculate Model Observability Matrix

Description

Calculate the singular values of the observability matrix of a model.

Usage

```
observability(model)
## S3 method for class 'ARMA':
observability(model)
## S3 method for class 'SS':
observability(model)
## S3 method for class 'TSEstModel':
observability(model)
```

Arguments

`model` An object containing a TSmodel.

Details

If all singular values are significantly different from zero the model is observable.

Value

The singular values of the observability matrix.

See Also

[reachability](#), [stability](#) [McMillanDegree](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
observability(model)
```

`percentChange.TSdata`

Calculate percent change

Description

Calculate the percent change relative to the data lag periods prior.

Usage

```
## S3 method for class 'TSdata':
percentChange(obj, base=NULL, lag=1, cumulate=FALSE, e=FALSE, ...)
## S3 method for class 'TSEstModel':
percentChange(obj, base=NULL, lag=1, cumulate=FALSE, e=FALSE, ...)
```

Arguments

<code>obj</code>	An object of class <code>TSdata</code> or <code>TSeSTModel</code>
<code>e</code>	see the default method.
<code>base</code>	see the default method.
<code>lag</code>	see the default method.
<code>cumulate</code>	see the default method.
<code>...</code>	arguments passed to other methods.

Details

See [percentChange](#).

Value

For an object of class `TSdata` the percent change calculation is done with the output data and the result is an object of class `TSdata` (or a list of objects of class `TSdata`). For an object of class `TSeSTModel` the percent change calculation is done with `estimates$pred` and the result is an object of class `TSdata` (or a list of objects of class `TSdata`).

See Also

[percentChange](#) [ytoypc](#)

Examples

```
data("eg1.DSE.data", package="dse1")
z <- percentChange(outputData(eg1.DSE.data))
```

<code>periodsInput</code>	<i>TSdata Periods</i>
---------------------------	-----------------------

Description

Apply a method to the input or output data.

Usage

```
periodsInput(x)
## S3 method for class 'TSdata':
periodsInput(x)
## S3 method for class 'TSeSTModel':
periodsInput(x)

periodsOutput(x)
## S3 method for class 'TSdata':
periodsOutput(x)
## S3 method for class 'TSeSTModel':
periodsOutput(x)

startInput(x)
```

```
## S3 method for class 'TSdata':
startInput(x)
## S3 method for class 'TSEstModel':
startInput(x)

startOutput(x)
## S3 method for class 'TSdata':
startOutput(x)
## S3 method for class 'TSEstModel':
startOutput(x)

endInput(x)
## S3 method for class 'TSdata':
endInput(x)
## S3 method for class 'TSEstModel':
endInput(x)

endOutput(x)
## S3 method for class 'TSdata':
endOutput(x)
## S3 method for class 'TSEstModel':
endOutput(x)

frequencyInput(x)
## S3 method for class 'TSdata':
frequencyInput(x)
## S3 method for class 'TSEstModel':
frequencyInput(x)

frequencyOutput(x)
## S3 method for class 'TSdata':
frequencyOutput(x)
## S3 method for class 'TSEstModel':
frequencyOutput(x)
```

Arguments

x An object containing TSdata.

Details

Apply a method to the input or output data so, for example, periodsInput(x) in theory does periods(inputData(x)), which returns the number of periods in input data. The actual implementation may not do periods(inputData(x)). For example, with TSPADIdata inputData(x) requires a database retrieval which may be fairly slow, while the number of periods may be available much more quickly.

Value

Depends.

Examples

```
data("eg1.DSE.data.diff", package="dse1")
periodsOutput(eg1.DSE.data.diff)
```

periods.TSdata *Specific Methods for tframed Data*

Description

See the generic function description.

Usage

```
## S3 method for class 'TSdata':
periods(x, ...)
## S3 method for class 'TSestModel':
periods(x)
## S3 method for class 'TSdata':
start(x, ...)
## S3 method for class 'TSestModel':
start(x, ...)
## S3 method for class 'TSdata':
end(x, ...)
## S3 method for class 'TSestModel':
end(x, ...)
## S3 method for class 'TSdata':
frequency(x, ...)
## S3 method for class 'TSestModel':
frequency(x, ...)
```

Arguments

<code>x</code>	a time series object.
<code>...</code>	(further arguments, currently disregarded).

See Also

[periods](#), [start](#), [end](#), [frequency](#)

plot.roots *Plot Model Roots*

Description

Calculate and plot roots of a model.

Usage

```
## S3 method for class 'roots':
plot(x, pch='*', fuzz=0, ...)
```

Arguments

- x** An object of class roots (a vector of complex (or real) values as returned by the function roots).
- pch** character to be used for the plot (passed to plot.default).
- fuzz** If non-zero then roots within fuzz distance are considered equal.
- ...** (further arguments, currently disregarded).

Value

The eigenvalues of the state transition matrix or the inverse of the roots of the determinant of the AR polynomial are returned invisibly.

Side Effects

The roots and a unit circle are plotted on the complex plane.

See Also

[addPlotRoots](#) [roots](#) [stability](#) [McMillanDegree](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
plot(roots(model))
```

Description

Polynomial utility functions used by DSE.

Usage

```
characteristicPoly(a)
companionMatrix(a)
polyvalue(coef, z)
polydet(a)
polyprod(a, b)
polysum(a, b)
polyrootDet(a)
```

Arguments

- a** An array representing a matrix polynomial.
- b** An array representing a matrix polynomial.
- coef** Coefficients of a polynomial.
- z** Value at which the polynomial is to be evaluated.

Details

These are utility functions used in some ARMA model calculations such as root and stability calculations.

Value

depends

See Also

[polyroot](#) [roots](#) [stability](#)

[Portmanteau](#)

Calculate Portmanteau statistic

Description

Calculate Portmanteau statistic.

Usage

```
Portmanteau(res)
```

Arguments

`res` A matrix with time-series residuals in columns.

See Also

[informationTests](#)

Examples

```
require("stats")
Portmanteau(matrix(rnorm(200), 100, 2)) # but typically with a residual
```

[print.TSdata](#)

Print Specific Methods

Description

See the generic function description.

Usage

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

Arguments

- x An object of class TSdata.
- ... arguments to be passed to other methods.

See Also

[print summary](#)

[print.TSestModel](#) *Display TSmodel Arrays*

Description

Display TSmodel arrays.

Usage

```
## S3 method for class 'SS':
print(x, digits=options()$digits, latex=FALSE, ...)
## S3 method for class 'ARMA':
print(x, digits=options()$digits, latex=FALSE, L=TRUE, fuzz=1e-10, ...)
## S3 method for class 'TSestModel':
print(x, ...)
```

Arguments

- x An object of class TSmodel or TSestModel.
- digits the number of significant digits
- L logical if TRUE then ARMA model arrays are displayed as a polynomial matrix with L indicating lags. Otherwise, each lag in the array is displayed as a matrix.
- latex logical. If TRUE additional context is added to make the output suitable for inclusion in a latex document.
- fuzz ARMA model polynomial elements with absolute value less than fuzz are not displayed (i.e.-as if they are zero)
- ... arguments passed to other methods.

Value

The object is returned invisibly.

Side Effects

The model arrays are displayed.

Note

BUG: digits cannot be controlled for some numbers (e.g.- 1.0 is printed as 0.9999999999)

See Also

[print](#), [summary](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
print(model)
print(model, digits=3)
print(model, digits=3, fuzz=0.001)
print(model, digits=3, fuzz=0.001, latex=TRUE)
```

reachability

Calculate Model Reachability Matrix

Description

Calculate the singular values of the reachability matrix of a model.

Usage

```
reachability(model)
## S3 method for class 'ARMA':
reachability(model)
## S3 method for class 'SS':
reachability(model)
## S3 method for class 'TSestModel':
reachability(model)
```

Arguments

model An object containing TSmodel.

Details

If all singular values are significantly different from zero the model is controllable.

Value

The singular values of the reachability matrix.

See Also

[observability](#), [stability](#) [roots](#) [McMillanDegree](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
reachability(model)
```

residualStats *Calculate Residuals Statistics and Likelihood*

Description

Calculate the residuals statistics and likelihood of a residual.

Usage

```
residualStats(pred, data, sampleT=nrow(pred), warn=TRUE)
```

Arguments

<code>pred</code>	A matrix with columns representing time series.
<code>data</code>	A matrix with columns representing time series.
<code>sampleT</code>	An integer indicating the sample to use.
<code>warn</code>	If FALSE certain warnings are suppressed.

Details

Residuals are calculated as `pred[1:sampleT,drop=FALSE] - data[1:sampleT,drop=FALSE]` and then statistics are calculated based on these residuals. If `pred` or `data` are NULL they are treated as zero.

Value

A list with elements like, cov, pred, and sampleT. `pred` and `sampleT` are as supplied (and are returned as this is a utility function called by other functions and it is convenient to pass them along). `cov` is the covariance of the residual and `like` is a vector of four elements representing the total, constant, determinant and covariance terms of the negative log likelihood function.

See Also

[1](#)

Examples

```
residualStats(matrix(rnorm(200), 100, 2), NULL) # but typically used for a residual
```

`residuals.TSestModel`

Calculate the residuals for an object

Description

Calculate the residuals for an object.

Usage

```
## S3 method for class 'TSestModel':
residuals(object, ...)
```

Arguments

<code>object</code>	an object.
<code>...</code>	additional arguments passed to <code>stats::residuals</code> .

Details

Calculates the residuals (prediction minus data).

Value

A time series matrix.

Author(s)

Paul Gilbert

`Riccati`

Riccati Equation

Description

Solve a Matrix Riccati Equation

Usage

```
Riccati(A, B, fuzz=1e-10, iterative=FALSE)
```

Arguments

<code>A</code>	A matrix.
<code>B</code>	A matrix.
<code>fuzz</code>	The tolerance used for testing convergence.
<code>iterative</code>	If TRUE an iterative solution technique is used.

Details

Solve Riccati equation $P = APA' + B$ by eigenvalue decomposition of a symplectic matrix or by iteration.

Value

xxx

Note

This procedure has not been tested.

References

- Vaccaro, R. J. and Vukina, T. (1993), A Solution to the Positivity Problem in the State-Space Approach to Modeling Vector-Valued Time Series. *Journal of Economic Dynamics and Control*, **17**, 401–421.
- Anderson, B. D. O. and Moore, J. B. (1979) *Optimal Filtering*. Prentice-Hall. (note sec 6.7.)
- Vaughan, D. (1970) A Nonrecursive Algebraic Solution for the Discrete Riccati Equation. *IEEE Tr AC*, 597–599.
- Laub, A. J. (1983) Numerical Aspects of Solving Algebraic Riccati Equations *Proc IEEE conf Decision and Control*, 183–186.
- Gudmundsson T., Kenney, C., and Laub, A. J. (1992) Scaling of the Discrete-Time Algebraic Riccati Equation to Enhance Stability of the Schur Solution Method *IEEE Tr AC*, **37**, 513–518.

See Also

[eigen](#)

roots

Calculate Model Roots

Description

Calculate roots of a TSmodel.

Usage

```
roots(obj, ...)
## S3 method for class 'SS':
roots(obj, fuzz=0, randomize=FALSE, ...)
## S3 method for class 'ARMA':
roots(obj, fuzz=0, randomize=FALSE, warn=TRUE, by.poly=FALSE, ...)
## S3 method for class 'TSEstModel':
roots(obj, ...)
```

Arguments

<code>obj</code>	An object of class <code>TSmodel</code> .
<code>fuzz</code>	If non-zero then roots within fuzz distance are considered equal.
<code>randomize</code>	Randomly arrange complex pairs of roots so the one with the positive imaginary part is not always first (so random experiments are not biased).
<code>warn</code>	If FALSE then warnings about unit roots added for TREND are not printed.
<code>by.poly</code>	If TRUE then roots are calculated by expanding the determinant of the A polynomial. Otherwise, they are calculated by converting to a state space representation and calculating the eigenvalues of F. This second method is preferable for speed, accuracy, and because of a limitation in the degree of a polynomial which can be handled by <code>polyroot</code> .
<code>...</code>	arguments passed to other methods.

Details

The equality of roots for equivalent state space and ARMA models is illustrated in *Gilbert (1993)*. The calculation of ARMA model roots is more stable if the model is converted to state space and the roots calculated from the state transition matrix (see *Gilbert,2000*). The calculation is done this way by default. If `by.poly=TRUE` then the determinant of the AR polynomial is expanded to get the roots.

Value

The eigenvalues of the state transition matrix or the inverse of the roots of the determinant of the AR polynomial are returned.

References

- Gilbert, P. D. (1993) State space and ARMA models: An overview of the equivalence. Working paper 93-4, Bank of Canada. Available at www.bank-banque-canada.ca/pgilbert
- Gilbert, P.D. (2000) A note on the computation of time series model roots. *Applied Economics Letters*, 7, 423–424

See Also

`stability`, `McMillanDegree`

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
roots(model)
```

<code>scale.TSdata</code>	<i>Scale Methods for TS objects</i>
---------------------------	-------------------------------------

Description

Scale data or a model by a given factor.

Usage

```
## S3 method for class 'TSdata':
scale(x, center=FALSE, scale=NULL)
## S3 method for class 'TSestModel':
scale(x, center=FALSE, scale=NULL)
## S3 method for class 'ARMA':
scale(x, center=FALSE, scale=NULL)
## S3 method for class 'innov':
scale(x, center=FALSE, scale=NULL)
## S3 method for class 'nonInnov':
scale(x, center=FALSE, scale=NULL)

checkScale(x, scale)
## S3 method for class 'TSestModel':
checkScale(x, scale)
## S3 method for class 'TSmodel':
checkScale(x, scale)
```

Arguments

<code>x</code>	TSdata, TSmodel or an object containing these.
<code>center</code>	to match generic arguments, not currently used.
<code>scale</code>	A list with two matrices or vectors, named input and output, giving the multiplication factor for inputs and outputs. Vectors are treated as diagonal matrices. <code>scale\$input</code> can be NULL if no transformation is to be applied (or the data or model has no input.)

Value

The resulting data or model is different from the original in proportion to scale. ie. if S and T are output and input scaling matrices then $y'(t) = S y(t)$ where y' is the new output $u'(t) = S u(t)$ where u' is the new input

For models the result has inputs and outputs (and innovations) which are scaled as if data scaling had been applied to them as above. Thus if the input and output scales are diagonal matrices or scalars the plot of the predictions and residuals for `l(scale(model,scale=somescale), scale(data, scale=somescale))` while have the same appearance as `l(model, data)` but will be scaled differently.

See Also

`scale`

Examples

```
data("eg1.DSE.data.diff", package="dse1")
# This is a simple example. Usually scale would have something
# to do with the magnitude of the data.
z <- scale(eg1.DSE.data.diff,
           scale=list(input=rep(2, nseriesInput(eg1.DSE.data.diff)),
                      output=rep(2,nseriesOutput(eg1.DSE.data.diff))))
model <- estVARXls(eg1.DSE.data.diff)
model <- scale(model,
               scale=list(input=rep(2, nseriesInput(eg1.DSE.data.diff)),
                          output=rep(2,nseriesOutput(eg1.DSE.data.diff))))
```

seriesNamesInput *TSdata Series Names*

Description

Extract or set names of input or output series in a TSdata object.

Usage

```
seriesNamesInput(x)
## S3 method for class 'TSdata':
seriesNamesInput(x)
## S3 method for class 'TSmodel':
seriesNamesInput(x)
## S3 method for class 'TSEstModel':
seriesNamesInput(x)

seriesNamesOutput(x)
## S3 method for class 'TSdata':
seriesNamesOutput(x)
## S3 method for class 'TSmodel':
seriesNamesOutput(x)
## S3 method for class 'TSEstModel':
seriesNamesOutput(x)

seriesNamesInput(x) <- value
seriesNamesOutput(x) <- value
```

Arguments

- | | |
|--------------|--|
| x | Object of class TSdata, TSmodel or TSEstModel. |
| value | value to be assigned to object. |

Value

The first usages gives a vector of strings with the series names. The second usages assigns a vector of strings to be the series names of data.

See Also

[seriesNames](#)

Examples

```
data("egl.DSE.data", package="dse1")
seriesNamesOutput(egl.DSE.data)
```

`seriesNames.TSdata` *Series Names Specific Methods*

Description

See the generic function description.

Usage

```
## S3 method for class 'TSdata':
seriesNames(x)
## S3 method for class 'TSmodel':
seriesNames(x)
## S3 method for class 'TSEstModel':
seriesNames(x)

## S3 replacement method for class 'TSdata':
seriesNames(x) <- value
## S3 replacement method for class 'TSmodel':
seriesNames(x) <- value
## S3 replacement method for class 'TSEstModel':
seriesNames(x) <- value
```

Arguments

- `x` an object from which series names can be extracted or to which series names are to be assigned.
`value` series names to be assigned to data.

See Also

[seriesNames](#)

`setArrays`*Set TSmodel Array Information***Description**

Complete parameter array information based on parameter vector settings. This function is used internally and is not normally called by a user.

Usage

```
setArrays(model, coefficients=NULL, constants=NULL)
## S3 method for class 'ARMA':
setArrays(model, coefficients=NULL, constants=NULL)
## S3 method for class 'SS':
setArrays(model, coefficients=NULL, constants=NULL)
## S3 method for class 'TSEstModel':
setArrays(model, coefficients=NULL, constants=NULL)
```

Arguments

- `model` An object of class TSmodel.
- `coefficients` A vector of new values for the model coefficients (parameters).

Value

A TSmodel object.

`setTSmodelParameters`*Set TSmodel Parameter Information***Description**

Complete parameter vector information based on parameter array settings. This function is used internally and is not normally called by a user.

Usage

```
setTSmodelParameters(model, constants=model$constants)
## S3 method for class 'TSEstModel':
setTSmodelParameters(model, constants=TSmodel(model)$constants)
## S3 method for class 'ARMA':
setTSmodelParameters(model, constants=model$constants)
## S3 method for class 'SS':
setTSmodelParameters(model, constants=model$constants)
```

Arguments

- model** An object of class TSmodel.
constants A list of logical arrays indicating TRUE for any model array entries that should be treated as constants.

Value

An object of class 'TSmodel'.

See Also

[setArrays](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
model <- setTSmodelParameters(model)
```

simulate

Simulate a TSmodel

Description

Simulate a model to produce artificial data.

Usage

```
simulate(model, ...)
## S3 method for class 'ARMA':
simulate(model, y0=NULL, input=NULL, input0=NULL,
         start=NULL, freq=NULL, sampleT=100, noise=NULL, sd=1, Cov=NULL,
         rng=NULL, noise.model=NULL, compiled=.DSEflags()$COMPILED, ...)
## S3 method for class 'SS':
simulate(model, input=NULL,
         start=NULL, freq=NULL, sampleT=100, noise=NULL, sd=1, Cov=NULL,
         rng=NULL, compiled=.DSEflags()$COMPILED, ...)
## S3 method for class 'TSestModel':
simulate(model, input=inputData(model),
         sd=NULL, Cov=NULL, ...)
```

Arguments

- model** An object of class TSmodel or TSestModel.
input Data for the exogenous variable if specified in the model.
sampleT The length of the sample to simulate.
start start date for resulting data.
freq freq for resulting data.

<code>y0, input0</code>	Lagged values prior to t=1 for y and u, in reverse order so <code>y0[1,]</code> and <code>input0[1,]</code> correspond to t=0. These arguments are not implemented for state space models. If not specified initial values are set to zero.
<code>noise</code>	Noise can be supplied. Otherwise it will be generated. If supplied it should be a list as described below in details.
<code>Cov</code>	The covariance of the noise process. If this is specified then <code>sd</code> is ignored. A vector or scalar is treated as a diagonal matrix. For an object of class <code>TSeestModel</code> , if neither <code>Cov</code> nor <code>sd</code> are specified, then <code>Cov</code> is set to the estimated covariance (<code>model\$estimates\$cov</code>).
<code>sd</code>	The standard deviation of the noise. This can be a vector.
<code>noise.model</code>	A <code>TSmodel</code> to be used for generating noise (not yet supported by SS methods).
<code>rng</code>	The random number generator information needed to regenerate a simulation.
<code>compiled</code>	Specifies the compiled version of the code should be used (instead of the S code version).
<code>...</code>	arguments passed to other methods.

Details

A state space or ARMA model (see `TSmodel`, [ARMA](#), and [SS](#) for more details) is simulated with pseudo random noise (The default noise is a normally distributed processes. An object of class `TSdata` is returned. This can be used as input to estimation algorithms. If `start` and `freq` are specified, or if `input` or `noise$w` (in that order) have time series properties, these are given to the output.

If `noise` is not supplied then random values will be generated using other supplied information or defaults. The `rng` will be set first if it is specified.

The default noise generation will be $N(0,I)$ If Q is not square in a non innovations state space model (i.e. the system noise has a dimension less than the state dimension), then it is padded with zeros, so generated noise of higher dimension has no effect. If `sd` is supplied, then `w` as describe below will be $N(0,sqr(sd))$. `sd` can be a vector of p elements corresponding to each of the p outputs.

If `noise` is supplied it should be a list of the necessary noise processes. For non-innovation form state space models the list must have elements `w`, `e`, and `w0`. (`w0` is `w` for $t=0$ in state space model and prior lags in ARMA models.) For innovation form state space models and ARMA models with MA components the list should have elements `w` and `w0`, but if `w0` is not specified it is set to zero. For ARMA models with no MA components (i.e. VAR models) the list needs only `w`. In this case, and in the innovations form state space model with `w0=0`, a matrix may be supplied in place of a list. `w` should be a `sampleT` by p matrix giving the noise for $t=1$ to `sampleT`. If `noise` is specified `sampleT` will be set to the number of periods in `w`.

If `noise$w0` is a matrix (rather than a vector) for a state space model simulation (as it is for ARMA simulations) then it is set to a vector of zeros. This provides compatibility with VAR models (ARMA models with no lags in B).

Input must be specified for ARMA models with `model$C` not `NULL` and state space models with `model$G` not `NULL`.

In general ARMA and SS simulations will not produce exactly the same results because it is impossible to determine necessary transformation of initial conditions and `w0`.

Value

The value returned is an object of class `TSdata` which can be supplied as an argument to estimation routines. (See `TSdata`). In addition to the usual elements (see the description of a `TSdata` object) there are some additional elements: `model`- the generating model, `rng` - the initial RNG and seed,

version - the version of S used (random number generators may vary) Cov as specified sd as specified noise - the noise details as provided in the argument or as generated. state - the state variable for state space models.

See Also

[makeTSnoise](#), [TSmodel](#), [TSdata](#), [ARMA](#), [SS](#)

Examples

```
mod1 <- ARMA(A=array(c(1,-.25,-.05), c(3,1,1)), B=array(1,c(1,1,1)))
AR    <- array(c(1, .5, .3, 0, .2, .1, 0, .2, .05, 1, .5, .3) ,c(3,2,2))
VAR   <- ARMA(A=AR, B=diag(1,2))
print(VAR)
simData <- simulate(VAR)

C      <- array(c(0.5,0,0,0.2),c(1,2,2))
VARX <- ARMA(A=AR, B=diag(1,2), C=C)
simData <- simulate(VARX, sampleT=150, input=matrix(rnorm(300),150,2))

MA    <- array(c(1, .2, 0, .1, 0, 0, 1, .3), c(2,2,2))
ARMA  <- ARMA(A=AR, B=MA, C=NULL)
simData <- simulate(ARMA, sampleT=200)

ARMAX <- ARMA(A=AR, B=MA, C=C)
simData <- simulate(ARMAX, sampleT=150, input=matrix(rnorm(300),150,2))

data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
simData <- simulate(model)

ss <- SS(F=array(c(.5, .3, .2, .4), c(2,2)),
          H=array(c(1, 0, 0, 1), c(2,2)),
          K=array(c(.5, .3, .2, .4), c(2,2)))

print(ss)
simData <- simulate(ss)

testEqual(simData, simulate(ss))
testEqual(simData, simulate(ss, rng=getRNG(simData)))

simData2 <- simulate(ss,
                      noise=list(w=matrix(runif(300), 150,2), w0=runif(2)))

simData3 <- simulate(ss, noise=matrix(runif(400), 200,2))
```

Description

Evaluate a state space model.

Usage

```
smoother(model, data, compiled=.DSEflags()$COMPILED)
## S3 method for class 'nonInnov':
smoother(model, data, compiled=.DSEflags()$COMPILED)
## S3 method for class 'TSmodel':
smoother(model, data, compiled=.DSEflags()$COMPILED)
## S3 method for class 'TSestModel':
smoother(model, data=TSdata(model),
          compiled=.DSEflags()$COMPILED)
```

Arguments

<code>model</code>	An object of class ‘TSestModel’ or ‘TSmodel’ with a model of class ‘nonInnov’ ‘SS’ ‘TSmodel’. If filter informatin is not provided (i.e. in a TSestModel) then smoother runs the Kalman filter (l.SS) first.
<code>data</code>	A TSdata object.
<code>compiled</code>	If TRUE the compiled version of the code is used. Otherwise the S version is used.

Details

Calculate fixed interval smoother state values for a model. Smoother first runs the filter and uses the filtered state to calculate a smoothed estimate of the state (sometimes called a two sided filter). The smoother requires an non-innovations form model. The method for a TSmodel gives an error message if the model does not inherit from class nonInnov.

Note: this does not allow the same option as l.SS for calculating over a sub-sample. Smoothing is done over the length of the available filter data (which will be calculated to the length of the data if not supplied). For models with an input, smoothing will only be done over the input data period if that is shorter than the filter data.

See [SS](#) for details of the model:

$$z(t) = Fz(t-1) + Gu(t) + Qe(t) \quad y(t) = Hz(t) + Rw(t)$$

Value

An object of class TSestModel with an additional element `smooth`. `smooth` is a list of `state`, the smoothed state, and `track`, the smoothed tracking error. The result will also contain the element `filter` with `state` and `track` (which may or may not have been in the original arguement).

References

- Anderson, B. D. O. and Moore, J. B. (1979) *Optimal Filtering*. Prentice-Hall.
- Shumway R. H. and Stoffer D.S. (1982) An Approach to Time Series Smoothing and Forecasting Using the EM Algorithm. *J. of Time Series Analysis*, 3, 253–264 (note appendix).
- Jazwinski, A. H. (1970) *Stochastic Processes and Filtering Theory*. Academic Press.

See Also

[state](#), [l](#), [SS](#) [l.SS](#) [TSmodel](#) [TSestModel](#).[object](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
#smoother requires an non-innovations form model
model <- TSmodel(toSSChol(estVARXls(eg1.DSE.data.diff)))
smoothed.model <- smoother(model, eg1.DSE.data.diff, compiled=FALSE)
tfplot(state(smoothed.model))
tfplot(state(smoothed.model, filter=TRUE))
#compare
tfplot(state(smoothed.model, smoother=TRUE), state(smoothed.model, filter=TRUE))
```

SS

State Space Models

Description

Construct a

Usage

```
SS(F=NULL, G=NULL, H=NULL, K=NULL, Q=NULL, R=NULL, z0=NULL, P0=NULL, rootP0=NULL,
    constants=NULL,
    description=NULL, names=NULL, input.names=NULL, output.names=NULL)
is.SS(obj)
is.innov.SS(obj)
is.nonInnov.SS(obj)
```

Arguments

F.	(nxn) state transition matrix.
H	(pxn) output matrix.
Q	(nxn) matrix specifying the system noise distribution.
R	(pxp) matrix specifying the output (measurement) noise distribution.
G	(nxp) input (control) matrix. G should be NULL if there is no input.
K	(nxp) matrix specifying the Kalman gain.
z0	vector indicating estimate of the state at time 0. Set to zero if not supplied.
rootP0	matrix indicating a square root of the initial tracking error (e.g. chol(P0)).
P0	matrix indicating initial tracking error P(t=1 t=0). Set to I if rootP0 or P0 are not supplied.
constants	NULL or a list of logical matrices with the same names as matrices above, indicating which elements should be considered constants.
description	String. An arbitrary description.
names	A list with elements input and output, each a vector of strings. Arguments input.names and output.names should not be used if argument names is used.
input.names	A vector of character strings indicating input variable names.
output.names	A vector of character strings indicating output variable names.
obj	an object.

Details

State space models have a further sub-class: innov or non-innov, indicating an innovations form or a non-innovations form.

The state space (SS) model is defined by:

$$z(t) = Fz(t-1) + Gu(t) + Qe(t)$$

$$y(t) = Hz(t) + Rw(t)$$

or the innovations model:

$$z(t) = Fz(t-1) + Gu(t) + Kw(t-1)$$

$$y(t) = Hz(t) + w(t)$$

Matrices are as specified above in the arguments, and

y is the p dimensional output data.

u is the m dimensional exogenous (input) data.

z is the n dimensional (estimated) state at time t, $E[z(t)|y(t-1), u(t)]$ denoted $E[z(t)|t-1]$. Note: In the case where there is no input u this corresponds to what would usually be called the predicted state - not the filtered state. An initial value for z can be specified as $z0$ and an initial one step ahead state tracking error (for non-innovations models) as $P0$. In the object returned by `l.ss, state` is a time series matrix corresponding to z .

$z0$ An initial value for z can be specified as $z0$.

$P0$ An initial one step ahead state tracking error (for non-innovations models) can be specified as $P0$.

$rootP0$ Alternatively, a square root of $P0$ can be specified. This can be an upper triangular matrix so that only the required number of parameters are used.

K, Q, R For sub-class `innov` the Kalman gain K is specified but not Q and R . For sub-class `non-innov` Q and R are specified but not the Kalman gain K .

e and w are typically assumed to be white noise in the non-innovations form, in which case the covariance of the system noise is QQ' and the covariance of the measurement noise is RR' . The covariance of e and w can be specified otherwise in the simulate method `simulate.SS` for this class of model, but the assumption is usually maintained when estimating models of this form (although, not by all authors).

Typically, a non-innovations form is harder to identify than an innovations form. Non-innovations form would typically be chosen when there is considerable theoretical or physical knowledge of the system (e.g. the system was built from known components with measured physical values).

By default, elements in parameter matrices are treated as constants if they are exactly 1.0 or 0.0, and as parameters otherwise. A value of 1.001 would be treated as a parameter, and this is the easiest way to initialize an element which is not to be treated as a constant of value 1.0. Any matrix elements can be fixed to constants by specifying the list `constants`. Matrices which are not specified in the list will be treated in the default way. An alternative for fixing constants is the function `codefixConstants`

Value

An SS TSmodel

References

Anderson, B. D. O. and Moore, J. B. (1979) *Optimal Filtering*. Prentice-Hall. (note p.39,44.)

See Also

[TSmodel](#) [ARMA](#) [simulate](#).[SS](#) [l.SS](#) [state](#) [smoother](#) [fixConstants](#)

Examples

```
f <- array(c(.5,.3,.2,.4),c(2,2))
h <- array(c(1,0,0,1),c(2,2))
k <- array(c(.5,.3,.2,.4),c(2,2))
ss <- SS(F=f,G=NULL,H=h,K=k)
is.SS(ss)
ss
```

stability

*Calculate Stability of a TSmodel***Description**

Calculate roots and their modulus and indicate stability.

Usage

```
stability(obj, fuzz=1e-4, eps=1e-15, digits=8, verbose=TRUE)
## S3 method for class 'ARMA':
stability(obj, fuzz=1e-4, eps=1e-15, digits=8, verbose=TRUE)
## S3 method for class 'roots':
stability(obj, fuzz=1e-4, eps=1e-15, digits=8, verbose=TRUE)
## S3 method for class 'TSmodel':
stability(obj, fuzz=1e-4, eps=1e-15, digits=8, verbose=TRUE)
## S3 method for class 'TSEstModel':
stability(obj, fuzz=1e-4, eps=1e-15, digits=8, verbose=TRUE)
```

Arguments

obj	An object of class TSmodel.
fuzz	Roots within fuzz are considered equal.
eps	Roots with modulus less than (1-eps) are considered stable.
digits	Printing precision.
verbose	Print roots and there moduli.

Details

eps prevents the indication of a stable model when the largest root is within rounding error of 1.0.

Value

TRUE or FALSE if the model is stable or not stable.

Side Effects

The eigenvalues of the state transition matrix or the roots of the determinant of the AR polynomial are printed if verbose is T.

See Also

[McMillanDegree](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
stability(model)
```

state

Extract State

Description

Extract state information from estimated SS model.

Usage

```
state(obj, smoother=FALSE, filter=!smoother)
```

Arguments

- obj** An object of class ‘TSestModel’ with state information (filter or smoother) or containing an ‘SS’ model from which to estimate the state.
- smoother** logical indicating if the smoother state should be returned.
- filter** logical indicating if the filtered state should be returned.
- .

Details

One and only one of smoother and filter should be TRUE).

Value

A time series matrix of the estimated state series.

See Also

[smoother](#), [SS](#), [l.SS](#)

<code>summary.TSdata</code>	<i>Specific Methods for Summary</i>
-----------------------------	-------------------------------------

Description

See the generic function description.

Usage

```
## S3 method for class 'TSdata':
summary(object, ...)
## S3 method for class 'SS':
summary(object, ...)
## S3 method for class 'ARMA':
summary(object, ...)
## S3 method for class 'TSEstModel':
summary(object, ...)
## S3 method for class 'summary.TSdata':
print(x, digits=options()$digits, ...)
## S3 method for class 'summary.SS':
print(x, digits=options()$digits, ...)
## S3 method for class 'summary.ARMA':
print(x, digits=options()$digits, ...)
## S3 method for class 'summary.TSEstModel':
print(x, digits=options()$digits, ...)
```

Arguments

- `object` an object to be summarized.
- `x` a summary object to be printed.
- `digits` number of significant digits to use for printing.
- `...` arguments passed to other methods.

See Also

[print](#), [summary](#)

<code>sumSqerror</code>	<i>Calculate sum of squared prediction errors</i>
-------------------------	---

Description

Calculate a weighted sum squared prediction errors for a parameterization.

Usage

```
sumSqerror(coefficients, model=NULL, data=NULL, error.weights=NULL)
```

Arguments

coefficients A vector of coefficients (parameters).
model an object of class TSmodel which gives the structure of the model for which coefficients are used. `coef(model)` should be the same length as `coefficients`.
data an object of class TSdata which gives the data with which the model is to be evaluated.
error.weights a vector of weights to be applied to the squared prediction errors.

Details

This function is primarily for use in parameter optimization, which requires that an objective function be specified by a vector of parameters. It returns only the sum of the weighted squared errors (eg. for optimization). The sample size is determined by `periodsOutput(data)`.

Value

The value of the sum squared errors for a prediction horizon given by the length of `error.weights`. Each period ahead is weighted by the corresponding weight in `error.weights`.

See Also

[1.1.SS 1.ARMA](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
sumSqerror(1e-10 + coef(model), model=TSmodel(model),
           data=TSdata(model), error.weights=c(1,1,10))
```

Description

See the generic function description.

Usage

```
## S3 method for class 'ARMA':
testEqual(obj1, obj2, fuzz=0)
## S3 method for class 'SS':
testEqual(obj1, obj2, fuzz=0)
## S3 method for class 'TSdata':
testEqual(obj1, obj2, fuzz=1e-16)
## S3 method for class 'TSmodel':
testEqual(obj1, obj2, fuzz=0)
## S3 method for class 'TSEstModel':
testEqual(obj1, obj2, fuzz=0)
```

Arguments

- `obj1` see generic method.
- `obj2` see generic method.
- `fuzz` see generic method.

See Also

[testEqual](#)

`tfplot.TSdata` *Tfplot Specific Methods*

Description

See the generic function description.

Usage

```
## S3 method for class 'TSdata':
tfplot(x, ...,
        tf=NULL, start=tfstart(tf), end=tfend(tf),
        select.inputs = seq(length=nseriesInput(x)),
        select.outputs = seq(length=nseriesOutput(x)),
        Title=NULL, xlab=NULL, ylab=NULL,
        graphs.per.page=5, mar=par()$mar, reset.screen=TRUE)
## S3 method for class 'TSEstModel':
tfplot(x, ...,
        tf=NULL, start=tfstart(tf), end=tfend(tf),
        select.inputs=NULL, select.outputs=NULL,
        Title=NULL, xlab=NULL, ylab=NULL,
        graphs.per.page=5, mar=par()$mar, reset.screen=TRUE)
```

Arguments

- `x` object to be plotted.
- `...` additional objects to be plotted.
- `start` start of plot.
- `end` end of plot.
- `tf` an alternate way to specify start and end of plot.
- `select.inputs` series to be plotted. (passed to selectSeries)
- `select.outputs` series to be plotted. (passed to selectSeries)
- `Title` string to use for plot title (passed to plot - see tfplot).
- `xlab` string to use for x label (passed to plot).
- `ylab` string to use for y label (passed to plot).

```

graphs.per.page
                  integer indicating number of graphs to place on a page.
mar
                  margins passed to plot. See par.)
reset.screen logical indicating if the plot window should be cleared before starting. If this is
not TRUE then mar values will have no effect.

```

See Also

[tfplot](#)

[tframed.TSdata](#)

Specific Methods for tframed Data

Description

See the generic function description.

Usage

```

## S3 method for class 'TSdata':
tframed(x, tf=NULL, names=NULL, ...)
## S3 replacement method for class 'TSdata':
tframe(x) <- value
## S3 method for class 'TSdata':
tfwindow(x, tf=NULL, start=tfstart(tf), end=tfend(tf), warn=TRUE)
## S3 method for class 'TSdata':
tbind(x, d2, ..., pad.start=TRUE, pad.end=TRUE, warn=TRUE)
## S3 method for class 'TSdata':
trimNA(x, startNAs=TRUE, endNAs=TRUE)
## S3 method for class 'TSdata':
window(x, start=NULL, end=NULL, tf=NULL, warn=TRUE, ...)

```

Arguments

x	See the generic function.
tf	a time frame. See the generic function.
value	a time frame to associate with x.
names	A list with elements input and output which are strings passed as names to the default method.
start	See the generic function.
startNAs	See the generic function.
end	See the generic function.
endNAs	See the generic function.
d2	See the generic function.
pad.start	See the generic function.
pad.end	See the generic function.
warn	logical indicating if some warning messages should be suppressed.
...	arguments passed to other functions.

Details

The generic function is applied to input and to output data.

See Also

[tframed](#), [tfwindow](#), [tbind](#), [trimNA](#)

[toARMA](#)

Convert to an ARMA Model

Description

Convert a state space model to an ARMA representation. The state is eliminated by a method which uses an equivalence that can be demonstrated by the Cayley Hamilton theorem. It is not very parsimonious.

Usage

```
toARMA(model, ...)
## S3 method for class 'ARMA':
toARMA(model, ...)
## S3 method for class 'SS':
toARMA(model, fuzz=1e-10, ...)
## S3 method for class 'TSestModel':
toARMA(model, ...)
```

Arguments

model	An object of class TSmodel.
fuzz	Parameters closer than fuzz to one or zero are set to 1.0 or 0.0 respectively
...	arguments to be passed to other methods.

Value

An object of class 'ARMA' 'TSmodel' containing an ARMA model.

References

- See, for example, Aoki, M. (1990) *State Space Modelling of Time Series*. 2d ed. rev. and enl., Springer-Verlag.
 Aoki, M. and Havenner, A. (1991) State Space Modeling of Multiple Time Series. *Econometric Reviews*, **10**, 1–59.

See Also

[toSS](#) [fixConstants](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
model <- toARMA(model)
```

toSSChol*Convert to Non-Innovation State Space Model***Description**

This function may not be working properly.

Convert to a non-innovations state space representation using the given matrix (Om) as the measurement noise covariance. Om would typically be an estimate of the output noise, such as returned in `$estimates$cov` of the function `l (l.SS or l.ARMA)`. This assumes that the noise processes in the arbitrary SS representation are white and uncorrelated.

Usage

```
toSSChol(model, ...)
## S3 method for class 'TSmodel':
toSSChol(model, Om=diag(1,nseriesOutput(model)), ...)
## S3 method for class 'TSEstModel':
toSSChol(model, Om=NULL, ...)
```

Arguments

- | | |
|--------------------|---|
| <code>model</code> | An object of class <code>TSmodel</code> . |
| <code>Om</code> | a matrix to be used as the measurement noise covariance. If <code>Om</code> is not supplied and <code>model</code> is of class <code>TSEstModel</code> then <code>model\$estimates\$cov</code> is used. Otherwise, <code>Om</code> is set to the identity matrix. |
| <code>...</code> | arguments to be passed to other methods. |

Details

Convert to a non-innovations SS representation using a Cholesky decomposition of `Om` as the coefficient matrix of the output noise.

Value

An object of class 'SS' 'TSmodel' containing a state space model which is not in innovations form.

See Also

[toSSInnov](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
model <- toSSChol(model)
```

toSSinnov*Convert to State Space Innovations Model*

Description

Convert to a state space innovations representation.

Usage

```
toSSinnov(model, ...)
```

Arguments

- | | |
|-------|------------------------------------|
| model | an object of class TSmodel. |
| ... | arguments passed to other methods. |

Value

If the argument is a TSmodel then the result is an object of class 'SS' 'TSmodel'. If the argument is a TSEstModel then the converted model is evaluated with the data and a TSEstModel is returned. The TSmodel is an innovations state space representation.

This assumes that the noise processes in the arbitrary SS representation are white and uncorrelated.

See Also

[toSS](#), [toSSOform](#) [toSSChol](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
model <- toSSinnov(model)
summary(model)

model2 <- SS(F=diag(1,3), H=matrix(c(1,0,0,1,0,0),2,3),
               Q=diag(0.5, 3, 3), R=diag(1.1, 2,2),
               description="test model", output.names=c("output 1", "output 2"))
model2 <- toSSinnov(model2)
summary(model2)
```

toSSOform*Convert to Oform*

Description

Convert a state space model to (observability?) form.

Usage

```
toSSOform(model)
## S3 method for class 'TSmodel':
toSSOform(model)
## S3 method for class 'TSEstModel':
toSSOform(model)
```

Arguments

`model` An object of class TSmodel.

Details

WARNING: This function does not work properly.

Convert to a SS innovations representation with a minimum number of parameters by converting as much of H as possible to I matrix. Any remaining reductions are done by converting part of ?? to I. It seems there should remain $n(m+2p)$ free parameters in F,G,H,K, and Om is determined implicitly by the residual.

Value

An object of class 'SS' 'TSmodel' containing a state space model in observability form (more or less).

See Also

[toSSinnov](#)

Examples

```
data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
```

toSS

Convert to State Space Model

Description

Convert a model to state space form.

Usage

```
toSS(model, ...)
## S3 method for class 'ARMA':
toSS(model, ...)
## S3 method for class 'SS':
toSS(model, ...)
## S3 method for class 'TSEstModel':
toSS(model, ...)
```

```

toSSaugment(model, ...)
## S3 method for class 'ARMA':
toSSaugment(model, fuzz=1e-14, ...)
## S3 method for class 'TSEstModel':
toSSaugment(model, ...)

toSSnested(model, ...)
## S3 method for class 'ARMA':
toSSnested(model, n=NULL, Aoki=FALSE, ...)
## S3 method for class 'SS':
toSSnested(model, n=NULL, Aoki=FALSE, ...)
## S3 method for class 'TSEstModel':
toSSnested(model, ...)

```

Arguments

model	An object of class TSmodel.
n	If n is specified then it is used as the state dimension when the markov parameter conversion technique is required.
Aoki	logical indicating if Aoki's method (which does not work in general) should be tried.
fuzz	if the zero lag term of polynomials A and B are within fuzz of the identity matrix then they are not inverted. (i.e. they are assumed to be identity.)
...	arguments to be passed to other methods.

Details

If the order of the AR polynomial equals or exceeds the MA polynomial (and the input polynomial) then the model is converted by state augmentation. Otherwise, it is converted by approximating the markov coefficients a la Mitnik. (This may not always work very well. Compare the results to check.)

Value

A state space model in an object of class 'SS' 'TSmodel'.

Examples

```

data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
model <- toSS(model)

```

TSdata.object *time series data object*

Description

Class "TSdata" of time series data objects for use with TSmodels.

Generation

This class of objects is returned by specific methods of the function TSdata or can be built according to the description below.

Methods

The TSdata class of objects has methods for the generic functions `print`, `plot`, `periods`, `start`, `end`, ..., `testEqual`, `seriesNames`, `seriesNamesInput`, `seriesNamesOutput`. Also, the function `is.TSdata` is supported.

Inheritance

Other data classes inherit from the class TSdata.

Structure

Objects are a list with class the most general class "TSdata". The native form for this package has elements `input` and `output`. Any other elements are ignored. `input` and `output` are matrices (or tframe or time series matrices) of the input and output data, with each series in a column. TSPADIdata objects inherit from this class but have a somewhat different structure. TSPADIdata makes it possible to retrieve data from an external database when it is needed. These subclass objects do not contain the actual data, but only the names of the series and the data base where they are located. The function `setTSPADIdata` can be used to set up an object of class `c("TSPADIdata", "TSdata")`.

See Also

`TSdata` `TSmodel` `TSeestModel`.`object` `TSPADIdata`

`TSdata`

Construct TSdata time series object

Description

Constructor for generating or extracting a "TSdata" object containing data for use by TSmodels.

Usage

```
TSdata(data=NULL, ...)
## Default S3 method:
TSdata(data=NULL, input=NULL, output=NULL, ...)
## S3 method for class 'TSdata':
TSdata(data, ...)
## S3 method for class 'TSeestModel':
TSdata(data, ...)
is.TSdata(obj)
as.TSdata(d)
```

Arguments

<code>data</code>	object of class TSdata, TSestModel, matrix, list with input and output matrices, or another object for which a constructor or TSdata extraction method has been defined.
<code>input</code>	a matrix of time series data.
<code>output</code>	a matrix of time series data.
<code>...</code>	arguments to be passed to other methods.
<code>obj</code>	an object.
<code>d</code>	an object from which a TSdata object can be extracted. See below.

Details

Generic method to construct or extract a TSdata object. The default method constructs a TSdata object. Specific methods extract the TSdata from other objects (which must contain TSdata). The function `is.TSdata(data)` returns TRUE if `data` inherits from "TSdata" and FALSE otherwise.

The function `as.TSdata` uses the elements `input` and `output` directly and strips away other class information and parts of the object (and does not make use of `inputData(data)` or `outputData(data)` which may do something special for certain classes).

See Also

`TSdata.object`, `freeze`, `freeze.TSPADIdata`, `TSmodel`, `TSestModel.object` `setTSPADIdata`

Examples

```
rain <- matrix(rnorm(86*17), 86,17)
radar <- matrix(rnorm(86*5), 86,5)
mydata <- TSdata(input=radar, output=rain)
```

Description

Object containing a time series model, data, and estimation information.

Usage

```
TSestModel(obj)
## S3 method for class 'TSestModel':
TSestModel(obj)
is.TSestModel(obj)
```

Arguments

<code>obj</code>	in the first usage an object from which a TSestModel object can be extracted (or constructed).
------------------	--

Details

The TSestModel class of objects are generated by estimation methods. See, for example, `estVARXls`. They contains a time series model (TSmodel), data (TSdata), and information obtained by evaluating the model with the data in an element called `estimates` containing:

`like` The negative log likelihood function value (a vector of the total, constant, the det part, and the cov part)

`cov` The estimated residual covariance.

`pred` The one step ahead predictions (see `predictT` below). These are aligned with output data so that residuals are `pred[1:sampleT,] - output[1:sampleT,]`

`sampleT` The end of the period (starting from 1) for which output is used for calculating one step ahead predictions.

`predictT` The end of the period for which the model is simulated. `sampleT` must be less than or equal `predictT`. If `predictT` is greater than `sampleT` then each step ahead beyond `sampleT` is based on the prediction of the previous step and not corrected by the prediction error.

The element `estimates` may optionally also contain and element `filter` which may have

`state` The one step ahead (filter) estimate of the state $E[z(t)|y(t-1), u(t)]$. Note: In the case where there is no input u this corresponds to what would usually be called the predicted state - not the filtered state.

`track` The estimated state tracking error $P(t|t-1)$. Again note, this corresponds to the predicted tracking error not the filtered tracking error. This is NULL for innovations models.

`smooth` a list of:

`state` The smoother (two sided filter) estimate of the state $E[z(t)|\text{sampleT}]$.

`track` The smoothed estimate of the state tracking error $P(t|\text{sampleT})$. This is NULL for innovations models.

See Also

`estVARXls`, `TSmodel`, `TSdata`

`TSmodel`

Time Series Models

Description

Construct or extract a "TSmodel" from objects.

Usage

```
TSmodel(obj, ...)
## S3 method for class 'TSmodel':
TSmodel(obj, ...)
## S3 method for class 'TSestModel':
TSmodel(obj, ...)
is.TSmodel(obj)
```

Arguments

- `obj` An object containing an object of class TSmodel or a list containing the information necessary to build an object of class "TSmodel".
`...` arguments passed to other methods.

Details

This is a generic method which will extract a "TSmodel" from an object (e.g. a TSestModel). The default method will try to build an ARMA or state-space "TSmodel" from a list, which must contain the necessary information.

This class of objects is returned by estimation methods or can be built according to the description for specific sub-classes (eg "ARMA", "SS").

The TSmodel class of objects has methods for the generic functions `print`, `testEqual`, `seriesNames`, `seriesNamesInput`, `seriesNamesOutput`, `l`, `roots`, `stability`, `forecast`, `featherForecasts`, `horizonForecasts`, `simulate`, `MonteCarloSimulations`. Also, the function `is.TSmodel` and the functions `toSS`, `toARMA`, `to.troll` are supported. Other model classes inherit from the class TSmodel.

This class of objects contains a time series model. It is the class of objects expected by many of the functions in this package.

Sub-class (e.g. ARMA and SS for linear, time-invariant ARMA and state space models.) are documented individually. Many of the functions in this package are designed for estimating and converting among various representations of these types of models.

See Also

[ARMA](#), [SS](#), [TSestModel](#), [TSdata](#)

Index

*Topic **algebra**
markovParms, 41
Riccati, 56

*Topic **datasets**
egl.DSE.data, 17
egJofF.1dec93.data, 18

*Topic **internal**
acf, 2
DSEutilities, 16
makeTSnoise, 40
residuals.TSestModel, 56
setArrays, 62
setTSmodelParameters, 62

*Topic **package**
00.dse.Intro, 1

*Topic **programming**
DSEflags, 13

*Topic **ts**
addPlotRoots, 2
ARMA, 3
balanceMitnik, 5
bestTSestModel, 6
checkBalance, 8
checkBalanceMitnik, 7
checkConsistentDimensions, 9
checkResiduals, 10
coef.TSmodel, 11
combine, 12
combine.TSdata, 12
dse-package, 14
DSEversion, 16
estBlackBox, 23
estBlackBox1, 19
estBlackBox2, 20
estBlackBox3, 21
estBlackBox4, 22
estMaxLik, 24
estSSfromVARX, 25
estSSMitnik, 26
estVARXar, 27
estVARXls, 29
estWtVariables, 30
fixConstants, 31

fixF, 32
gmap, 32
informationTests, 34
informationTestsCalculations, 33
inputData, 35
l, 37
l.ARMA, 36
l.SS, 38
markovParms, 41
McMillanDegree, 42
MitnikReducedModels, 43
MitnikReduction, 43
nseriesInput, 45
nstates, 46
observability, 46
percentChange.TSdata, 47
periods.TSdata, 50
periodsInput, 48
plot.roots, 50
Polynomials, 51
Portmanteau, 52
print.TSdata, 52
print.TSestModel, 53
reachability, 54
residualStats, 55
Riccati, 56
roots, 57
scale.TSdata, 59
seriesNames.TSdata, 61
seriesNamesInput, 60
simulate, 63
smoother, 65
SS, 67
stability, 69
state, 70
summary.TSdata, 71
sumSqerror, 71
testEqual.ARMA, 72
tfplot.TSdata, 73
tframed.TSdata, 74
toARMA, 75
toSS, 78

toSSChol, 76
 toSSInnov, 77
 toSSOform, 77
 TSdata, 80
 TSdata.object, 79
 TSestModel, 81
 TSmodel, 82
 .DSEflags (*DSEflags*), 13
 00.dse.Intro, 1

 acf, 2, 2
 addPlotRoots, 2, 51
 ar, 28
 ARMA, 3, 14, 25, 36, 37, 64, 65, 69, 83
 as.TSdata (*TSdata*), 80

 balanceMittnik, 5, 45
 bestTSestModel, 6
 bft, 15, 25–28, 30, 45
 bft (*estBlackBox4*), 22

 characteristicPoly (*Polynomials*),
 51
 checkBalance, 7, 8
 checkBalanceMittnik, 7, 8
 checkConsistentDimensions, 9
 checkResiduals, 10, 15
 checkScale (*scale.TSdata*), 59
 coef.TSestModel (*coef.TSmodel*), 11
 coef.TSmodel, 11
 coef<- (*coef.TSmodel*), 11
 combine, 12
 combine.TSdata, 12
 companionMatrix (*Polynomials*), 51
 criteria.table.heading
 (*DSEutilities*), 16
 criteria.table.legend
 (*DSEutilities*), 16
 criteria.table.nheading
 (*DSEutilities*), 16

 dse (*dse-package*), 14
 dse-package, 1
 dse-package, 14
 DSE.ar, 28
 DSE.ar (*DSEutilities*), 16
 dse.Intro (*dse-package*), 14
 dseclass (*DSEutilities*), 16
 dseclass<- (*DSEutilities*), 16
 DSEflags, 13
 DSEutilities, 16
 DSEversion, 16

 eg1.dat (*eg1.DSE.data*), 17

 eg1.DSE.data, 17
 egJoffF.1dec93.data, 18
 eigen, 57
 end, 50
 end.TSdata (*periods.TSdata*), 50
 end.TSestModel (*periods.TSdata*),
 50
 endInput (*periodsInput*), 48
 endOutput (*periodsInput*), 48
 estBlackBox, 23, 30
 estBlackBox1, 7, 19, 21–23
 estBlackBox2, 7, 20, 22, 23
 estBlackBox3, 7, 21, 21, 23
 estBlackBox4, 7, 21, 22
 EstEval, 15
 estMaxLik, 4, 15, 24, 26, 28, 30
 estSSfromVARX, 15, 25, 28, 30
 estSSMittnik, 26, 26, 28, 30
 estVARXar, 6, 14, 27, 30
 estVARXls, 4, 6, 14, 25–28, 29, 30, 45, 82
 estVARXmean.correction
 (*DSEutilities*), 16
 estWtVariables, 30

 fake.TSestModel.missing.data
 (*DSEutilities*), 16
 featherForecasts, 15
 fixConstants, 4, 25, 31, 32, 69, 75
 fixF, 31, 32
 forecast, 15
 forecastCov, 15
 freeze, 81
 freeze.TSPADIdata, 81
 frequency, 50
 frequency.TSdata
 (*periods.TSdata*), 50
 frequency.TSestModel
 (*periods.TSdata*), 50
 frequencyInput (*periodsInput*), 48
 frequencyOutput (*periodsInput*), 48

 gmap, 32

 horizonForecasts, 15

 informationTests, 11, 15, 33, 34, 45, 52
 informationTestsCalculations, 7,
 19, 21–23, 33, 34, 45
 inputData, 35
 inputData<- (*inputData*), 35
 is.ARMA (ARMA), 3
 is.innov.SS (SS), 67
 is.nonInnov.SS (SS), 67

is.SS (*SS*), 67
 is.TSdata (*TSdata*), 80
 is.TSestModel (*TSestModel*), 81
 is.TSmodel (*TSmodel*), 82
 1, 14, 15, 25, 37, 37, 39, 55, 66, 72
 1.ARMA, 36, 37, 39, 72
 1.SS, 37, 38, 66, 69, 70, 72
 makeTSnoise, 40, 65
 markovParms, 41
 McMillanDegree, 15, 42, 47, 51, 54, 58, 70
 MittnikReducedModels, 43
 MittnikReduction, 6–8, 27, 41, 43, 43
 MittnikReduction.from.Hankel
 (*MittnikReduction*), 43
 MonteCarloSimulations, 15
 nlm, 25
 nseriesInput, 45, 46
 nseriesOutput (*nseriesInput*), 45
 nstates, 46
 observability, 46, 54
 old.estVARXar (*estVARXar*), 27
 optim, 25
 outputData (*inputData*), 35
 outputData<– (*inputData*), 35
 percentChange, 48
 percentChange.TSdata, 47
 percentChange.TSestModel
 (*percentChange.TSdata*), 47
 periods, 50
 periods.TSdata, 50
 periods.TSestModel
 (*periods.TSdata*), 50
 periodsInput, 48
 periodsOutput (*periodsInput*), 48
 plot.roots, 3, 50
 polydet (*Polynomials*), 51
 Polynomials, 51
 polyprod (*Polynomials*), 51
 polyroot, 52
 polyrootDet (*Polynomials*), 51
 polysum (*Polynomials*), 51
 polyvalue (*Polynomials*), 51
 Portmanteau, 11, 52
 print, 53, 54, 71
 print.ARMA (*print.TSestModel*), 53
 print.SS (*print.TSestModel*), 53
 print.summary.ARMA
 (*summary.TSdata*), 71
 print.summary.SS
 (*summary.TSdata*), 71
 print.summary.TSdata
 (*summary.TSdata*), 71
 print.summary.TSestModel
 (*summary.TSdata*), 71
 print.TSdata, 52
 print.TSestModel, 53
 printTestValue (*DSEutilities*), 16
 reachability, 47, 54
 residuals.TSestModel, 56
 residualStats, 55
 Riccati, 56
 roots, 15, 51, 52, 54, 57
 scale, 59
 scale.ARMA (*scale.TSdata*), 59
 scale.innov (*scale.TSdata*), 59
 scale.nonInnov (*scale.TSdata*), 59
 scale.TSdata, 59
 scale.TSestModel (*scale.TSdata*),
 59
 selectSeries, 35
 seriesNames, 61
 seriesNames.TSdata, 61
 seriesNames.TSestModel
 (*seriesNames.TSdata*), 61
 seriesNames.TSmodel
 (*seriesNames.TSdata*), 61
 seriesNames<-.TSdata
 (*seriesNames.TSdata*), 61
 seriesNames<-.TSestModel
 (*seriesNames.TSdata*), 61
 seriesNames<-.TSmodel
 (*seriesNames.TSdata*), 61
 seriesNamesInput, 46, 60
 seriesNamesInput<–
 (*seriesNamesInput*), 60
 seriesNamesOutput, 46
 seriesNamesOutput
 (*seriesNamesInput*), 60
 seriesNamesOutput<–
 (*seriesNamesInput*), 60
 setArrays, 62, 63
 setTSmodelParameters, 62
 setTSPADIdata, 81
 simulate, 14, 63
 simulate.ARMA, 4
 simulate.SS, 69
 smoother, 14, 39, 65, 69, 70
 SS, 14, 25, 39, 64–66, 67, 70, 83
 stability, 15, 42, 47, 51, 52, 54, 58, 69

start, 50
start.TSdata (*periods.TSdata*), 50
start.TSestModel
 (*periods.TSdata*), 50
startInput (*periodsInput*), 48
startOutput (*periodsInput*), 48
state, 39, 66, 69, 70
summary, 53, 54, 71
summary.ARMA (*summary.TSdata*), 71
summary.SS (*summary.TSdata*), 71
summary.TSdata, 71
summary.TSestModel
 (*summary.TSdata*), 71
sumSqerror, 71
svd.criteria (*DSEutilities*), 16
SVDbalanceMittnik, 41
SVDbalanceMittnik
 (*balanceMittnik*), 5

tbinding, 75
tbinding.TSdata (*tframed.TSdata*), 74
testEqual, 73
testEqual.ARMA, 72
testEqual.SS (*testEqual.ARMA*), 72
testEqual.TSdata
 (*testEqual.ARMA*), 72
testEqual.TSestModel
 (*testEqual.ARMA*), 72
testEqual.TSmodel
 (*testEqual.ARMA*), 72
tfplot, 74
tfplot.TSdata, 73
tfplot.TSestModel
 (*tfplot.TSdata*), 73
tframe<- TSdata (*tframed.TSdata*),
 74
tframed, 75
tframed.TSdata, 74
tfwindow, 75
tfwindow.TSdata (*tframed.TSdata*),
 74
toARMA, 14, 75
toSS, 14, 26, 75, 77, 78
toSSaugment (*toSS*), 78
toSSChol, 76, 77
toSSInnov, 76, 77, 78
toSSnested (*toSS*), 78
toSSOform, 77, 77
trimNA, 75
trimNA.TSdata (*tframed.TSdata*), 74
TSdata, 14, 16–18, 35, 65, 80, 80, 82, 83
TSdata.object, 79, 81
TSestModel, 14, 39, 81, 83
TSestModel.object, 16, 37, 39, 66, 80,
 81
TSmodel, 4, 14, 16, 25, 37, 39, 65, 66, 69, 80,
 81, 82, 82
TSPADIdata, 80
window.TSdata (*tframed.TSdata*), 74
ytoypc, 48