

Package ‘fnets’

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Type Package

Title Factor-Adjusted Network Estimation and Forecasting for High-Dimensional Time Series

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Description

Implements methods for network estimation and forecasting of high-dimensional time series exhibiting strong serial and cross-sectional correlations under a factor-adjusted vector autoregressive model.

See Barigozzi, Cho and Owens (2022) <[arXiv:2201.06110](#)> for further descriptions of FNETS methodology and

Owens, Cho and Barigozzi (2023) <<https://drive.google.com/file/d/1Rw-xgpijF8ZIBUzjIU9emr-ucAvAhKL4/view?usp=sharing>> accompanying the R package.

Depends R (>= 4.1.0)

Imports lpSolve,

parallel,

doParallel,

foreach,

MASS,

fields,

igraph,

RColorBrewer

License GPL (>= 3)

Encoding UTF-8

LazyData false

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Suggests testthat (>= 3.0.0)

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common.predict	<i>Forecasting the factor-driven common component</i>
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Description

Produces forecasts of the common component for a given forecasting horizon by estimating the best linear predictors

Usage

```
common.predict(object, x, h = 1, fc.restricted = TRUE, r = c("ic", "er"))
```

Arguments

object	fnets object
x	input time series matrix, with each row representing a variable
h	forecasting horizon
fc.restricted	whether to forecast using a restricted or unrestricted, blockwise VAR representation of the common component
r	number of restricted factors, or a string specifying the factor number selection method when <code>fc.restricted = TRUE</code> ; possible values are: <ul style="list-style-type: none"> • "ic" information criteria of Bai and Ng (2002) • "er" eigenvalue ratio

Value

a list containing	
is	in-sample estimator of the common component
fc	forecasts of the common component for a given forecasting horizon h
r	restricted factor number
h	forecast horizon

References

- Ahn, S. C. & Horenstein, A. R. (2013) Eigenvalue ratio test for the number of factors. *Econometrica*, 81(3), 1203–1227.
- Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series. *arXiv preprint arXiv:2201.06110*.
- Forni, M., Hallin, M., Lippi, M. & Reichlin, L. (2005). The generalized dynamic factor model: one-sided estimation and forecasting. *Journal of the American Statistical Association*, 100(471), 830–840.
- Forni, M., Hallin, M., Lippi, M. & Zaffaroni, P. (2017). Dynamic factor models with infinite-dimensional factor space: Asymptotic analysis. *Journal of Econometrics*, 199(1), 74–92.
- Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

Examples

```
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
x <- common$data + idio$data
out <- fnets(x, q = NULL, var.order = 1, var.method = "lasso",
do.lrpc = FALSE, var.args = list(n.cores = 2))
cpre <- common.predict(out, x, h = 1, r = NULL)
ipre <- idio.predict(out, x, cpre, h = 1)
```

factor.number

Factor number selection methods

Description

Methods to estimate the number of factor. When method = 'er', the factor number is estimated by maximising the ration of successive eigenvalues. When method = 'ic', the information criterion-methods discussed in Hallin and Liška (2007) (when fm.restricted = FALSE) and Alessi, Barigozzi and Capasso (2010) (when fm.restricted = TRUE) are implemented, with the information criterion called by ic.op = 5 recommended by default.

Usage

```
factor.number(
  x,
  fm.restricted = FALSE,
  method = c("ic", "er"),
  q.max = NULL,
  do.plot = FALSE,
  center = TRUE
)
```

Arguments

<code>x</code>	input time series matrix, with each row representing a variable
<code>fm.restricted</code>	whether to estimate the number of restricted or unrestricted factors
<code>method</code>	A string specifying the factor number selection method; possible values are: <ul style="list-style-type: none"> • "ic" information criteria-based methods of Alessi, Barigozzi & Capasso (2010) when <code>fm.restricted = TRUE</code> or Hallin and Liška (2007) when <code>fm.restricted = FALSE</code> • "er" eigenvalue ratio of Ahn and Horenstein (2013)
<code>q.max</code>	maximum number of factors; if <code>q.max = NULL</code> , a default value is selected as <code>min(50, floor(sqrt(min(dim(x)[2] - 1, dim(x)[1])))</code>
<code>do.plot</code>	whether to plot the information criteria values
<code>center</code>	whether to de-mean the input <code>x</code> row-wise

Details

For further details, see references.

Value

if `method = "ic"`, a vector containing minimisers of the six information criteria, otherwise, the maximiser of the eigenvalue ratio

References

- Ahn, S. C. & Horenstein, A. R. (2013) Eigenvalue ratio test for the number of factors. *Econometrica*, 81(3), 1203–1227.
- Alessi, L., Barigozzi, M., and Capasso, M. (2010) Improved penalization for determining the number of factors in approximate factor models. *Statistics & Probability Letters*, 80(23-24):1806–1813.
- Bai, J. & Ng, S. (2002) Determining the number of factors in approximate factor models. *Econometrica*. 70: 191-221.
- Hallin, M. & Liška, R. (2007) Determining the number of factors in the general dynamic factor model. *Journal of the American Statistical Association*, 102(478), 603–617.
- Owens, D., Cho, H. & Barigozzi, M. (2022) *fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling*

Examples

```
library(fnets)

set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
x <- common$data * apply(idio$data, 1, sd) / apply(common$data, 1, sd) + idio$data
```

```

h1 <- factor.number(x, fm.restricted = FALSE, do.plot = TRUE)
h1
library(fnets)

set.seed(123)
n <- 500
p <- 50
common <- sim.restricted(n, p)
idio <- sim.var(n, p)
x <- common$data * apply(idio$data, 1, sd) / apply(common$data, 1, sd) + idio$data

abc <- factor.number(x, fm.restricted = TRUE, do.plot = TRUE)
abc

er <- factor.number(x, method = "er", fm.restricted = TRUE, do.plot = TRUE)
er

```

fnets

Factor-adjusted network estimation

Description

Operating under a factor-adjusted vector autoregressive (VAR) model, the function estimates the spectral density and autocovariance matrices of the factor-driven common component and the idiosyncratic VAR process, the impulse response functions and common shocks for the common component, and VAR parameters, innovation covariance matrix and long-run partial correlations for the idiosyncratic component.

Usage

```

fnets(
  x,
  center = TRUE,
  fm.restricted = FALSE,
  q = c("ic", "er"),
  ic.op = NULL,
  kern.bw = NULL,
  common.args = list(factor.var.order = NULL, max.var.order = NULL, trunc.lags = 20,
    n.perm = 10),
  var.order = 1,
  var.method = c("lasso", "ds"),
  var.args = list(n.iter = NULL, n.cores = min(parallel::detectCores() - 1, 3)),
  do.threshold = FALSE,
  do.lrpc = TRUE,
  lrpc.adaptive = FALSE,
  tuning.args = list(tuning = c("cv", "bic"), n.folds = 1, penalty = NULL, path.length
    = 10, do.plot = FALSE)
)

```

Arguments

<code>x</code>	input time series matrix, with each row representing a variable
<code>center</code>	whether to de-mean the input <code>x</code> row-wise
<code>fm.restricted</code>	whether to estimate a restricted factor model using static PCA
<code>q</code>	<p>Either the number of factors or a string specifying the factor number selection method; possible values are:</p> <ul style="list-style-type: none"> • "ic" information criteria-based methods of Alessi, Barigozzi & Capasso (2010) when <code>fm.restricted = TRUE</code> or Hallin and Liška (2007) when <code>fm.restricted = FALSE</code> modifying Bai and Ng (2002) • "er" eigenvalue ratio of Ahn and Horenstein (2013) <p>see factor.number.</p>
<code>ic.op</code>	choice of the information criterion penalty, see factor.number for further details
<code>kern.bw</code>	a positive integer specifying the kernel bandwidth for dynamic PCA; by default, it is set to $\text{floor}(4 * (\text{dim}(x)[2] / \log(\text{dim}(x)[2]))^{(1/3)})$. When <code>fm.restricted = TRUE</code> , it is used to compute the number of lags for which autocovariance matrices are estimated
<code>common.args</code>	<p>a list specifying the tuning parameters required for estimating the impulse response functions and common shocks. It contains:</p> <ul style="list-style-type: none"> • <code>factor.var.order</code> order of the blockwise VAR representation of the common component. If <code>factor.var.order = NULL</code>, it is selected blockwise by Schwarz criterion • <code>max.var.order</code> maximum blockwise VAR order for the Schwarz criterion • <code>trunc.lags</code> truncation lag for impulse response function estimation • <code>n.perm</code> number of cross-sectional permutations involved in impulse response function estimation
<code>var.order</code>	order of the idiosyncratic VAR process; if a vector of integers is supplied, the order is chosen via tuning
<code>var.method</code>	<p>a string specifying the method to be adopted for idiosyncratic VAR process estimation; possible values are:</p> <ul style="list-style-type: none"> • "lasso" Lasso-type l1-regularised M-estimation • "ds" Dantzig Selector-type constrained l1-minimisation
<code>var.args</code>	<p>a list specifying the tuning parameters required for estimating the idiosyncratic VAR process. It contains:</p> <ul style="list-style-type: none"> • <code>n.iter</code> maximum number of descent steps, by default depends on <code>var.order</code>; applicable when <code>var.method = "lasso"</code> • <code>n.cores</code> number of cores to use for parallel computing, see makePSOCK-cluster; applicable when <code>var.method = "ds"</code>
<code>do.threshold</code>	whether to perform adaptive thresholding of all parameter estimators with threshold
<code>do.lrpc</code>	whether to estimate the long-run partial correlation
<code>lrpc.adaptive</code>	whether to use the adaptive estimation procedure

- `tuning.args` a list specifying arguments for tuning for selecting the tuning parameters involved in VAR parameter and (long-run) partial correlation matrix estimation. It contains:
- `tuning` a string specifying the selection procedure for `var.order` and `lambda`; possible values are:
 - `"cv"` cross validation
 - `"bic"` information criterion
 - `n.folds` if `tuning = "cv"`, positive integer number of folds
 - `penalty` if `tuning = "bic"`, penalty multiplier between 0 and 1; if `penalty = NULL`, it is set to $1/(1+\exp(\dim(x)[1]/\dim(x)[2]))$ by default
 - `path.length` positive integer number of regularisation parameter values to consider; a sequence is generated automatically based in this value
 - `do.plot` whether to plot the output of the cross validation step

Details

See Barigozzi, Cho and Owens (2022) and Owens, Cho and Barigozzi (2022) for further details. List arguments do not need to be specified with all list components; any missing entries will be filled in with the default argument.

Value

an S3 object of class `fnets`, which contains the following fields:

<code>q</code>	number of factors
<code>spec</code>	if <code>fm.restricted = FALSE</code> a list containing estimates of the spectral density matrices for <code>x</code> , common and idiosyncratic components
<code>acv</code>	a list containing estimates of the autocovariance matrices for <code>x</code> , common and idiosyncratic components
<code>loadings</code>	if <code>fm.restricted = TRUE</code> , factor loadings; if <code>fm.restricted = FALSE</code> and <code>q >= 1</code> , a list containing estimators of the impulse response functions (as an array of dimension $(p, q, \text{trunc.lags} + 2)$)
<code>factors</code>	if <code>fm.restricted = TRUE</code> , factor series; else, common shocks (an array of dimension (q, n))
<code>idio.var</code>	a list containing the following fields: <ul style="list-style-type: none"> • <code>beta</code> estimate of VAR parameter matrix; each column contains parameter estimates for the regression model for a given variable • <code>Gamma</code> estimate of the innovation covariance matrix • <code>lambda</code> regularisation parameter • <code>var.order</code> VAR order
<code>lrpc</code>	see the output of par.lrpc
<code>mean.x</code>	if <code>center = TRUE</code> , returns a vector containing row-wise sample means of <code>x</code> ; if <code>center = FALSE</code> , returns a vector of zeros
<code>var.method</code>	input parameter
<code>do.lrpc</code>	input parameter
<code>kern.bw</code>	input parameter

References

- Ahn, S. C. & Horenstein, A. R. (2013) Eigenvalue ratio test for the number of factors. *Econometrica*, 81(3), 1203–1227.
- Alessi, L., Barigozzi, M., & Capasso, M. (2010) Improved penalization for determining the number of factors in approximate factor models. *Statistics & Probability Letters*, 80(23-24):1806–1813.
- Bai, J. & Ng, S. (2002) Determining the number of factors in approximate factor models. *Econometrica*. 70: 191-221.
- Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series. arXiv preprint arXiv:2201.06110.
- Hallin, M. & Liška, R. (2007) Determining the number of factors in the general dynamic factor model. *Journal of the American Statistical Association*, 102(478), 603–617.
- Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

See Also

[predict.fnets](#), [plot.fnets](#)

Examples

```
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
x <- common$data + idio$data
out <- fnets(x,
  q = NULL, var.order = 1, var.method = "lasso", do.threshold = TRUE,
  do.lrpc = TRUE, tuning.args = list(tuning = "cv", n.folds = 1, path.length = 10, do.plot = TRUE),
  var.args = list(n.cores = 2)
)
pre <- predict(out, x, h = 1, common.method = "unrestricted")
plot(out, type = "granger", display = "network")
plot(out, type = "lrpc", display = "heatmap")
```

fnets.factor.model	<i>Factor model estimation</i>
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Description

Performs factor modelling under either restricted (static) or unrestricted (dynamic) factor models

Usage

```
fnets.factor.model(
  x,
  center = TRUE,
  fm.restricted = FALSE,
  q = c("ic", "er"),
  ic.op = NULL,
  kern.bw = NULL,
  common.args = list(factor.var.order = NULL, max.var.order = NULL, trunc.lags = 20,
    n.perm = 10)
)
```

Arguments

<code>x</code>	input time series matrix, with each row representing a variable
<code>center</code>	whether to de-mean the input <code>x</code> row-wise
<code>fm.restricted</code>	whether to estimate a restricted factor model using static PCA
<code>q</code>	<p>Either a string specifying the factor number selection method when <code>fm.restricted = TRUE</code>; possible values are:</p> <ul style="list-style-type: none"> • "ic" information criteria of Hallin and Liška (2007) or Bai and Ng (2002), see factor.number • "er" eigenvalue ratio <p>or the number of unrestricted factors.</p>
<code>ic.op</code>	choice of the information criterion penalty, see hl.factor.number or abc.factor.number for further details
<code>kern.bw</code>	kernel bandwidth for dynamic PCA; by default, it is set to $4 * \text{floor}((\text{dim}(x)[2]/\log(\text{dim}(x)[2]))^{1/4})$. When <code>fm.restricted = TRUE</code> , it is used to compute the number of lags for which autocovariance matrices are estimated
<code>common.args</code>	<p>a list specifying the tuning parameters required for estimating the impulse response functions and common shocks. It contains:</p> <ul style="list-style-type: none"> • <code>factor.var.order</code> order of the blockwise VAR representation of the common component. If <code>factor.var.order = NULL</code>, it is selected blockwise by Schwarz criterion • <code>max.var.order</code> maximum blockwise VAR order for the Schwarz criterion • <code>trunc.lags</code> truncation lag for impulse response function estimation • <code>n.perm</code> number of cross-sectional permutations involved in impulse response function estimation

Details

See Barigozzi, Cho and Owens (2022) for further details.

Value

an S3 object of class `fm`, which contains the following fields:

<code>q</code>	number of factors
<code>spec</code>	if <code>fm.restricted = FALSE</code> a list containing estimates of the spectral density matrices for <code>x</code> , common and idiosyncratic components
<code>acv</code>	a list containing estimates of the autocovariance matrices for <code>x</code> , common and idiosyncratic components
<code>loadings</code>	if <code>fm.restricted = TRUE</code> , factor loadings; if <code>fm.restricted = FALSE</code> and <code>q >= 1</code> , a list containing estimators of the impulse response functions (as an array of dimension $(p, q, \text{trunc.lags} + 2)$)
<code>factors</code>	if <code>fm.restricted = TRUE</code> , factor series; else, common shocks (an array of dimension (q, n))
<code>mean.x</code>	if <code>center = TRUE</code> , returns a vector containing row-wise sample means of <code>x</code> ; if <code>center = FALSE</code> , returns a vector of zeros

References

- Alessi, L., Barigozzi, M., & Capasso, M. (2010) Improved penalization for determining the number of factors in approximate factor models. *Statistics & Probability Letters*, 80(23-24):1806–1813.
- Bai, J. & Ng, S. (2002) Determining the number of factors in approximate factor models. *Econometrica*. 70: 191-221.
- Barigozzi, M., Cho, H. & Owens, D. (2022) Factor-adjusted network estimation and forecasting for high-dimensional time series. *arXiv preprint arXiv:2201.06110*.
- Hallin, M. & Liška, R. (2007) Determining the number of factors in the general dynamic factor model. *Journal of the American Statistical Association*, 102(478), 603–617.
- Owens, D., Cho, H. & Barigozzi, M. (2022) *fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling*

Examples

```
set.seed(123)
n <- 500
p <- 50
common <- sim.restricted(n, p)
x <- common$data
out <- fnets.factor.model(x, fm.restricted = TRUE)
```

fnets.var

*l1-regularised Yule-Walker estimation for VAR processes***Description**

Estimates the VAR parameter matrices via l1-regularised Yule-Walker estimation and innovation covariance matrix via constrained l1-minimisation.

Usage

```
fnets.var(
  x,
  center = TRUE,
  method = c("lasso", "ds"),
  lambda = NULL,
  var.order = 1,
  tuning.args = list(tuning = c("cv", "bic"), n.folds = 1, penalty = NULL, path.length
    = 10, do.plot = FALSE),
  do.threshold = FALSE,
  n.iter = NULL,
  tol = 0,
  n.cores = min(parallel::detectCores() - 1, 3)
)
```

Arguments

x	input time series matrix, with each row representing a variable
center	whether to de-mean the input x row-wise
method	a string specifying the method to be adopted for VAR process estimation; possible values are: <ul style="list-style-type: none"> • "lasso" Lasso-type l1-regularised M-estimation • "ds" Dantzig Selector-type constrained l1-minimisation
lambda	regularisation parameter; if lambda = NULL, tuning is employed to select the parameter
var.order	order of the VAR process; if a vector of integers is supplied, the order is chosen via tuning
tuning.args	a list specifying arguments for tuning for selecting the regularisation parameter (and VAR order). It contains: <ul style="list-style-type: none"> • tuning a string specifying the selection procedure for var.order and lambda; possible values are: <ul style="list-style-type: none"> – "cv" cross validation – "bic" information criterion • n.folds if tuning = "cv", positive integer number of folds

	<ul style="list-style-type: none"> • <code>penalty</code> if <code>tuning = "bic"</code>, penalty multiplier between 0 and 1; if <code>penalty = NULL</code>, it is set to $1/(1+\exp(\dim(x)[1]/\dim(x)[2]))$ by default • <code>path.length</code> positive integer number of regularisation parameter values to consider; a sequence is generated automatically based in this value • <code>do.plot</code> whether to plot the output of the cross validation step
<code>do.threshold</code>	whether to perform adaptive thresholding of VAR parameter estimator with threshold
<code>n.iter</code>	maximum number of descent steps, by default depends on <code>var.order</code> ; applicable when <code>method = "lasso"</code>
<code>tol</code>	numerical tolerance for increases in the loss function; applicable when <code>method = "lasso"</code>
<code>n.cores</code>	number of cores to use for parallel computing, see makePSOCKcluster ; applicable when <code>method = "ds"</code>

Details

Further information can be found in Barigozzi, Cho and Owens (2022).

Value

a list which contains the following fields:

<code>beta</code>	estimate of VAR parameter matrix; each column contains parameter estimates for the regression model for a given variable
<code>Gamma</code>	estimate of the innovation covariance matrix
<code>lambda</code>	regularisation parameter
<code>var.order</code>	VAR order
<code>mean.x</code>	if <code>center = TRUE</code> , returns a vector containing row-wise sample means of <code>x</code> ; if <code>center = FALSE</code> , returns a vector of zeros

References

Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series. arXiv preprint arXiv:2201.06110.

Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

Examples

```
library(fnets)

set.seed(123)
n <- 500
p <- 50
idio <- sim.var(n, p)
x <- idio$data
```

```
fv <- fnets.var(x,
  center = TRUE, method = "lasso", var.order = 1,
  tuning.args = list(tuning = "cv", n.folds = 1, path.length = 10, do.plot = TRUE),
  n.cores = 2
)
norm(fv$beta - t(idio$A), "F") / norm(t(idio$A), "F")
```

idio.predict	<i>Forecasting idiosyncratic VAR process</i>
--------------	--

Description

Produces forecasts of the idiosyncratic VAR process for a given forecasting horizon by estimating the best linear predictors

Usage

```
idio.predict(object, x, cpre, h = 1)
```

Arguments

object	fnets object
x	input time series matrix, with each row representing a variable
cpre	output of common.predict
h	forecast horizon

Value

a list containing	
is	in-sample estimator of the idiosyncratic component
fc	forecasts of the idiosyncratic component for a given forecasting horizon h
h	forecast horizon

References

Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series. arXiv preprint arXiv:2201.06110.

Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

Examples

```
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
x <- common$data + idio$data
out <- fnets(x, q = NULL, var.order = 1, var.method = "lasso",
do.lrpc = FALSE, var.args = list(n.cores = 2))
cpre <- common.predict(out, x, h = 1, r = NULL)
ipre <- idio.predict(out, x, cpre, h = 1)
```

par.lrpc	<i>Parametric estimation of long-run partial correlations of factor-adjusted VAR processes</i>
----------	--

Description

Returns a parametric estimate of long-run partial correlations of the VAR process from the VAR parameter estimates and the inverse of innovation covariance matrix obtained via constrained l1-minimisation.

Usage

```
par.lrpc(
  object,
  x,
  eta = NULL,
  tuning.args = list(n.folds = 1, path.length = 10, do.plot = FALSE),
  lrpc.adaptive = FALSE,
  eta.adaptive = NULL,
  do.correct = TRUE,
  do.threshold = FALSE,
  n.cores = min(parallel::detectCores() - 1, 3)
)
```

Arguments

object	fnets object
x	input time series matrix; with each row representing a variable
eta	regularisation parameter; if eta = NULL, it is selected by cross validation
tuning.args	a list specifying arguments for the cross validation procedure for selecting the tuning parameter involved in long-run partial correlation matrix estimation. It contains: <ul style="list-style-type: none"> • n.folds positive integer number of folds

	<ul style="list-style-type: none"> • <code>path.length</code> positive integer number of regularisation parameter values to consider; a sequence is generated automatically based in this value • <code>do.plot</code> whether to plot the output of the cross validation step, and if <code>do.threshold = TRUE</code>, plot the thresholding output
<code>lrpc.adaptive</code>	whether to use the adaptive estimation procedure
<code>eta.adaptive</code>	regularisation parameter for Step 1 of the adaptive estimation procedure; if <code>eta.adaptive = NULL</code> , defaults to $2 * \sqrt{\log(\dim(x)[1]) / \dim(x)[2]}$
<code>do.correct</code>	whether to correct for any negative entries in the diagonals of the inverse of long-run covariance matrix
<code>do.threshold</code>	whether to perform adaptive thresholding of Delta and Omega parameter estimators with threshold
<code>n.cores</code>	number of cores to use for parallel computing, see makePSOCKcluster

Details

See Barigozzi, Cho and Owens (2022) for further details, and Cai, Liu and Zhou (2016) for further details on the adaptive estimation procedure.

Value

	a list containing
<code>Delta</code>	estimated inverse of the innovation covariance matrix
<code>Omega</code>	estimated inverse of the long-run covariance matrix
<code>pc</code>	estimated innovation partial correlation matrix
<code>lrpc</code>	estimated long-run partial correlation matrix
<code>eta</code>	regularisation parameter
<code>lrpc.adaptive</code>	input argument

References

- Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series.
- Cai, T. T., Liu, W., & Zhou, H. H. (2016) Estimating sparse precision matrix: Optimal rates of convergence and adaptive estimation. *The Annals of Statistics*, 44(2), 455-488.
- Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

Examples

```
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
```

```

x <- common$data + idio$data
out <- fnets(x, q = NULL, var.method = "lasso", do.lrpc = FALSE, var.args = list(n.cores = 2))
plrpc <- par.lrpc(out, x,
  tuning.args = list(n.folds = 1, path.length = 10, do.plot = TRUE), n.cores = 2)
out$lrpc <- plrpc
out$do.lrpc <- TRUE
plot(out, type = "pc", display = "network", threshold = .05)
plot(out, type = "lrpc", display = "heatmap", threshold = .05)

```

plot.fnets

Plotting the networks estimated by fnets

Description

Plotting method for S3 objects of class `fnets`. Produces a plot visualising three networks underlying factor-adjusted VAR processes: (i) directed network representing Granger causal linkages, as given by estimated VAR transition matrices summed across the lags, (ii) undirected network representing contemporaneous linkages after accounting for lead-lag dependence, as given by partial correlations of VAR innovations, (iii) undirected network summarising (i) and (ii) as given by long-run partial correlations of VAR processes. Edge widths are determined by edge weights.

Usage

```

## S3 method for class 'fnets'
plot(
  x,
  type = c("granger", "pc", "lrpc"),
  display = c("network", "heatmap"),
  names = NA,
  groups = NA,
  threshold = 0,
  ...
)

```

Arguments

<code>x</code>	<code>fnets</code> object
<code>type</code>	a string specifying which of the above three networks (i)–(iii) to visualise; possible values are <ul style="list-style-type: none"> • "granger" directed network representing Granger causal linkages • "pc" undirected network representing contemporaneous linkages; available when <code>x\$do.lrpc = TRUE</code> • "lrpc" undirected network summarising Granger causal and contemporaneous linkages; available when <code>x\$do.lrpc = TRUE</code>
<code>display</code>	a string specifying how to visualise the network; possible values are:

- "network" as an igraph object, see [plot.igraph](#)
 - "heatmap" as a heatmap, see [imagePlot](#)
- names a character vector containing the names of the vertices
- groups an integer vector denoting any group structure of the vertices
- threshold if threshold > 0, hard thresholding is performed on the matrix giving rise to the network of interest
- ... additional arguments

Details

See Barigozzi, Cho and Owens (2022) for further details.

References

Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series. arXiv preprint arXiv:2201.06110.

Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

See Also

[fnets](#)

predict.fm	<i>Forecasting for factor models</i>
------------	--------------------------------------

Description

Produces forecasts of the data for a given forecasting horizon by estimating the best linear predictors of the common component

Usage

```
## S3 method for class 'fm'
predict(object, x, h = 1, fc.restricted = TRUE, r = c("ic", "er"), ...)
```

Arguments

- object fm object
- x input time series matrix, with each row representing a variable
- h forecasting horizon
- fc.restricted whether to forecast using a restricted or unrestricted, blockwise VAR representation of the common component
- r number of restricted factors, or a string specifying the factor number selection method when fc.restricted = TRUE; possible values are:

- "ic" information criteria of Alessi, Barigozzi & Capasso (2010)
 - "er" eigenvalue ratio
- ... not used

Value

a list containing

is in-sample predictions

forecast forecasts for the given forecasting horizon

r factor number

References

Ahn, S. C. & Horenstein, A. R. (2013) Eigenvalue ratio test for the number of factors. *Econometrica*, 81(3), 1203–1227.

Alessi, L., Barigozzi, M., & Capasso, M. (2010) Improved penalization for determining the number of factors in approximate factor models. *Statistics & Probability Letters*, 80(23-24):1806–1813.

Barigozzi, M., Cho, H. & Owens, D. (2022) Factor-adjusted network estimation and forecasting for high-dimensional time series. *arXiv preprint arXiv:2201.06110*.

Owens, D., Cho, H. & Barigozzi, M. (2022) *fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling*

See Also

[fnets.factor.model](#), [common.predict](#)

predict.fnets

Forecasting by fnets

Description

Produces forecasts of the data for a given forecasting horizon by separately estimating the best linear predictors of common and idiosyncratic components

Usage

```
## S3 method for class 'fnets'
predict(object, x, h = 1, fc.restricted = TRUE, r = c("ic", "er"), ...)
```

Arguments

object	fnets object
x	input time series matrix, with each row representing a variable
h	forecasting horizon
fc.restricted	whether to forecast using a restricted or unrestricted, blockwise VAR representation of the common component
r	number of restricted factors, or a string specifying the factor number selection method when <code>fc.restricted = TRUE</code> ; possible values are: <ul style="list-style-type: none"> • "ic" information criteria of Bai and Ng (2002) • "er" eigenvalue ratio
...	not used

Value

a list containing	
forecast	forecasts for the given forecasting horizon
common.pred	a list containing forecasting results for the common component
idio.pred	a list containing forecasting results for the idiosyncratic component
mean.x	mean.x argument from object

References

- Ahn, S. C. & Horenstein, A. R. (2013) Eigenvalue ratio test for the number of factors. *Econometrica*, 81(3), 1203–1227.
- Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series. *arXiv preprint arXiv:2201.06110*.
- Owens, D., Cho, H. & Barigozzi, M. (2022) *fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling*

See Also

[fnets](#), [common.predict](#), [idio.predict](#)

Examples

```
set.seed(123)
n <- 500
p <- 50
common <- sim.restricted(n, p)
idio <- sim.var(n, p)
x <- common$data + idio$data
out <- fnets(x, q = 2, var.order = 1, var.method = "lasso",
do.lrpc = FALSE, var.args = list(n.cores = 2))
cpre.unr <- common.predict(out, x, h = 1, fc.restricted = FALSE, r = NULL)
cpre.res <- common.predict(out, x, h = 1, fc.restricted = TRUE, r = NULL)
ipre <- idio.predict(out, x, cpre.res, h = 1)
```

sim.restricted	<i>Simulate data from a restricted factor model</i>
----------------	---

Description

Simulate the common component following an unrestricted factor model that admits a restricted representation; see the model (C2) in the reference.

Usage

```
sim.restricted(n, p, q = 2, heavy = FALSE)
```

Arguments

n	sample size
p	dimension
q	number of unrestricted factors; number of restricted factors is given by $2 * q$
heavy	if heavy = FALSE, common shocks are generated from <code>rnorm</code> whereas if heavy = TRUE, from <code>rt</code> with <code>df = 5</code> and then scaled by <code>sqrt(3 / 5)</code>

Value

a list containing	
data	generated series
q	number of factors
r	number of restricted factors

References

Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series.

Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

Examples

```
common <- sim.restricted(500, 50)
```

sim.unrestricted	<i>Simulate data from an unrestricted factor model</i>
------------------	--

Description

Simulate the common component following an unrestricted factor model that does not admit a restricted representation; see the model (C1) in the reference.

Usage

```
sim.unrestricted(n, p, q = 2, heavy = FALSE)
```

Arguments

n	sample size
p	dimension
q	number of unrestricted factors
heavy	if heavy = FALSE, common shocks are generated from <code>rnorm</code> whereas if heavy = TRUE, from <code>rt</code> with <code>df = 5</code> and then scaled by <code>sqrt(3 / 5)</code>

Value

a list containing	
data	generated series
q	number of factors

References

Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series. arXiv preprint arXiv:2201.06110

Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

Examples

```
common <- sim.unrestricted(500, 50)
```

sim.var	<i>Simulate a VAR(1) process</i>
---------	----------------------------------

Description

Simulate a VAR(1) process; see the reference for the generation of the transition matrix.

Usage

```
sim.var(n, p, Gamma = diag(1, p), heavy = FALSE)
```

Arguments

n	sample size
p	dimension
Gamma	innovation covariance matrix; ignored if heavy = TRUE
heavy	if heavy = FALSE, common shocks are generated from <code>rnorm</code> whereas if heavy = TRUE, from <code>rt</code> with <code>df = 5</code> and then scaled by <code>sqrt(3 / 5)</code>

Value

a list containing	
data	generated series
A	transition matrix
Gamma	innovation covariance matrix

References

Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network estimation and forecasting for high-dimensional time series.

Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

Examples

```
idio <- sim.var(500, 50)
```

threshold	<i>Edge selection for VAR parameter, inverse innovation covariance, and long-run partial correlation matrices</i>
-----------	---

Description

Threshold the entries of the input matrix at a data-driven level to perform edge selection

Usage

```
threshold(mat, path.length = 500, do.plot = FALSE)
```

Arguments

mat	input parameter matrix
path.length	number of candidate thresholds
do.plot	whether to plot thresholding output

Details

See Liu, Zhang, and Liu (2021) for more information on the threshold selection process

Value

a list which contains the following fields:

threshold	data-driven threshold
thr.mat	thresholded input matrix

References

Barigozzi, M., Cho, H. & Owens, D. (2022) FNETS: Factor-adjusted network analysis for high-dimensional time series. arXiv preprint arXiv:2201.06110.

Liu, B., Zhang, X. & Liu, Y. (2021) Simultaneous Change Point Inference and Structure Recovery for High Dimensional Gaussian Graphical Models. *Journal of Machine Learning Research*, 22(274), 1–62.

Owens, D., Cho, H. & Barigozzi, M. (2022) fnets: An R Package for Network Estimation and Forecasting via Factor-Adjusted VAR Modelling

Examples

```
set.seed(123)
A <- diag(.7, 50) + rnorm(50^2, 0, .1)
threshold.A <- threshold(A)
```

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