# Package 'fnets'

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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 factoradjusted vector autoregressive model.  See Barigozzi, Cho and Owens (2022) <arxiv:2201.06110> for further descriptions of FNETS methodology and Owens, Cho and Barigozzi (2023) <arxiv:2301.11675> accompanying the R package.</arxiv:2301.11675></arxiv:2201.06110>
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factor.number

2 factor.number

	fnets.var	10
	network	12
	network.fnets	12
	par.lrpc	15
	plot.factor.number	17
	plot.fnets	18
	plot.threshold	20
	predict.fm	21
	predict.fnets	22
	print.factor.number	24
	print.fm	25
	print.fnets	26
	print.threshold	27
	sim.restricted	28
	sim.unrestricted	29
	sim.var	30
	threshold	31
Index		32

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,
    center = TRUE
)
```

### **Arguments**

x input time series matrix, with each row representing a variable and each column containing the observations at a given time

fm.restricted whether to estimate the number of restricted or unrestricted factors

method A string specifying the factor number selection method; possible values are:

factor.number 3

```
• "ic" information criteria-based methods of Alessi, Barigozzi & Capasso (2010) when fm.restricted = TRUE or Hallin and Liška (2007) when fm.restricted = FALSE
```

• "er" eigenvalue ratio of Ahn and Horenstein (2013)

```
q.max maximum number of factors; if q.max = NULL, a default value is selected as min(50, floor(sqrt(min(dim(x)[2] - 1, dim(x)[1]))))
```

center whether to de-mean the input x row-wise

#### **Details**

For further details, see references.

#### Value

S3 object of class factor.number. 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. arXiv preprint arXiv:2301.11675.

#### See Also

plot.factor.number, print.factor.number

```
library(fnets)
## Alessi, Barigozzi, and Capasso method for restricted models
set.seed(123)
n <- 500
p <- 50
common <- sim.restricted(n, p)
idio <- sim.var(n, p)
x <- common$$\frac{1}{2}$ apply(idio$$\frac{1}{2}$ data, 1, sd) / apply(common$$\frac{1}{2}$ data
abc <- factor.number(x, fm.restricted = TRUE)
print(abc)
plot(abc)</pre>
```

4 fnets

```
## Eigenvalue ratio method
er <- factor.number(x, method = "er", fm.restricted = TRUE)
print(er)
plot(er)

## Hallin and Liška method for unrestricted models
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
x <- common$$\frac{1}{2}$ apply(idio$$\frac{1}{2}$ data, 1, sd) / apply(common$$\frac{1}{2}$ data

hl <- factor.number(x, fm.restricted = FALSE)
print(hl)
plot(hl)</pre>
```

fnets

Factor-adjusted network estimation

### **Description**

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(
  Х,
  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
)
```

5 fnets

#### **Arguments**

input time series matrix, with each row representing a variable and each column Х

containing the observations at a given time

whether to de-mean the input x row-wise center

whether to estimate a restricted factor model using static PCA fm.restricted

> Either the number of factors or a string specifying the factor number selection method; possible values are:

- "ic" information criteria-based methods of Alessi, Barigozzi & Capasso (2010) when fm. restricted = TRUE or Hallin and Liška (2007) when fm. restricted = FALSE modifying Bai and Ng (2002)
- "er" eigenvalue ratio of Ahn and Horenstein (2013)

see factor.number.

choice of the information criterion penalty, see factor.number for further details

kern.bw a positive integer specifying the kernel bandwidth for dynamic PCA; by default, it is set to floor  $(4 * (\dim(x)[2]/\log(\dim(x)[2]))^(1/3))$ . When fm. restricted = TRUE, it is used to compute the number of lags for which au-

tocovariance matrices are estimated

a list specifying the tuning parameters required for estimating the impulse recommon.args sponse functions and common shocks. It contains:

- factor.var.order order of the blockwise VAR representation of the common component. If factor.var.order = NULL, it is selected blockwise by Schwarz criterion
- max.var.order maximum blockwise VAR order for the Schwarz criterion
- trunc.lags truncation lag for impulse response function estimation
- n.perm number of cross-sectional permutations involved in impulse response function estimation

var.order order of the idiosyncratic VAR process; if a vector of integers is supplied, the order is chosen via tuning

> a string specifying the method to be adopted for idiosyncratic VAR process estimation; possible values are:

- "lasso" Lasso-type 11-regularised M-estimation
- "ds" Dantzig Selector-type constrained 11-minimisation

a list specifying the tuning parameters required for estimating the idiosyncratic VAR process. It contains:

- n.iter maximum number of descent steps, by default depends on var.order; applicable when var.method = "lasso"
- n.cores number of cores to use for parallel computing, see makePSOCKcluster; applicable when var.method = "ds"

do.threshold whether to perform adaptive thresholding of all parameter estimators with thresh-

do.lrpc whether to estimate the long-run partial correlation whether to use the adaptive estimation procedure lrpc.adaptive

ic.op

var.method

var.args

6 fnets

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

#### **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:

q	number of factors
spec	if fm.restricted = FALSE a list containing estimates of the spectral density matrices for x, common and idiosyncratic components
acv	a list containing estimates of the autocovariance matrices for $\boldsymbol{x}$ , common and idiosyncratic components
loadings	if fm.restricted = TRUE, factor loadings; if fm.restricted = FALSE and q >= 1, a list containing estimators of the impulse response functions (as an array of dimension (p, q, trunc.lags + 2))
factors	if fm.restricted = TRUE, factor series; else, common shocks (an array of dimension $(q, n)$ )
idio.var	a list containing the following fields:
	• beta estimate of VAR parameter matrix; each column contains parameter estimates for the regression model for a given variable
	Gamma estimate of the innovation covariance matrix
	lambda regularisation parameter
	• var.order VAR order
lrpc	see the output of par.lrpc
mean.x	if center = TRUE, returns a vector containing row-wise sample means of $x$ ; if center = FALSE, returns a vector of zeros
var.method	input parameter
do.lrpc	input parameter
kern.bw	input parameter

fnets.factor.model 7

#### 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. arXiv preprint arXiv:2301.11675.

#### See Also

predict.fnets, plot.fnets, print.fnets

### **Examples**

fnets.factor.model

Factor model estimation

#### **Description**

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

8 fnets.factor.model

#### Usage

```
fnets.factor.model(
  Х,
  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**

kern.bw

input time series matrix, with each row representing a variable and each column Χ

containing the observations at a given time

whether to de-mean the input x row-wise center

fm.restricted whether to estimate a restricted factor model using static PCA

Either a string specifying the factor number selection method when fm. restricted = TRUE; possible values are:

• "ic" information criteria of Hallin and Liška (2007) or Bai and Ng (2002),

see factor.number • "er" eigenvalue ratio

or the number of unrestricted factors.

ic.op choice of the information criterion penalty, see hl.factor.number or abc.factor.number

for further details

kernel bandwidth for dynamic PCA; by default, it is set to  $4 * floor((dim(x)[2]/log(dim(x)[2]))^{(1)}$ When fm.restricted = TRUE, it is used to compute the number of lags for

which autocovariance matrices are estimated

a list specifying the tuning parameters required for estimating the impulse recommon.args sponse functions and common shocks. It contains:

> • factor.var.order order of the blockwise VAR representation of the common component. If factor.var.order = NULL, it is selected blockwise by Schwarz criterion

- max.var.order maximum blockwise VAR order for the Schwarz criterion
- trunc.lags truncation lag for impulse response function estimation
- n.perm number of cross-sectional permutations involved in impulse response function estimation

#### **Details**

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

fnets.factor.model 9

#### Value

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

q	number of factors
spec	if fm.restricted = FALSE a list containing estimates of the spectral density matrices for x, common and idiosyncratic components
acv	a list containing estimates of the autocovariance matrices for $\boldsymbol{x}$ , common and idiosyncratic components
loadings	if fm.restricted = TRUE, factor loadings; if fm.restricted = FALSE and q >= 1, a list containing estimators of the impulse response functions (as an array of dimension (p, q, trunc.lags + 2))
factors	if fm.restricted = TRUE, factor series; else, common shocks (an array of dimension $(q, n)$ )
mean.x	if center = TRUE, returns a vector containing row-wise sample means of x; if center = FALSE, 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. arXiv preprint arXiv:2301.11675.

### See Also

print.fm, predict.fm

```
set.seed(1234)
n <- 500
p <- 50
common <- sim.restricted(n, p)
x <- common$\data + matrix(rnorm(n * p), nrow = p)
out <- fnets.factor.model(x, fm.restricted = FALSE)</pre>
```

10 fnets.var

fnets.var

11-regularised Yule-Walker estimation for VAR processes

#### **Description**

Estimates the VAR parameter matrices via 11-regularised Yule-Walker estimation and innovation covariance matrix via constrained 11-minimisation.

### Usage

```
fnets.var(
  Х,
  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
  do.threshold = FALSE,
 n.iter = NULL,
 tol = 0,
  n.cores = min(parallel::detectCores() - 1, 3)
)
```

#### **Arguments**

lambda

input time series matrix, with each row representing a variable and each column

containing the observations at a given time

whether to de-mean the input x row-wise center

method a string specifying the method to be adopted for VAR process estimation; possible values are:

• "lasso" Lasso-type 11-regularised M-estimation

• "ds" Dantzig Selector-type constrained 11-minimisation

11-regularisation parameter; if lambda = NULL, tuning is employed to select the parameter

order of the VAR process; if a vector of integers is supplied, the order is chosen var.order via tuning

tuning.args a list specifying arguments for tuning for selecting the regularisation parameter (and VAR order). 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

fnets.var 11

• 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.threshold whether to perform adaptive thresholding of VAR parameter estimator with

threshold

n.iter maximum number of descent steps, by default depends on var.order; applica-

ble when method = "lasso"

tol numerical tolerance for increases in the loss function; applicable when method

= "lasso"

n. cores number of cores to use for parallel computing, see makePSOCKcluster; appli-

cable when method = "ds"

#### **Details**

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

#### Value

a list which contains the following fields:

beta estimate of VAR parameter matrix; each column contains parameter estimates

for the regression model for a given variable

Gamma estimate of the innovation covariance matrix

lambda 11-regularisation parameter

var.order VAR order

mean.x if center = TRUE, returns a vector containing row-wise sample means of x; if

center = FALSE, 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. arXiv preprint arXiv:2301.11675.

12 network.fnets

network

Convert networks into igraph objects

### Description

Convert networks into igraph objects

### Usage

```
network(object, ...)
```

### **Arguments**

object object

... additional arguments

#### See Also

network.fnets

network.fnets

Convert networks estimated by fnets into igraph objects

#### **Description**

Converts S3 objects of class fnets into a network. Produces an igraph object for the 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.

Converts S3 objects of class fnets into a network. Produces an igraph object for the 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.

network.fnets 13

#### Usage

```
## S3 method for class 'fnets'
network(
  object,
  type = c("granger", "pc", "lrpc"),
  names = NA,
  groups = NA,
  ...
)

## S3 method for class 'fnets'
network(
  object,
  type = c("granger", "pc", "lrpc"),
  names = NA,
  groups = NA,
  ...
)
```

#### **Arguments**

object fnets object

type a string specifying which of the above three networks (i)–(iii) to visualise; pos-

sible values are

• "granger" directed network representing Granger causal linkages

• "pc" undirected network representing contemporaneous linkages; available when object\$do.lrpc = TRUE

• "1rpc" undirected network summarising Granger causal and contemporaneous linkages; available when x\$do.1rpc = TRUE

names a character vector containing the names of the vertices

groups an integer vector denoting any group structure of the vertices

.. additional arguments to igraph::graph\_from\_adjacency\_matrix

threshold if threshold > 0, hard thresholding is performed on the matrix giving rise to

the network of interest

### Details

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

### Value

A list with the following fields:

network igraph object names input argument 14 network.fnets

```
groups input argument
grp.col colours corresponding to each group
S3 object of type igraph
```

#### 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. arXiv preprint arXiv:2301.11675.

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. arXiv preprint arXiv:2301.11675.

#### See Also

```
fnets, plot.fnets
fnets, plot.fnets
```

```
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)</pre>
idio <- sim.var(n, p)</pre>
x <- common$data + idio$data
out <- fnets(x,
  do.threshold = TRUE,
  var.args = list(n.cores = 2)
)
network(out, type = "granger")
network(out, type = "pc")
network(out, type = "lrpc")
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)</pre>
idio <- sim.var(n, p)</pre>
x <- common$data + idio$data
out <- fnets(x,
  do.threshold = TRUE,
  var.args = list(n.cores = 2)
network(out, type = "granger")
```

par.lrpc 15

```
network(out, type = "pc")
network(out, type = "lrpc")
```

par.lrpc

Parametric estimation of long-run partial correlations of factoradjusted VAR processes

### 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 11-minimisation.

### Usage

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

fnets object

### Arguments

object

x	input time series matrix with each row representing a variable
eta	11-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> <li>n.folds positive integer number of folds</li> </ul>
	<ul> <li>path.length positive integer number of regularisation parameter values to consider; a sequence is generated automatically based in this value</li> </ul>
lrpc.adaptive	whether to use the adaptive estimation procedure
eta.adaptive	11-regularisation parameter for Step 1 of the adaptive estimation procedure; if eta.adaptive = NULL, the default choice is $2 * sqrt(log(dim(x)[1])/dim(x)[2])$
do.correct	whether to correct for any negative entries in the diagonals of the inverse of long-run covariance matrix
do.threshold	whether to perform adaptive thresholding of Delta and Omega parameter estimators with threshold
n.cores	number of cores to use for parallel computing, see makePSOCKcluster

par.lrpc

#### **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

Delta estimated inverse of the innovation covariance matrix

Omega estimated inverse of the long-run covariance matrix

pc estimated innovation partial correlation matrix

lrpc estimated long-run partial correlation matrix

eta 11-regularisation parameter

lrpc.adaptive input argument

#### 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.

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. arXiv preprint arXiv:2301.11675.

```
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
x <- common$\forall data + idio$\forall data
out <- fnets(x, do.lrpc = FALSE, var.args = list(n.cores = 2))
plrpc <- par.lrpc(out, x, n.cores = 2)
out$\forall lrpc <- plrpc
out$\forall lrpc <- TRUE
plot(out, type = "pc", display = "network")
plot(out, type = "lrpc", display = "heatmap")</pre>
```

plot.factor.number 17

plot.factor.number

Plot factor number

### **Description**

Plots the eigenvalue ratio or information criteria from a factor.number object

### Usage

```
## S3 method for class 'factor.number' plot(x, ...)
```

#### **Arguments**

```
x factor.number object ... not used
```

#### Value

NULL, printed to console

#### 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

factor.number

```
library(fnets)
## Alessi, Barigozzi, and Capasso method for restricted models
set.seed(123)
n <- 500
p <- 50
common <- sim.restricted(n, p)
idio <- sim.var(n, p)
x <- common$$\frac{1}{2}$ apply(idio$$\frac{1}{2}$ data, 1, sd) / apply(common$$\frac{1}{2}$ data, 1, sd) + idio$$\frac{1}{2}$ data
abc <- factor.number(x, fm.restricted = TRUE)
print(abc)
plot(abc)
## Eigenvalue ratio method</pre>
```

18 plot.fnets

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

## Hallin and Liška method for unrestricted models
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
x <- common$$\frac{1}{2}$$ apply(idio$$\frac{1}{2}$$ data, 1, sd) / apply(common$$\frac{1}{2}$$ data

hl <- factor.number(x, fm.restricted = FALSE)
print(hl)
plot(hl)</pre>
```

plot.fnets

Plotting the networks estimated by fnets

### Description

Plotting method for S3 objects of class fnets. When display = "network" or "heatmap", it 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. When display = "tuning", it produces up to two plots (when do.larpc = TRUE) visualising the outcome of CV or IC adopted for selecting the 11-regularisation parameters and the VAR order.

#### Usage

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

### **Arguments**

Χ

fnets object

plot.fnets 19

type a string specifying which of the above three networks (i)—(iii) to visualise when display = "network" or display = "heatmap"; possible values are

- "granger" directed network representing Granger causal linkages
- "pc" undirected network representing contemporaneous linkages; available when x\$do.1rpc = TRUE
- "1rpc" undirected network summarising Granger causal and contemporaneous linkages; available when x\$do.1rpc = TRUE

display a string specifying which plot to produce; possible values are

- "network" visualise the network as an igraph object, see plot.igraph
- "heatmap" visualise the network as a heatmap, see imagePlot
- "tuning" visualise the outcome from CV or IC (specified by tuning.args\$tuning of fnets) for selecting 11-regularisation parameters and the VAR order

names a character vector containing the names of the network vertices
groups an integer vector denoting any group structure of the network vertices
v.colours a vector denoting vertex colours corresponding to groups
additional arguments

#### **Details**

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

#### Value

A plot produced as per the input arguments

#### 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. arXiv preprint arXiv:2301.11675.

### See Also

fnets

```
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,
    do.threshold = TRUE,</pre>
```

20 plot.threshold

```
var.args = list(n.cores = 2)
)
plot(out, type = "granger", display = "network",
groups = rep(c(1,2), p/2), v.colours = c("orange","blue"))
plot(out, type = "lrpc", display = "heatmap")
plot(out, display = "tuning")
```

plot.threshold

Plotting the thresholding procedure

### **Description**

Plotting method for S3 objects of class threshold. Produces a plot visualising three diagnostics for the thresholding procedure, with threshold values t\_k (x axis) against (i) Ratio\_k, the ratio of edges to non-edges in the matrix entries (ii) Diff\_k, the first difference of Ratio\_k (iii) |CUSUM\_k|, the absolute scaled cumulative sums of Diff\_k

### Usage

```
## S3 method for class 'threshold'
plot(x, plots = c(TRUE, FALSE, TRUE), ...)
```

#### **Arguments**

```
x threshold object
plots logical vector, which plots to use (Ratio, Diff, CUSUM respectively)
... additional arguments
```

#### **Details**

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

#### Value

A network plot produced as per the input arguments

#### 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. arXiv preprint arXiv:2301.11675.

### See Also

threshold

predict.fm 21

### **Examples**

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

predict.fm

Forecasting for factor models

### 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 = NULL, h = 1, fc.restricted = TRUE, r = c("ic", "er"), ...)
```

### **Arguments**

object	fm object
х	input time series matrix, with each row representing a variable; by default, uses input to object $% \left\{ 1,2,\ldots,4,3,4,4,4,4,4,4,4,4,4,4,4,4,4,4,4,4,4$
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:
	<ul><li>"ic" information criteria of Alessi, Barigozzi &amp; Capasso (2010)</li><li>"er" eigenvalue ratio</li></ul>
	not used

### Value

a list containing

is in-sample predictions

forecast for the given forecasting horizon

r factor number

22 predict.fnets

#### 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. arXiv preprint arXiv:2301.11675.

#### See Also

fnets.factor.model

#### **Examples**

```
set.seed(123)
n <- 500
p <- 50
common <- sim.restricted(n, p)
x <- common$\data + rnorm(n*p)
out <- fnets.factor.model(x, fm.restricted = TRUE)
pre <- predict(out)</pre>
```

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 = NULL, h = 1, fc.restricted = TRUE, r = c("ic", "er"), ...)
```

#### Arguments

object	fnets object
X	input time series matrix, with each row representing a variable; by default, uses input to $\mbox{object}$
h	forecasting horizon
fc.restricted	whether to forecast using a restricted or unrestricted, blockwise VAR representation of the common component

predict.fnets 23

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 Bai and Ng (2002)
- "er" eigenvalue ratio

... not used

#### Value

### a list containing

forecast 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

```
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,
do.lrpc = FALSE, var.args = list(n.cores = 2))
pre.unr <- predict(out, fc.restricted = FALSE)
pre.res <- predict(out, fc.restricted = TRUE)</pre>
```

24 print.factor.number

### **Description**

Prints a summary of a factor.number object

### Usage

```
## S3 method for class 'factor.number'
print(x, ...)
```

### **Arguments**

```
x factor.number object ... not used
```

#### Value

NULL, printed to console

#### 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

factor.number

```
library(fnets)
## Alessi, Barigozzi, and Capasso method for restricted models
set.seed(123)
n <- 500
p <- 50
common <- sim.restricted(n, p)
idio <- sim.var(n, p)
x <- common$$\frac{1}{2}$ apply(idio$$\frac{1}{2}$ data, 1, sd) / apply(common$$\frac{1}{2}$ data, 1, sd) + idio$$\frac{1}{2}$ data
abc <- factor.number(x, fm.restricted = TRUE)
print(abc)
plot(abc)
## Eigenvalue ratio method</pre>
```

print.fm 25

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

## Hallin and Liška method for unrestricted models
set.seed(123)
n <- 500
p <- 50
common <- sim.unrestricted(n, p)
idio <- sim.var(n, p)
x <- common$$\frac{1}{2}$$ apply(idio$$\frac{1}{2}$data, 1, sd) / apply(common$$\frac{1}{2}$$ data

hl <- factor.number(x, fm.restricted = FALSE)
print(hl)
plot(hl)</pre>
```

print.fm

Print factor model

### Description

Prints a summary of a fm object

### Usage

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

### Arguments

x fm object ... not used

#### Value

NULL, printed to console

#### 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.factor.model

26 print.fnets

### **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.factor.model(x, q = 2)
print(out)</pre>
```

print.fnets

Print fnets

### **Description**

Prints a summary of a fnets object

### Usage

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

### **Arguments**

```
x fnets object ... not used
```

### Value

NULL, printed to console

### 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

print.threshold 27

### **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,
do.lrpc = FALSE, var.args = list(n.cores = 2))
print(out)
x <- idio$data
out <- fnets.var(x,
n.cores = 2)
print(out)</pre>
```

print.threshold

Print threshold

### **Description**

Prints a summary of a threshold object

### Usage

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

#### **Arguments**

x threshold object ... not used

### Value

NULL, printed to console

#### 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

threshold

28 sim.restricted

#### **Examples**

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

sim.restricted

Simulate data from a restricted factor model

### **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
р	dimension

q number of unrestricted factors; number of restricted factors is given by 2 \* q heavy if heavy = FALSE, common shocks are generated from rnorm whereas if heavy

= TRUE, from rt with df = 5 and then scaled by sqrt(3 / 5)

### Value

a list containing

data time series matrix with n rows and p columns

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

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

sim.unrestricted 29

Sim	unrestricte	1

Simulate data from an unrestricted factor model

### 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
р	dimension

q number of unrestricted factors

heavy if heavy = FALSE, common shocks are generated from rnorm whereas if heavy

= TRUE, from rt with df = 5 and then scaled by sqrt(3 / 5)

### Value

a list containing

data time series matrix with n rows and p columns

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. arXiv preprint arXiv:2301.11675.

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

30 sim.var

sim.var	Simulate a VAR(1) process
---------	---------------------------

### 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 rnorm whereas if heavy

= TRUE, from rt with df = 5 and then scaled by sqrt(3 / 5)

#### Value

a list containing

data time series matrix with n rows and p columns

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

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

threshold 31

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

### Description

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

### Usage

```
threshold(mat, path.length = 500)
```

### **Arguments**

mat input parameter matrix

path.length number of candidate thresholds

#### **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. arXiv preprint arXiv:2301.11675.

### See Also

plot.threshold, print.threshold

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

## **Index**

```
abc.factor.number, 8
factor.number, 2, 5, 8, 17, 24
fnets, 4, 14, 19, 23, 26
fnets.factor.model, 7, 22, 25
fnets.var, 10
hl.factor.number, 8
imagePlot, 19
makePSOCKcluster, 5, 11, 15
network, 12
network.fnets, 12, 12
par.lrpc, 6, 15
plot.factor.number, 3, 17
plot.fnets, 7, 14, 18
plot.igraph, 19
plot.threshold, 20, 31
predict.fm, 9, 21
predict.fnets, 7, 22
print.factor.number, 3, 24
print.fm, 9, 25
print.fnets, 7, 26
print.threshold, 27, 31
sim.restricted, 28
sim.unrestricted, 29
sim.var, 30
threshold, 5, 11, 15, 20, 27, 31
```