

Package ‘kalmanfilter’

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Type Package

Title Kalman Filter

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Description 'Rcpp' implementation of the multivariate Kalman filter for state space models that can handle missing values and exogenous data in the observation and state equations. Kim, Chang-Jin and Charles R. Nelson (1999) ``State-Space Models with Regime Switching: Classical and Gibbs-Sampling Approaches with Applications'' <<http://econ.korea.ac.kr/cjkim/doi:10.7551/mitpress/6444.001.0001>><<http://econ.korea.ac.kr/~{}cjkim/>>.

License GPL (>= 2)

Imports Rcpp (>= 1.0.9)

LinkingTo Rcpp, RcppArmadillo

RoxygenNote 7.2.1

Suggests data.table (>= 1.14.2), maxLik (>= 1.5-2), ggplot2 (>= 3.3.6), gridExtra (>= 2.3), knitr, rmarkdown, testthat

VignetteBuilder knitr

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NeedsCompilation yes

Author Alex Hubbard [aut, cre]

Maintainer Alex Hubbard <hubbard.alex@gmail.com>

Depends R (>= 3.5.0)

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contains	<i>Check if list contains a name</i>
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Description

Check if list contains a name

Usage

`contains(s, L)`

Arguments

s	a string name
L	a list object

Value

boolean

gen_inv	<i>Generalized matrix inverse</i>
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Description

Generalized matrix inverse

Usage

`gen_inv(m)`

Arguments

m	matrix
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Value

matrix inverse of m

kalmanfilter*Kalman Filter*

Description

kalmanfilter Rcpp implementation of the multivariate Kalman filter for state space models that can handle missing values and exogenous data in the observation and state equations. See the package vignette using `browseVignettes("kalmanfilter")` to view it in your browser.

Author(s)

Alex Hubbard

kalman_filter

Kalman Filter

Description

Kalman Filter

Usage

```
kalman_filter(ssm, yt, Xo = NULL, Xs = NULL, weight = NULL, smooth = FALSE)
```

Arguments

ssm	list describing the state space model, must include names B0 - N_b x 1 matrix, initial guess for the unobserved components P0 - N_b x N_b matrix, initial guess for the covariance matrix of the unobserved components Dm - N_b x 1 matrix, constant matrix for the state equation Am - N_y x 1 matrix, constant matrix for the observation equation Fm - N_b X p matrix, state transition matrix Hm - N_y x N_b matrix, observation matrix Qm - N_b x N_b matrix, state error covariance matrix Rm - N_y x N_y matrix, state error covariance matrix betaO - N_y x N_o matrix, coefficient matrix for the observation exogenous data betaS - N_b x N_s matrix, coefficient matrix for the state exogenous data
yt	N x T matrix of data
Xo	N_o x T matrix of exogenous observation data
Xs	N_s x T matrix of exogenous state
weight	column matrix of weights, T x 1
smooth	boolean indication whether to run the backwards smoother

Value

list of matrices and cubes output by the Kalman filter

Examples

```
#Nelson-Siegel dynamic factor yield curve
library(kalmanfilter)
library(data.table)
data(treasuries)
tau = unique(treasuries$maturity)

#Set up the state space model
ssm = list()
ssm[["Fm"]] = rbind(c(0.9720, -0.0209, -0.0061),
                     c(0.1009 , 0.8189, -0.1446),
                     c(-0.1226, 0.0192, 0.8808))
ssm[["Dm"]] = matrix(c(0.1234, -0.2285, 0.2020), nrow = nrow(ssm[["Fm"]]), ncol = 1)
ssm[["Qm"]] = rbind(c(0.1017, 0.0937, 0.0303),
                     c(0.0937, 0.2267, 0.0351),
                     c(0.0303, 0.0351, 0.7964))
ssm[["Hm"]] = cbind(rep(1, 11),
                     -(1 - exp(-tau*0.0423))/(tau*0.0423),
                     (1 - exp(-tau*0.0423))/(tau*0.0423) - exp(-tau*0.0423))
ssm[["Am"]] = matrix(0, nrow = length(tau), ncol = 1)
ssm[["Rm"]] = diag(c(0.0087, 0, 0.0145, 0.0233, 0.0176, 0.0073,
                     0, 0.0016, 0.0035, 0.0207, 0.0210))
ssm[["B0"]] = matrix(c(5.9030, -0.7090, 0.8690), nrow = nrow(ssm[["Fm"]]), ncol = 1)
ssm[["P0"]] = diag(rep(0.0001, nrow(ssm[["Fm"]])))

#Convert to an NxT matrix
yt = dcast(treasuries, "date ~ maturity", value.var = "value")
yt = t(yt[, 2:ncol(yt)])
kf = kalman_filter(ssm, yt, smooth = TRUE)
```

Rginv

R's implementation of the Moore-Penrose pseudo matrix inverse

Description

R's implementation of the Moore-Penrose pseudo matrix inverse

Usage

`Rginv(m)`

Arguments

`m` matrix

Value

matrix inverse of `m`

sw_dcf*Stock and Watson Dynamic Common Factor Data Set*

Description

Stock and Watson Dynamic Common Factor Data Set

Usage

```
data(sw_dcf)
```

Format

data.table with columns DATE, VARIABLE, VALUE, and MATURITY The data is monthly frequency with variables ip (industrial production), gmyxpg (total personal income less transfer payments in 1987 dollars), mtq (total manufacturing and trade sales in 1987 dollars), lpnag (employees on nonagricultural payrolls), and dcoinc (the coincident economic indicator)

Source

Kim, Chang-Jin and Charles R. Nelson (1999) "State-Space Models with Regime Switching: Classical and Gibbs-Sampling Approaches with Applications" <doi:10.7551/mitpress/6444.001.0001>http://econ.korea.ac.kr/~cjkim/switching.pdf

treasuries*Treasuries*

Description

Treasuries

Usage

```
data(treasuries)
```

Format

data.table with columns DATE, VARIABLE, VALUE, and MATURITY The data is quarterly frequency with variables DGS1MO, DGS3MO, DGS6MO, DGS1, DGS2, DGS3, DGS5, DGS7, DGS10, DGS20, and DGS30

Source

FRED

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