# Package 'dse1'

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# **Description**

This is a library of functions for time series modeling. The library works with the S (Splus) and R languages. A "Brief User's Guide" is available at <a href="www.bank-banque-canada.ca/pgilbert">www.bank-banque-canada.ca/pgilbert</a> This help and other information is also available at that web location.

The library implements an object oriented approach to time series modeling. This means that different model and data representations can be implemented with fairly simple extensions to the library.

The library includes multi-variate state space and ARMA (including VAR) models. The library also implements Troll models as an example of another class of model. (Troll models are run by completely separate software, Troll from Intex Inc, as if they were an integral part of the library. Models and data are passed back and forth to Troll and the results can be analyzed with already existing functions in the library. Note: the Troll interface is broken in the current version.)

The library includes methods for simulating, estimating, and converting among different model representations. These form the basic part of the library. Methods for studying estimation techniques and for examining the forecasting properties of models are also documented in the User's Guide.

There are also functions for forecasting and for evaluating the performance of forecasting models as well as functions for evaluating model estimation techniques.

DSE requires libraries tframe and setRNG. The first provides a kernel of functions for programming time series methods relatively independently of the representation of time. The second provides some utilities for setting and resetting the random number generator and normal transformations, and for generating the same random numbers in S and R differences. For programmers, these libraries may of interest by themselves. Tframe is intended to make it easier to write code which can use any new/better time representations when they appear. It also provides plotting, time windowing, and some other utility functions which are specifically intended for time series.

Relative to commercial packages the library is especially useful for time series research (such as studying estimation methods). For usual time series applications there may be commercial packages which are preferable.

The functions described in the Brief User's Guide should work fairly reliably, however, many of the functions described in the help facility are still under development and may not work. In addition,

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there may be functions described in the help facility for which the code is not yet distributed. 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 an ongoing research effort. Constructive suggestions and comments are welcomed. I can be reached at  $\langle pgilbert@bank-banque-canada.ca \rangle$  or by phone at (613) 782-7346.

There is also a mailing list available which I hope will let users help one another, as I cannot always be available. Please subscibe if you use the library extensively, especially if you can help other users. To subscribe send mail to  $\langle boc\_list@bank-banque-canada.ca \rangle$  with the body line subscribe boc\_dse Your Name

### Usage

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

### **MainObjects**

The main objects are classes "TSdata", "TSmodel", and "TSestModel". These are, respectively, time series data objects, models, and objects with a model, data and some estimation information. For each of these there are several sub classes.

#### See Also

```
TSdata, TSmodel, TSestModel.object
```

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.
В	The moving-average polynomial, an bxpxp array.
C	The input polynomial, an expxm 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.

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#### **Details**

```
The ARMA model is defined by: A(L)y(t) = B(L)w(t) + C(L)u(t) + TREND(t) \label{eq:arma} where
```

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

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 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 codefixConstants

# Value

An ARMA TSmodel

# See Also

TSmodel simulate.ARMA

# **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)</pre>
```

6 DSEutilities

DSEflags

Flags to Indicate Use of Compiled Code

# **Description**

Determines if compiled code should be used or not.

# Usage

```
.DSEflags(new)
.DSEflags()
```

# **Arguments**

new

A list which must have elements COMPILED and DUP.

# **Examples**

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

**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)
read.int(prmt)
svd.criteria(sv)
criteria.table.heading()
criteria.table.legend()
criteria.table.nheading()
DSE.ar(data, ...)
```

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Print Version Information

# **Description**

Print version information.

### Usage

```
DSEversion()
```

### **Examples**

```
DSEversion()
```

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.

# See Also

```
stability
```

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# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
McMillanDegree(model)</pre>
```

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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
   z <- MittnikReducedModels(toSS(estVARXls(eg1.DSE.data.diff)))</pre>
```

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MittnikReduction Balance and Reduce a Model		
	MittnikReduction	Balance and Reduce a Model

### **Description**

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

#### **Usage**

# 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.

Polynomials Polynomials

#### References

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

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.

Mittnik, S. (1989), Multivariate Time Series Analysis With State Space Models. *Computers Math Appl.* **17**, 1189–1201.

Mittnik, S. (1990), Macroeconomic Forecasting Experience With Balance State Space Models. *International Journal Of Forecasting*, **6**, 337–348.

Mittnik, S. (1990), Forecasting With Balanced State Space Representations of Multivariate Distributed Lag Models. *J. of Forecasting*, **9**, 207–218.

#### See Also

estVARXls bft balanceMittnik informationTests informationTestsCalculations

# **Examples**

```
if(is.R()) data("egJofF.1dec93.data", package="dse1")
model <- toSS(estVARXls(egJofF.1dec93.data))
newmodel <-MittnikReduction(model, criterion="taic")</pre>
```

Polynomials

Polynomial Utilities

# **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.

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# 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**

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

Riccati

Riccati Equation

# **Description**

Solve a Matrix Riccati Equation

# Usage

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

# Arguments

A A matrix.

B A matrix.

fuzz The tolerance used for testing convergence.
iterative If TRUE an iterative solution technique is used.

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#### **Details**

Solve Riccati equation P = APA' + B by eigenvalue decompostion 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

SS

State Space Models

# **Description**

Construct a

# Usage

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#### **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.
zO	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)).
PO	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 matices 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:

$$\begin{split} z(t) = &Fz(t-1) + Gu(t) + Qe(t) \\ y(t) = &Hz(t) + Rw(t) \\ \text{or the innovations model:} \\ z(t) = &Fz(t-1) + Gu(t) + Kw(t-1) \\ y(t) = &Hz(t) + w(t) \end{split}$$

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 1.ss, state is a time series matrix corresponding to z.
- z0 An initial value for z can be specified as z0.
- PO An initial one step ahead state tracking error (for non-innovations models) can be specified as PO.
- 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.

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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, an non-innovations form is harder to identify than an innovations form. Non-innovations form would typically be choosen 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 1.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</pre>
```

TSdata

Construct TSdata time series object

# **Description**

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

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#### 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 'TSestModel':
TSdata(data, ...)
is.TSdata(obj)
as.TSdata(d)
```

# **Arguments**

data	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.
input	a matrix of time series data.
output	a matrix of time series data.
	arguments to be passed to other methods.
obj	an object.
d	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 is 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)</pre>
```

TSdata.object time series data object

# **Description**

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

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#### 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 library 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 TSestModel.object TSPADIdata
```

TSestModel

Estimated Time Series Model

# **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**

obj

in the first usage an object from which a TSestModel object can be extracted (or constructed).

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#### **Details**

The TSestModel class of objects are generated by estimation methods. See, for example, estVARX1s. 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)| sampleT].
- track The smoothed estimate of the state tracking error P(t|sampleT). This is NULL for innovations models.

# See Also

```
estVARX1s, 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)
```

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#### **Arguments**

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, 1, 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 library.

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 library are designed for estimating and converting among various representations of these types of models.

### See Also

ARMA, SS, TSestModel, TSdata

TSrestrictedModel Restrict Model Coefficients

# **Description**

Put restrictions on a "TSmodel".

#### **Usage**

# **Arguments**

model An object containing an object of class TSmodel (unrestricted).

coefficients coefficients of the restricted model.

restriction function that maps the coefficients of the restricted model into the coefficients of the unrestricted model.

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#### **Details**

### BEWARE THIS CLASS OF MODELS IS EXPERIMENTAL AND MAY CHANGE.

This is the constructor for a "TSrestrictedModel", "TSmodel""

The argument model should specify a TSmodel which is not a TSrestrictedModel.

This constructor defines a new class of model TSrestrictedModel, which is composed of an unrestricted TSmodel, a new set of coefficients, and a function called restriction which maps the new set of coefficients into the unrestricted TSmodel. So, if mod1 is a TSrestrictedModel, then mod1\$restriction(mod1\$TSmodel, mod1\$coefficients) should return an unrestricted TSmodel with it coefficients specified. Methods like 1.TSrestrictedModel sets the arrays using the set of coefficients for the TSrestrictedModel and restrictions/equalities defined in the function restriction.

It is possible the class is misnamed. It should be possible to construct some extentions of TSmodels by definining restrictions on an enlarged model (but I have not played with this yet).

### See Also

```
TSmodel, ARMA, SS, TSestModel, TSdata
```

### **Examples**

```
rngValue10 <- list(seed=10, kind="Mersenne-Twister", normal.kind="Inversion")</pre>
  example 1
z \leftarrow ARMA(A=c(1, 0.3), B=1)
mod1 <- TSrestrictedModel(z,</pre>
            coefficients=coef(z),
            restriction=function(m, AllCoef){setArrays(m, 2*AllCoef)})
mod1
  example 2
mod2 <- TSrestrictedModel(z, coefficients=c(3, coef(z)),</pre>
       restriction=function(m, AllCoef){ setArrays(m, AllCoef[1]*AllCoef[-1])})
mod2
  example 3
z \leftarrow toss(ARMA(A=c(1, 0.3, 0.1), B=1))
mod3 <- TSrestrictedModel(z, coefficients=c(2,.3, 4, coef(z)),</pre>
       restriction=function(m, AllCoef){
                       mm <- setArrays(m, AllCoef[-(1:3)])</pre>
                       P0 \leftarrow matrix(0,2,2)
                        P0[,1] <- AllCoef[1:2]
                       P0[,2] <- AllCoef[2:3]
                       mm$P0 <- P0
                        setTSmodelParameters(mm)
                        })
mod3
  example 4
  Starting P0 ("big k") symmetric with off diagonal element smaller than diag.
P0 <- matrix(1e6,4,4)
```

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```
diag(P0 )<- 1e7
# lower triangle will be parameters
P0[outer(1:4, 1:4, ">=")]
length(P0[outer(1:4, 1:4, ">=")]) # number of parameters
Hloadings <- t(matrix(c(</pre>
    8.8, 5.2,
   23.8, -12.6,
   5.2, -2.0,
   36.8, 16.9,
   -2.8, 31.0,
   2.6, 47.6), 2,6))
z <- SS(F=t(matrix(c(</pre>
                  0.8, 0.04, 0.2, 0,
                  0.2, 0.5, 0, -0.3,
                    1, 0, 0, -0.2,
                    0,
                         1,
                               0, 0), c(4,4)),
               H=cbind(Hloadings, matrix(0,6,2)),
               Q=diag(c(1, 1, 0, 0), 4),
               R=diag(1,6),
               z0=c(10, 20, 30, 40),
               P0=NULL
# The restriction constructions PO from the lower triangle
mod4 <- TSrestrictedModel(z,</pre>
       coefficients=c(P0[outer(1:4, 1:4, ">=")], coef(z)),
       restriction=function(m, AllCoef){
                      mm <- setArrays(m, AllCoef[-(1:10)])</pre>
                      P0 <- matrix(0,4,4)
                      P0[outer(1:4, 1:4, ">=")] <- AllCoef[1:10]
                      P0 < - P0 + t(P0)
                      diag(P0) \leftarrow diag(P0)/2
                      mm$P0 <- P0
                       setTSmodelParameters(mm)
                       })
mod4
z <- simulate(SS(F=t(matrix(c(</pre>
                  0.8, 0.04, 0.2, 0,
                  0.2, 0.5, 0, -0.3,
1, 0, 0, -0.2,
                                0, 0 ), c(4,4))),
                    Ο,
                          1,
               H=cbind(Hloadings, matrix(0,6,2)),
               Q=diag(c(1, 1, 0, 0), 4),
               R=diag(1,6),
               z0=c(10, 20, 30, 40),
               P0=diag(c(10, 10, 10, 10))),
               rng=rngValue10)
state.sim <- z$state # for comparison below
y.sim <- outputData(z) # simulated indicators</pre>
coef(mod4)
coef(l(mod4, TSdata(output=y.sim)))
```

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```
summary(l(mod4, TSdata(output=y.sim)))
zz <- smoother(l(mod4, TSdata(output=y.sim)))</pre>
summary(zz)
tfplot(state.sim, state(zz))
tfplot(state.sim,state(zz, smoother=TRUE))
est.mod4 <- estMaxLik(mod4, TSdata(output=y.sim),</pre>
        algorithm.args=list(method="BFGS", upper=Inf, lower=-Inf, hessian=TRUE,
                         control=list(maxit=10000))
summary(est.mod4)
sest.mod4 <- smoother(est.mod4)</pre>
summary(sest.mod4)
tfplot(sest.mod4, graphs.per.page=3)
tfplot(state.sim, state(sest.mod4, smoother=TRUE))
tfplot(state.sim, state(sest.mod4, filter=TRUE))
coef(sest.mod4)
coef(mod4)
```

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.

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# 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**

v An object containing a TSmodel.

pch Character to use for plotting.

fuzz 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.

# See Also

```
plot.roots
```

# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
plot(roots(model))
addPlotRoots(toSS(model))</pre>
```

balanceMittnik 23

balanceMittnik	Balance a state space model	
----------------	-----------------------------	--

### **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  $p \times (m+p)$  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  $p \times p = 1$ .

See also MittnikReduction and references.

#### Value

A state space model in a TSestModel object.

#### References

See references for MittnikReduction.

### See Also

```
estVARXls, estVARXar MittnikReduction
```

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#### **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
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)</pre>
```

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 subsample 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.estBlackBox2estBlackBox3estBlackBox4informationTestsCalculations

# Examples

```
 if (is.R()) \ data("eg1.DSE.data.diff", package="dse1") \\ models <- list(estVARXls(eg1.DSE.data.diff), estVARXar(eg1.DSE.data.diff)) \\ z <- bestTSestModel(models)
```

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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.

# See Also

```
checkBalanceMittnik MittnikReduction
```

# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
checkBalance(model)</pre>
```

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.

### See Also

```
checkBalance MittnikReduction
```

### **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
checkBalanceMittnik(model)</pre>
```

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)
```

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# **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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
checkConsistentDimensions(model)</pre>
```

checkResiduals

**Autocorrelations Diagnostics** 

# **Description**

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

# Usage

# **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.

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```
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.

# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
checkResiduals(model)</pre>
```

coef.TSmodel

Extract 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, ...)
## S3 method for class 'TSrestrictedModel':
coef(object, ...)
coef(object) <- value
## Default S3 method:
coef(object) <- value</pre>
```

# **Arguments**

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

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### Value

A vector of parameter values.

# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
coef(model)
coef(model) <- 0.1 + coef(model)</pre>
```

combine

Combine two objects.

# **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

tbind, combine.TSdata, combine.forecastCov

# **Examples**

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combine. TSdata

Combine series from two TSdata objects.

# **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**

diffLog

Calculate the difference of log data

# **Description**

Calculate the difference from lag periods prior for log of data.

# Usage

```
diffLog(x, lag=1, base = exp(1), ...)
```

# Arguments

x	A time series.
lag	The difference is calculated relative to lag periods prior.
base	Base to use when calculating logrithms.
	(further arguments, currently disregarded).

eg1.DSE.data 31

#### Value

A time series of the difference relative to lag periods prior for the log of the data. lag data points are lost from the beginning of the series. Negative values will result in NAs.

### **Examples**

```
if(is.R()) data("eg1.DSE.data", package="dsel")
z <- diffLog(outputData(eg1.DSE.data))</pre>
```

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 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", "GDPl2" and "CPI". GDPl2 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." Working Paper 93-4, Bank of Canada. Also available at <a href="https://www.bank-banque-canada.ca/pgilbert">www.bank-banque-canada.ca/pgilbert</a>.

### See Also

**TSdata** 

32 egJofF.1dec93.data

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

estBlackBox

Estimate a TSmodel

# **Description**

Estimate a TSmodel.

# Usage

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

### **Arguments**

data Data in an object of class TSdata.

. . . Optional arguments depent 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**

```
if(is.R()) data("egJofF.1dec93.data", package="dse1")
goodmodel <- estBlackBox(egJofF.1dec93.data)</pre>
```

estBlackBox1

Estimate a TSmodel

# Description

Estimate a TSmodel.

# Usage

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

#### **Arguments**

data Data in an object of class TSdata.
estimation Initial estimation method to be used.

reduction Reduction method to be used.

 ${\tt criterion} \qquad {\tt Criterion} \ {\tt to} \ {\tt be} \ {\tt used} \ {\tt for} \ {\tt model} \ {\tt selection}. \ {\tt see} \ {\tt informationTestsCalculations}.$ 

trend logical indicating if a trend should be estimated.

subtract.means

logical indicating if the mean should be subtracted from data before estimation.

verbose logical indicating if information should be printed during estimation.

max.laq integer indicating the maximum number of lags to consider.

#### Value

A state space model in an object of class TSestModel.

#### See Also

informationTestsCalculations

# **Examples**

```
if(is.R()) data("egJofF.1dec93.data", package="dse1")
goodmodel <- estBlackBox1(egJofF.1dec93.data)</pre>
```

estBlackBox2

Estimate a TSmodel

### **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

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.

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 TRUE 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 proce-

dures is printed.

The number of lags to include in the VAR estimation. max.lag

### **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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
z <- estBlackBox2(eg1.DSE.data.diff)</pre>
```

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.

If TRUE then additional information from the estimation and reduction proce-

The number of lags to include in the VAR estimation.

sample.start The starting point to use for calculating information criteria.

dures is printed.

#### **Details**

verbose

max.lag

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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
z <- estBlackBox3(eg1.DSE.data.diff)</pre>
```

estBlackBox4 37

### **Description**

Estimate a TSmodel with Brute Force Technique.

# Usage

# **Arguments**

A TSdata object. data estimation a character string indicating the estimation method to use. lag.weight weighting to apply to lagged observations. variable.weights weighting to apply to series if estimation method is estWtVariables. reduction character string indicating reduction procedure to use. criterion criterion to be used for model selection. see informationTestsCalculations. if TRUE include a trend in the model. trend 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. if TRUE the data is transformed so that all variables have the same variance. standardize verbose if TRUE then additional information from the estimation and reduction procedures is printed. VAR estimation is done for each lag up to max.lag. max.lag sample.start the starting point to use for calculating information criteria in the final selection. logical indicating if warning messages should be suppressed. warn 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).

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### 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**

```
 \begin{array}{lll} \mbox{if(is.R()) data("eg1.DSE.data.diff", package="dse1")} \\ \mbox{z <- } \mbox{bft(eg1.DSE.data.diff)} \\ \end{array}
```

estMaxLik

Maximum Likelihood Estimation

### **Description**

Maximum likelihood estimation.

estMaxLik(obj1, obj2, ...)

# Usage

# **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.
```

estSSMittnik 39

#### **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 minimimize 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 alogrithm). 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\$converceCode, 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.

### See Also

```
optim, nlm, estVARXls, bft, TSmodel, 1, SS, ARMA, fixConstants
```

# **Examples**

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)
```

40 estSSfromVARX

### **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 Mittnik p1195 Comp.Math Appl.v17,1989. 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 MittnikReduction 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 MittnikReduction.

# See Also

MittnikReduction estVARXls bft

## **Examples**

```
if(is.R()) data("egJofF.1dec93.data", package="dse1")
## Not run: model <- estSSMittnik(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 <- estSSMittnik(egJofF.1dec93.data, n=3)</pre>
```

estSSfromVARX

Estimate a state space TSmodel using VAR estimation

# **Description**

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

# Usage

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

estVARXar 41

## **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 arguements to estVARXIs

### **Details**

This function uses the functions estVARXIs 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 <a href="https://www.bank-banque-canada.ca/pgilbert">www.bank-banque-canada.ca/pgilbert</a>.
```

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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
model <-estSSfromVARX(eg1.DSE.data.diff)</pre>
```

estVARXar

Estimate a VAR TSmodel

# 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.
```

42 estVARXar

unstandardize

If TRUE and standardize is TRUE then the returned model is adjusted to corre-

spond 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 exogeneous 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 conventon 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 <a href="https://www.bank-banque-canada.ca/pgilbert">www.bank-banque-canada.ca/pgilbert</a>

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 estVARXls estMaxLik ar DSE.ar
```

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
model <- estVARXar(eg1.DSE.data.diff)</pre>
```

estVARXIs 43

### **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 corre-

spond 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 TSestModel object containing a TSmodel object which is a VAR model.

44 estWtVariables

#### References

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

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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
model <- estVARXls(eg1.DSE.data.diff)</pre>
```

estWtVariables

Weighted Estimation

## **Description**

estWtVariables

### Usage

# **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. Estmation 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 45

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**

model an object of class TSmodel.

fuzz absolute difference to be considered equivalent.

constants NULL or a list of logical arrays.

### **Details**

If constants is not NULL then parameters within fuzz of 0.0 or 1.0 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

46 gmap

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

model

An object of class TSmodel.

### Value

An SS TSmodel object.

### See Also

fixConstants

# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
model <- toSS(estVARXls(eg1.DSE.data.diff))
model <- fixF(model)</pre>
```

gmap

Basis Transformation of a Model.

# **Description**

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

# Usage

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

# Arguments

g An invertible matrix

model 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.

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### Value

An equivalent model transformed using g.

# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
gmap(2, model)</pre>
```

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.

# See Also

```
informationTestsCalculations
```

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model1 <- estVARXls(eg1.DSE.data.diff)
model2 <- estVARXar(eg1.DSE.data.diff)
informationTests(model1, model2)</pre>
```

informationTestsCalculations

Calculate selection criteria

# **Description**

Calculates several model selection criteria.

### Usage

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

# **Arguments**

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

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
informationTestsCalculations(model)</pre>
```

inputData 49

inputData	TSdata Series
-----------	---------------

# **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</pre>
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
```

```
if(is.R()) data("eg1.DSE.data", package="dse1")
outputData(eg1.DSE.data)
```

1.ARMA

## **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	3
	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 l.ARMA directly. l.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.

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### See Also

```
ARMA 1, 1.SS TSmodel TSestModel.object
```

### **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- TSmodel(estVARXls(eg1.DSE.data.diff))
evaluated.model <- l(model,eg1.DSE.data.diff)</pre>
```

1

Evaluate a TSmodel

### **Description**

Evaluate a model with data.

# Usage

```
1(obj1, obj2, ...)
## S3 method for class 'TSdata':
1(obj1, obj2, ...)
## S3 method for class 'TSestModel':
1(obj1, obj2, ...)
## S3 method for class 'TSrestrictedModel':
1(obj1, obj2, result=NULL, ...)
```

# Arguments

```
obj1 a TSmodel, TSdata, or TSestModel object.
obj2 a TSmodel or TSdata object.
result specifies part of TSestModel object to returned
arguments to be passed to other methods.
```

### **Details**

For state space models 1. SS is called and for ARMA models 1. 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

```
1.SS, 1.ARMA
```

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- toSS(TSmodel(estVARXls(eg1.DSE.data.diff)))
evaluated.model <- l(model, eg1.DSE.data.diff)</pre>
```

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1.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	3
	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 $tack$ 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 $\ensuremath{\mathrm{S/R}}$ 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).

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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 return (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) 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 11.ARMA TSmodel TSestModel.object state smoother
```

```
if(is.R()) 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))</pre>
```

54 makeTSnoise

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

# **Arguments**

sampleT	an integer indicating the number of periods.	
p	an integer indicating the number of series.	
lags	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.	
noise	Noise can be supplied. Otherwise it will be generated. If supplied it should be a list as described below under returned value.	
Cov	The covariance of the noise process. If this is specified then sd is ignored. A vector or scalar is treated as a diagonal matrix. For an object of class TSestModel, if neither Cov nor sd are specified, then Cov is set to the estimated covariance (model\$estimates\$cov).	
sd	The standard deviation of the noise. This can be a vector.	
noise.model	A TSmodel to be used for generating noise (not yet supported by SS methods).	
noise.baseline		
	a constant or matrix to be added to noise. Alternately this can be a vector of	
	length p, each value of which is treated as a constant to add to the coresponding noise series.	
rng	length p, each value of which is treated as a constant to add to the coresponding	
rng tf	length p, each value of which is treated as a constant to add to the coresponding noise series.	
_	length p, each value of which is treated as a constant to add to the coresponding noise series.  The random number generator information needed to regenerate a simulation.	

# Value

A time series matrix.

markovParms 55

markovParms

Markov Parameters

# **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 [M0|...|Mi] giving the Markov parameters Mi, i+1 = blocks where each Mi 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 Mittnik p1190. If 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 MittnikReduction.

### See Also

SVDbalanceMittnik

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
markovParms(model)</pre>
```

56 nseriesInput

nseriesInput

Number of Series in in Input or Output

### **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

 $Object\ of\ class\ TS data,\ TS model\ or\ TS est Model.$ 

# Value

An integer indicating the number of series.

### See Also

```
seriesNamesInput seriesNamesOutput
```

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

observability 57

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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
observability(model)</pre>
```

percentChange

Calculate percent change

# **Description**

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

58 percentChange

### Usage

```
percentChange(obj, ...)
## Default S3 method:
percentChange(obj, base=NULL, lag=1, cumulate=FALSE, e=FALSE, ...)
## 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**

obj	An object of class TSdata or TSestModel, a time series matrix, a matrix with columns corresponding to series (which are treated individually), or a list of one of these kinds of objects. (called m below)
е	If e is TRUE the exponent of the series is used (after cumulating if cumulate is TRUE). e can be a logical vector with elements corresponding to columns of m.
base	If base is provided it is treated as the first period value (that is, prior to differencing). It is prefixed to the m prior to cumulating. It should be a vector of length dim(m)[2]. (If e is TRUE then base should be log of the original data).
lag	integer indicating the number of periods relative to which the percent change should be calculated.
cumulate	logical indicating if the series should be cumulated before the percent change is calculated.
	arguments passed to other methods.

# Details

Calculate the percent change relative to the data lag periods prior. obj should be a matrix or vector.

If codecumulate is TRUE then the data is cumulated first. cumulate can be a logical vector with elements corresponding to columns of m.

For a TSmodel the percent change calculation is done with input and output and the result is an object of class TSdata.

For a TSestModel the percent change calculation is done with estimates pred and the result is an object of class TSdata

### 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

```
ytoypc
```

```
if(is.R()) data("eg1.DSE.data", package="dse1")
z <- percentChange(outputData(eg1.DSE.data))</pre>
```

periods.TSdata 59

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**

```
x a time series object.... (further arguments, currently disregarded).
```

# See Also

```
periods, start, end, frequency
```

periodsInput

TSdata Periods

# **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)
```

60 periodsInput

```
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 periods(inputData(x)) requires a

plot.roots 61

database retrieval which may be fairly slow, while the number of periods may be available much more quickly.

# Value

Depends.

# **Examples**

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

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

х	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.

# See Also

```
{\tt addPlotRoots}\ {\tt roots}\ {\tt stability}\ {\tt McMillanDegree}
```

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
model <- estVARXls(eg1.DSE.data.diff)
plot(roots(model))</pre>
```

62 print.TSestModel

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
```

# **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=le-10, ...)
## S3 method for class 'TSestModel':
print(x, ...)
## S3 method for class 'TSrestrictedModel':
print(x, ...)
```

# Arguments

Х	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.

reachability 63

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.

### Note

BUG: digits cannot be controlled for some numbers (e.g.- 1.0 is printed as 0.999999999)

### See Also

```
print summary
```

# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
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)</pre>
```

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

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.

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### See Also

```
observability, stability roots McMillanDegree
```

# **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
reachability(model)</pre>
```

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**

pred A matrix with columns representing time series.

data A matrix with columns representing time series.

sampleT An integer indicating the sample to use.

warn 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 sample T. pred and sample T 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

```
residual Stats(\texttt{matrix}(\texttt{rnorm}(200),\ 100,2),\ \texttt{NULL})\ \#\ \texttt{but\ typically\ used\ for\ a\ residual}
```

residuals.TSestModel 65

```
residuals.TSestModel
```

Calculate the residuals for an object

# **Description**

Calculate the residuals for an object.

### Usage

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

# Arguments

```
object an object.... additional arguments passed to stats::residuals.
```

### **Details**

Calculates the residuals (prediction minus data).

# Value

A time series matrix.

# Author(s)

Paul Gilbert

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, ...)
```

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### **Arguments**

obj An object of class TSmodel.

fuzz If non-zero then roots within fuzz distance are considered equal.

randomize Randomly arrange complex pairs of roots so the one with the positive imaginary

part is not always first (so random experiments are not biased).

warn If FALSE then warnings about unit roots added for TREND are not printed.

by .poly If TRUE then roots are calculated by expanding the determinant of the A poly-

nomial. 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 polyroot.

... 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 <a href="https://www.bank-banque-canada.ca/pgilbert">www.bank-banque-canada.ca/pgilbert</a>

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

## See Also

```
stability, McMillanDegree
```

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
model <- estVARXls(eg1.DSE.data.diff)
roots(model)</pre>
```

scale.TSdata 67

scale.TSdata

Scale Methods for TS objects

# **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**

scale

x TSdata, TSmodel or an object containing these.
center to match generic arguments, not currently used.

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. scale\$input 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

68 seriesNames.TSdata

### **Examples**

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 method for class 'TSdata':
seriesNames(x) <- value
## S3 method for class 'TSmodel':
seriesNames(x) <- value
## S3 method for class 'TSestModel':
seriesNames(x) <- value</pre>
```

# **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

seriesNamesInput 69

```
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</pre>
seriesNamesOutput(x) <- value</pre>
```

# **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
```

```
if(is.R()) data("eg1.DSE.data", package="dse1")
seriesNamesOutput(eg1.DSE.data)
```

70 setTSmodelParameters

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)
## S3 method for class 'ARMA':
setArrays(model, coefficients=NULL)
## S3 method for class 'SS':
setArrays(model, coefficients=NULL)
## S3 method for class 'TSestModel':
setArrays(model, coefficients=NULL)
## S3 method for class 'TSrestrictedModel':
setArrays(model, coefficients=coef(model))
```

### **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

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## **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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
model <- setTSmodelParameters(model)</pre>
```

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.

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y0, input0 Lagged values prior to t=1 for y and u, in reverse order so y0[1,] and input0[1,]correspond

to t=0. These arguments are not implemented for state space models. If not

specified initial values are set to zero.

noise Noise can be supplied. Otherwise it will be generated. If supplied it should be a

list as described below in details.

Cov The covariance of the noise process. If this is specified then sd is ignored. A vec-

tor or scalar is treated as a diagonal matrix. For an object of class TSestModel, if neither Cov nor sd are specified, then Cov is set to the estimated covariance

(model\$estimates\$cov).

sd The standard deviation of the noise. This can be a vector.

noise.model A TSmodel to be used for generating noise (not yet supported by SS methods).

The random number generator information needed to regenerate a simulation.

compiled Specifies the compiled version of the code should be used (instead of the S code

version).

. . . 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 compatability 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,

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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 \leftarrow ARMA(A=array(c(1,-.25,-.05), c(3,1,1)), B=array(1,c(1,1,1)))
   <- 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)</pre>
     \leftarrow array(c(0.5,0,0,0.2),c(1,2,2))
VARX <- ARMA(A=AR, B=diag(1,2), C=C)</pre>
simData <- simulate(VARX, sampleT=150, input=matrix(rnorm(300),150,2))</pre>
     \leftarrow 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)</pre>
ARMAX <- ARMA(A=AR, B=MA, C=C)
simData <- simulate(ARMAX, sampleT=150, input=matrix(rnorm(300),150,2))</pre>
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)</pre>
simData <- simulate(model)</pre>
ss \leftarrow 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)</pre>
testEqual(simData, simulate(ss))
testEqual(simData, simulate(ss, rng=getRNG(simData)))
simData2 <- simulate(ss,</pre>
    noise=list(w=matrix(runif(300), 150,2), w0=runif(2)))
simData3 <- simulate(ss, noise=matrix(runif(400), 200,2))</pre>
```

smoother

Evaluate a smoother with a TSmodel

## **Description**

Evaluate a state space model.

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

## **Arguments**

model 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.

data A TSdata object.

compiled 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 1.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) 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 argument).

## 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, 1, SS 1.SS TSmodel TSestModel.object
```

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#### **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))</pre>
```

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.

## See Also

```
McMillanDegree
```

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## **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
stability(model)</pre>
```

standardize

standardize data

## **Description**

Convert data to mean zero, standard deviation one. This does not remove cross correlations between series.

# Usage

```
standardize(ser)
```

#### **Arguments**

ser

A vector time matrix.

#### Value

A time series matrix.

## **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
z <- standardize(outputData(eg1.DSE.data.diff))</pre>
```

state

Extract State

## **Description**

Extract state information from estimated SS model.

# Usage

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

# Arguments

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.

.

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#### **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, 1.SS
```

sumSqerror

Calculate sum of squared prediction errors

## **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

```
11.SS 1.ARMA
```

#### **Examples**

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```
summary. TSdata Specific Methods for Summary
```

## **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**

```
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
```

```
testEqual.ARMA Specific Methods for Testing Equality
```

#### **Description**

See the generic function description.

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#### 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.

```
## 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)
```

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#### **Arguments**

```
object to be plotted.
                   additional objects to be plotted.
                   start of plot.
start
                   end of plot.
end
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)
                   string to use for plot title (passed to plot).
Title
                   string to use for x label (passed to plot).
xlab
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.

```
## S3 method for class 'TSdata':
tframed(x, tf=NULL, names=NULL)
## S3 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, ...)</pre>
```

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## **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.

toSS toSS

#### 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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- toSS(estVARXls(eg1.DSE.data.diff))
model <- toARMA(model)</pre>
```

toSS

Convert to State Space Model

## **Description**

Convert a model to state space form.

```
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, ...)
```

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# 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 identitity 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 Mittnik. (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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dsel")
model <- estVARXls(eg1.DSE.data.diff)
model <- toSS(model)</pre>
```

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 1 (1.SS or 1.ARMA). This assumes that the noise processes in the arbitrary SS representation are white and uncorrelated.

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

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## **Arguments**

model An object of class TSmodel.

Om a matrix to be used as the measurement noise covariance. If Om is not supplied and model is of class TSestModel then model\$estimates\$cov is used. Otherwise, Om is set to the identity matrix.

... 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**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)
model <- toSSChol(model)</pre>
```

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.

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#### Value

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

#### See Also

```
toSSinnov
```

## **Examples**

```
if(is.R()) data("eg1.DSE.data.diff", package="dse1")
model <- estVARXls(eg1.DSE.data.diff)</pre>
```

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 an 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**

```
if(is.R()) 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)</pre>
```

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ytoypc

Convert to year to year percent change

# **Description**

Convert level data to year over year percent change.

# Usage

```
ytoypc(ser)
```

## **Arguments**

ser

A time series.

#### Value

A vector time series of the year over year percent change. This uses percentChange with lag=frequency(ser).

#### See Also

```
percentChange
```

# **Examples**

```
if(is.R()) data("eg1.DSE.data", package="dsel")
z <- ytoypc(outputData(eg1.DSE.data))</pre>
```

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