

Package ‘rrcovNA’

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Title Scalable Robust Estimators with High Breakdown Point for Incomplete Data

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Description Robust Location and Scatter Estimation and Robust Multivariate Analysis with High Breakdown Point for Incomplete Data (missing values) (Todorov et al. (2010) <[doi:10.1007/s11634-010-0075-2](https://doi.org/10.1007/s11634-010-0075-2)>).

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Imports stats4, lattice, norm, cluster

Suggests grid

LazyLoad yes

License GPL (>= 2)

URL <https://github.com/valentint/rrcovNA>

BugReports <https://github.com/valentint/rrcovNA/issues>

NeedsCompilation yes

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bush10	<i>Campbell Bushfire Data with added missing data items (10 percent)</i>
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Description

This data set is based on the bushfire data set which was used by Campbell (1984) to locate bushfire scars - see [bushfire](#) in package `robustbase`. The original dataset contains satellite measurements on five frequency bands, corresponding to each of 38 pixels. The data set is very well studied (Maronna and Yohai, 1995; Maronna and Zamar, 2002). There are 12 clear outliers: 33-38, 32, 7-11 and 12 and 13 are suspect.

Usage

```
data(bush10)
```

Format

A data frame with 38 observations on 6 variables.

The original data set consists of 38 observations in 5 variables. Based on it four new data sets are created in which some of the data items are replaced by missing values with a simple "missing completely at random " mechanism. For this purpose independent Bernoulli trials are realized for each data item with a probability of success 0.1 where success means that the corresponding item is set to missing.)

Source

Maronna, R.A. and Yohai, V.J. (1995) The Behaviour of the Stahel-Donoho Robust Multivariate Estimator. *Journal of the American Statistical Association* **90**, 330–341.

Beguin, C. and Hulliger, B. (2004) Multivariate outlier detection in incomplete survey data: the epidemic algorithm and transformed rank correlations. *Journal of the Royal Statistical Society: Series B (Statistical Methodology)* **127**, 2, 275–294.

Examples

```
## The following code will result in exactly the same output
## as the one obtained from the original data set
data(bush10)
plot(bush10)
CovNAMcd(bush10)

## Not run:
## This is the code with which the missing data were created:
## Creates a data set with missing values (for testing purposes)
## from a complete data set 'x'. The probability of
## each item being missing is 'pr'.
##
getmiss <- function(x, pr=0.1){
  library(Rlab)
  n <- nrow(x)
  p <- ncol(x)
  bt <- rbern(n*p, pr)
  btmat <- matrix(bt, nrow=n)
  btmiss <- ifelse(btmat==1, NA, 0)
  x+btmiss
}

## End(Not run)
```

ces

Consumer Expenditure Survey Data

Description

This data set has been derived from the Quarterly Interview Survey of the Consumer Expenditure Survey (CES) undertaken by the U.S. Department of Labor, Bureau of Labor Statistics and is available at <https://www.bls.gov/cex/> where also more details about this survey can be found. The original data set comprises 869 households in 34 variables of which one is unique ID, five characterize the size of the household, further 6 variables contain other characteristics of the household like age, education ethnicity, etc. and 22 variables represent the household expenditures. We will consider a reduced set of only 8 expenditure variables. This reduced data set was analyzed by Hubert et al. (2009) in the context of PCA and the first step of the analysis showed that all variables are highly skewed. They applied the robust PCA method of Serneels and Verdonck based on the EM algorithm, since some of the data are incomplete.

Usage

```
data(ces)
```

Format

A data frame with 869 observations on the following 8 variables:

EXP Total household expenditure
 FDH0 Food and nonalcoholic beverages consumed at home
 FDAW Food and nonalcoholic beverages consumed away from home
 SHEL Housing expenditure
 TELE Telephone services
 CLOT Clothing
 HEAL Health care
 ENT Entertainment

Source

<https://www.bls.gov/cex/>

References

Hubert, M, Rousseeuw, P.J. and Verdonck, T., (2009). Robust PCA for skewed data and its outlier map, *Computational Statistics & Data Analysis*, **53**, 6, pp. 2264-2274

Examples

```
data(ces)
summary(ces)
plot(ces)
```

CovNA-class	<i>Class "CovNA" – a base class for estimates of multivariate location and scatter for incomplete data</i>
-------------	--

Description

The class CovNA represents an estimate of the multivariate location and scatter of a data set. The objects of class CovNA contain the classical estimates and serve as base for deriving other estimates, i.e. different types of robust estimates.

Objects from the Class

Objects can be created by calls of the form `new("CovNA", ...)`, but the usual way of creating CovNA objects is a call to the function `CovNA` which serves as a constructor.

Slots

call: Object of class "language"
cov: covariance matrix
center: location
n.obs: number of observations used for the computation of the estimates
mah: mahalanobis distances
det: determinant
flag: flags (FALSE if suspected an outlier)
method: a character string describing the method used to compute the estimate: "Classic"
singularity: a list with singularity information for the covariance matrix (or NULL of not singular)
X: data

Extends

Class "**Cov-class**", directly.

Methods

getDistance signature(obj = "CovNA"): distances
getFlag signature(obj = "CovNA"): Flags observations as outliers if the corresponding mahalanobis distance is larger then `qchisq(prob, p)` where prob defaults to 0.975.
summary signature(object = "CovNA"): calculate summary information

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

Examples

```
showClass("CovNA")
```

CovNAClassic	<i>Classical Estimates of Multivariate Location and Scatter for incomplete data (EM Algorithm)</i>
--------------	--

Description

Computes the classical estimates of multivariate location and scatter. Returns an S4 class `CovNAClassic` with the estimated center, cov, Mahalanobis distances and weights based on these distances.

Usage

```
CovNAClassic(x, unbiased=TRUE)
CovNA(x, unbiased=TRUE)
```

Arguments

x	a matrix or data frame. As usual, rows are observations and columns are variables.
unbiased	whether to return the unbiased estimate of the covariance matrix. Default is unbiased = TRUE

Value

An object of class "CovNAClassic".

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[Cov-class](#), [CovClassic-class](#), [CovNAClassic-class](#)

Examples

```
data(bush10)
cv <- CovNAClassic(bush10)
cv
summary(cv)
```

CovNAClassic-class	<i>Class "CovNAClassic" - classical estimates of multivariate location and scatter for incomplete data (EM algorithm)</i>
--------------------	---

Description

The class `CovNAClassic` represents an estimate of the multivariate location and scatter of an incomplete data set. The class `CovNAClassic` objects contain the classical estimates.

Objects from the Class

Objects can be created by calls of the form `new("CovNAClassic", ...)`, but the usual way of creating `CovNAClassic` objects is a call to the function `CovNAClassic` which serves as a constructor.

Slots

`call`: Object of class "language"
`cov`: covariance matrix
`center`: location
`n.obs`: number of observations used for the computation of the estimates
`mah`: mahalanobis distances
`method`: a character string describing the method used to compute the estimate: "Classic"
`singularity`: a list with singularity information for the covariance matrix (or NULL if not singular)
`X`: data

Methods

plot signature(x = "CovNAClassic"): plot the object

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

Examples

```
data(bush10)
cv <- CovNAClassic(bush10)
cv
summary(cv)
```

Description

Computes a robust multivariate location and scatter estimate with a high breakdown point for incomplete data, using the ‘Fast MCD’ (Minimum Covariance Determinant) estimator.

Usage

```
CovNAMcd(x, alpha = 1/2, nsamp = 500, seed = NULL, trace = FALSE,
use.correction = TRUE, impMeth = c("norm", "seq", "rseq"), control)
```

Arguments

<code>x</code>	a matrix or data frame.
<code>alpha</code>	numeric parameter controlling the size of the subsets over which the determinant is minimized, i.e., $\alpha \cdot n$ observations are used for computing the determinant. Allowed values are between 0.5 and 1 and the default is 0.5.
<code>nsamp</code>	number of subsets used for initial estimates or "best" or "exact". Default is <code>nsamp = 500</code> . For <code>nsamp="best"</code> exhaustive enumeration is done, as long as the number of trials does not exceed 5000. For "exact", exhaustive enumeration will be attempted however many samples are needed. In this case a warning message will be displayed saying that the computation can take a very long time.
<code>seed</code>	starting value for random generator. Default is <code>seed = NULL</code>
<code>trace</code>	whether to print intermediate results. Default is <code>trace = FALSE</code>
<code>use.correction</code>	whether to use finite sample correction factors. Default is <code>use.correction=TRUE</code>
<code>impMeth</code>	select imputation method to use - choose one of "norm", "seq" or "rseq". The default is "norm"
<code>control</code>	a control object (S4) of class CovControlMcd-class containing estimation options - same as these provided in the function specification. If the control object is supplied, the parameters from it will be used. If parameters are passed also in the invocation statement, they will override the corresponding elements of the control object.

Details

This function computes the minimum covariance determinant estimator of location and scatter and returns an S4 object of class [CovMcd-class](#) containing the estimates. The implementation of the function is similar to the existing R function [covMcd\(\)](#) which returns an S3 object. The MCD method looks for the $h(> n/2)$ observations (out of n) whose classical covariance matrix has the lowest possible determinant. The raw MCD estimate of location is then the average of these h points, whereas the raw MCD estimate of scatter is their covariance matrix, multiplied by a consistency factor and a finite sample correction factor (to make it consistent at the normal model

and unbiased at small samples). Both rescaling factors are returned also in the vector `raw.cnp2` of length 2. Based on these raw MCD estimates, a reweighting step is performed which increases the finite-sample efficiency considerably - see Pison et al. (2002). The rescaling factors for the reweighted estimates are returned in the vector `cnp2` of length 2. Details for the computation of the finite sample correction factors can be found in Pison et al. (2002). The finite sample corrections can be suppressed by setting `use.correction=FALSE`. The implementation in `rrcov` uses the Fast MCD algorithm of Rousseeuw and Van Driessen (1999) to approximate the minimum covariance determinant estimator.

Value

An S4 object of class `CovNAMcd` which is a subclass of the virtual class `CovNARobust`.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

- V. Todorov, M. Templ and P. Filzmoser. Detection of multivariate outliers in business survey data with incomplete information. *Advances in Data Analysis and Classification*, **5** 37–56, 2011.
- P. J. Rousseeuw and K. van Driessen (1999) A fast algorithm for the minimum covariance determinant estimator. *Technometrics* **41**, 212–223.
- Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

Examples

```
data(bush10)
mcd <- CovNAMcd(bush10)
mcd
summary(mcd)

plot(mcd)
plot(mcd, which="pairs")
plot(mcd, which="xydistance")
plot(mcd, which="xyqqchi2")
```

CovNAMcd-class

MCD Estimates of Multivariate Location and Scatter for incomplete data

Description

This class, derived from the virtual class "CovRobust" accomodates MCD Estimates of multivariate location and scatter computed by the 'Fast MCD' algorithm.

Objects from the Class

Objects can be created by calls of the form `new("CovMcd", ...)`, but the usual way of creating CovMcd objects is a call to the function `CovMcd` which serves as a constructor.

Slots

alpha: Object of class "numeric" - the size of the subsets over which the determinant is minimized (the default is $(n+p+1)/2$)

quan: Object of class "numeric" - the number of observations on which the MCD is based. If `quan` equals `n.obs`, the MCD is the classical covariance matrix.

best: Object of class "Uvector" - the best subset found and used for computing the raw estimates. The size of `best` is equal to `quan`

raw.cov: Object of class "matrix" the raw (not reweighted) estimate of location

raw.center: Object of class "vector" - the raw (not reweighted) estimate of scatter

raw.mah: Object of class "Uvector" - mahalanobis distances of the observations based on the raw estimate of the location and scatter

raw.wt: Object of class "Uvector" - weights of the observations based on the raw estimate of the location and scatter

raw.cnp2: Object of class "numeric" - a vector of length two containing the consistency correction factor and the finite sample correction factor of the raw estimate of the covariance matrix

cnp2: Object of class "numeric" - a vector of length two containing the consistency correction factor and the finite sample correction factor of the final estimate of the covariance matrix.

iter, crit, wt: from the "CovRobust-class" class.

call, cov, center, n.obs, mah, method, singularity, X: from the "Cov-class" class.

Extends

Class "CovRobust-class", directly. Class "Cov-class", by class "CovRobust-class".

Methods

No methods defined with class "CovMcd" in the signature.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[CovMcd](#), [Cov-class](#), [CovRobust-class](#)

Examples

```
showClass("CovNAMcd")
```

CovNAOgk

*Robust Location and Scatter Estimation - Ortoogonalized
Gnanadesikan-Kettenring (OGK) for incomplete data*

Description

Computes a robust multivariate location and scatter estimate with a high breakdown point for incomplete data, using the pairwise algorithm proposed by Marona and Zamar (2002) which in turn is based on the pairwise robust estimator proposed by Gnanadesikan-Kettenring (1972).

Usage

```
CovNAOgk(x, niter = 2, beta = 0.9, impMeth = c("norm" , "seq", "rseq"), control)
```

Arguments

x	a matrix or data frame.
niter	number of iterations, usually 1 or 2 since iterations beyond the second do not lead to improvement.
beta	coverage parameter for the final reweighted estimate
impMeth	select imputation method to use - choose one of "norm" , "seq" or "rseq". The default is "norm"
control	a control object (S4) of class CovControlOgk-class containing estimation options - same as these provided in the function specification. If the control object is supplied, the parameters from it will be used. If parameters are passed also in the invocation statement, they will override the corresponding elements of the control object. The control object contains also functions for computing the robust univariate location and dispersion estimate <code>mrob</code> and for computing the robust estimate of the covariance between two random variables <code>vrob</code> .

Details

The method proposed by Marona and Zamar (2002) allows to obtain positive-definite and almost affine equivariant robust scatter matrices starting from any pairwise robust scatter matrix. The default robust estimate of covariance between two random vectors used is the one proposed by Gnanadesikan and Kettenring (1972) but the user can choose any other method by redefining the function in slot `vrob` of the control object `CovControlOgk`. Similarly, the function for computing the robust univariate location and dispersion used is the tau scale defined in Yohai and Zamar (1998) but it can be redefined in the control object.

The estimates obtained by the OGK method, similarly as in `CovMcd` are returned as 'raw' estimates. To improve the estimates a reweighting step is performed using the coverage parameter `beta` and these reweighted estimates are returned as 'final' estimates.

Value

An S4 object of class [CovNAOgk](#) which is a subclass of the virtual class [CovNARobust](#).

Note

If the user does not specify a scale and covariance function to be used in the computations or specifies one by using the arguments `smrob` and `svrob` (i.e. the names of the functions as strings), a native code written in C will be called which is by far faster than the R version.

If the arguments `mrob` and `vrob` are not NULL, the specified functions will be used via the pure R implementation of the algorithm. This could be quite slow.

See [CovControlOgk](#) for details.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Yohai, R.A. and Zamar, R.H. (1998) High breakdown point estimates of regression by means of the minimization of efficient scale *JASA* **86**, 403–413.

Gnanadesikan, R. and John R. Kettenring (1972) Robust estimates, residuals, and outlier detection with multiresponse data. *Biometrics* **28**, 81–124.

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[CovNAMcd](#)

Examples

```
data(bush10)
CovNAOgk(bush10)

## the following three statements are equivalent
c1 <- CovNAOgk(bush10, niter=1)
c2 <- CovNAOgk(bush10, control = CovControlOgk(niter=1))

## direct specification overrides control one:
c3 <- CovNAOgk(bush10, beta=0.95,
               control = CovControlOgk(beta=0.99))
c1
```

CovNAOgk-class	<i>OGK Estimates of Multivariate Location and Scatter for incomplete data</i>
----------------	---

Description

This class, derived from the virtual class "CovRobust" accomodates OGK Estimates of multivariate location and scatter computed by the algorithm proposed by Marona and Zamar (2002).

Objects from the Class

Objects can be created by calls of the form `new("CovOgk", ...)`, but the usual way of creating CovOgk objects is a call to the function `CovOgk` which serves as a constructor.

Slots

`raw.cov`: Object of class "matrix" the raw (not reweighted) estimate of covariance matrix
`raw.center`: Object of class "vector" - the raw (not reweighted) estimate of the location vector
`raw.mah`: Object of class "Uvector" - mahalanobis distances of the observations based on the raw estimate of the location and scatter
`raw.wt`: Object of class "Uvector" - weights of the observations based on the raw estimate of the location and scatter
`iter, crit, wt`: from the "CovRobust-class" class.
`call, cov, center, n.obs, mah, method, singularity, X`: from the "Cov-class" class.

Extends

Class "CovRobust-class", directly. Class "Cov-class", by class "CovRobust-class".

Methods

No methods defined with class "CovNAOgk" in the signature.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[CovMcd-class](#), [CovMest-class](#)

Examples

```
showClass("CovNAOgk")
```

CovNARobust

Robust Location and Scatter Estimation for incomplete data

Description

Computes a robust multivariate location and scatter estimate with a high breakdown point for incomplete data, using one of the available estimators.

Usage

```
CovNARobust(x, control, impMeth=c("norm" , "seq" , "rseq"))
```

Arguments

x	a matrix or data frame.
control	a control object (S4) for one of the available control classes, e.g. CovControlMcd-class , CovControlOgk-class , CovControlSest-class , etc., containing estimation options. The class of this object defines which estimator will be used. Alternatively a character string can be specified which names the estimator - one of auto, sde, mcd, ogk, m, mve, sfast, surreal, bisquare, rocke. If 'auto' is specified or the argument is missing, the function will select the estimator (see below for details)
impMeth	select imputation method to use - choose one of "norm" , "seq" or "rseq". The default is "norm"

Details

This function is based on imputation and then estimation with a selected high breakdown point method. Thus first imputation with the selected method will be performed and then the function CovRobust will be called. For details see [CovRobust](#).

Value

An object derived from a CovRobust object, depending on the selected estimator.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

V. Todorov, M. Templ and P. Filzmoser. Detection of multivariate outliers in business survey data with incomplete information. *Advances in Data Analysis and Classification*, **5** 37–56, 2011.

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

Examples

```
data(bush10)
CovNARobust(bush10)
CovNARobust(bush10, CovControlSest())
```

CovNARobust-class	<i>Class "CovNARobust" - virtual base class for robust estimates of multivariate location and scatter for incomplete data</i>
-------------------	---

Description

CovNARobust is a virtual base class used for deriving the concrete classes representing different robust estimates of multivariate location and scatter for incomplete data. Here are implemented the standard methods common for all robust estimates like show, summary and plot. The derived classes can override these methods and can define new ones.

Objects from the Class

A virtual Class: No objects may be created from it.

Slots

iter: number of iterations used to compute the estimates
crit: value of the criterion function
wt: weights
call, cov, center, n.obs, mah, method, singularity, X: from the "[Cov-class](#)" class.

Extends

Classes "[CovNA](#)" and "[CovRobust-class](#)", directly.

Methods

plot signature(x = "CovNARobust"): plot the object
summary signature(object = "CovNARobust"): display additional information for the object

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[CovNA](#), [CovNAMcd](#), [CovNA0gk](#), [CovNASde](#) , [CovNASest](#)

Examples

```
data(hbk)
hbk.x <- data.matrix(hbk[, 1:3])
cv <- CovMest(hbk.x)           # it is not possible to create an object of
                                # class CovRobust, since it is a VIRTUAL class

cv
summary(cv)                   # summary method for class CovRobust
plot(cv)                      # plot method for class CovRobust
```

CovNASde	<i>Stahel-Donoho Estimates of Multivariate Location and Scatter for incomplete data</i>
----------	---

Description

Compute a robust estimate of location and scale using the Stahel-Donoho projection based estimator

Usage

```
CovNASde(x, nsamp, maxres, tune = 0.95, eps = 0.5, prob = 0.99,
impMeth = c("norm" , "seq", "rseq"), seed = NULL, trace = FALSE, control)
```

Arguments

x	a matrix or data frame.
nsamp	a positive integer giving the number of resamples required; nsamp may not be reached if too many of the p-subsamples, chosen out of the observed vectors, are in a hyperplane. If nsamp = 0 all possible subsamples are taken. If nsamp is omitted, it is calculated to provide a breakdown point of eps with probability prob.
maxres	a positive integer specifying the maximum number of resamples to be performed including those that are discarded due to linearly dependent subsamples. If maxres is omitted it will be set to 2 times nsamp.
tune	a numeric value between 0 and 1 giving the fraction of the data to receive non-zero weight. Defaults to 0.95
prob	a numeric value between 0 and 1 specifying the probability of high breakdown point; used to compute nsamp when nsamp is omitted. Defaults to 0.99.
impMeth	select imputation method to use - choose one of "norm" , "seq" or "rseq". The default is "norm"
eps	a numeric value between 0 and 0.5 specifying the breakdown point; used to compute nsamp when nresamp is omitted. Defaults to 0.5.

seed	starting value for random generator. Default is seed = NULL.
trace	whether to print intermediate results. Default is trace = FALSE.
control	a control object (S4) of class <code>CovControlSde-class</code> containing estimation options - same as these provided in the function specification. If the control object is supplied, the parameters from it will be used. If parameters are passed also in the invocation statement, they will override the corresponding elements of the control object.

Value

An S4 object of class `CovNASde` which is a subclass of the virtual class `CovNARobust`.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

- R. A. Maronna and V.J. Yohai (1995) The Behavior of the Stahel-Donoho Robust Multivariate Estimator. *Journal of the American Statistical Association* **90** (429), 330–341.
- R. A. Maronna, D. Martin and V. Yohai (2006). *Robust Statistics: Theory and Methods*. Wiley, New York.
- Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

Examples

```
data(bush10)
CovNASde(bush10)

## the following four statements are equivalent
c0 <- CovNASde(bush10)
c1 <- CovNASde(bush10, nsamp=2000)
c2 <- CovNASde(bush10, control = CovControlSde(nsamp=2000))
c3 <- CovNASde(bush10, control = new("CovControlSde", nsamp=2000))

## direct specification overrides control one:
c4 <- CovNASde(bush10, nsamp=100,
               control = CovControlSde(nsamp=2000))

c1
summary(c1)
```

CovNASde-class	<i>Stahel-Donoho Estimates of Multivariate Location and Scatter for incomplete data</i>
----------------	---

Description

This class, derived from the virtual class "CovRobust" accomodates Stahel-Donoho estimates of multivariate location and scatter.

Objects from the Class

Objects can be created by calls of the form `new("CovSde", ...)`, but the usual way of creating CovSde objects is a call to the function `CovSde` which serves as a constructor.

Slots

`iter`, `crit`, `wt`: from the "[CovRobust-class](#)" class.

`call`, `cov`, `center`, `n.obs`, `mah`, `method`, `singularity`, `X`: from the "[Cov-class](#)" class.

Extends

Class "[CovRobust-class](#)", directly. Class "[Cov-class](#)", by class "[CovRobust-class](#)".

Methods

No methods defined with class "CovNASde" in the signature.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[CovSde](#), [Cov-class](#), [CovRobust-class](#)

Examples

```
showClass("CovNASde")
```

Description

Computes S-Estimates of multivariate location and scatter based on Tukey's biweight function for incomplete data using a fast algorithm similar to the one proposed by Salibián-Barrera and Yohai (2006) for the case of regression. Alternativley, the Ruppert's SURREAL algorithm, bisquare or Rocke type estimation can be used.

Usage

```
CovNASest(x, bdp = 0.5, arp = 0.1, eps = 1e-5, maxiter = 120,
          nsamp = 500, impMeth = c("norm", "seq", "rseq"), seed = NULL,
          trace = FALSE, tolSolve = 1e-13,
          scalefn,
          method = c("sfast", "surreal", "bisquare", "rocke", "suser", "sdet"), control,
          t0, S0, initcontrol)
```

Arguments

x	a matrix or data frame.
bdp	a numeric value specifying the required breakdown point. Allowed values are between $(n - p) / (2 * n)$ and 1 and the default is <code>bdp=0.5</code> .
arp	a numeric value specifying the asymptotic rejection point (for the Rocke type S estimates), i.e. the fraction of points receiving zero weight (see Rocke (1996)). Default is <code>arp=0.1</code> .
eps	a numeric value specifying the relative precision of the solution of the S-estimate (bisquare and Rocke type). Default is to <code>eps=1e-5</code> .
maxiter	maximum number of iterations allowed in the computation of the S-estimate (bisquare and Rocke type). Default is <code>maxiter=120</code> .
nsamp	the number of random subsets considered. Default is <code>nsamp = 500</code> .
impMeth	select imputation method to use - choose one of "norm", "seq" or "rseq". The default is "norm"
seed	starting value for random generator. Default is <code>seed = NULL</code> .
trace	whether to print intermediate results. Default is <code>trace = FALSE</code> .
tolSolve	numeric tolerance to be used for inversion (solve) of the covariance matrix in mahalanobis .
scalefn	function to compute a robust scale estimate or character string specifying a rule determining such a function. Used for computing the "deterministic" S-estimates (<code>method="sdet"</code>). If <code>scalefn</code> is missing or is <code>NULL</code> , the function is selected depending on the data set size, following the recommendation of Hubert et al. (2012) - Qn if $n \leq 1000$ and scaleTau2 otherwise.

method	Which algorithm to use: 'sfast'=FAST-S, 'surreal'=SURREAL, 'bisquare', 'rocke' or 'sdet', which will invoke the deterministic algorithm of Hubert et al. (2012).
control	a control object (S4) of class <code>CovControlSest-class</code> containing estimation options - same as these provided in the function specification. If the control object is supplied, the parameters from it will be used. If parameters are passed also in the invocation statement, they will override the corresponding elements of the control object.
t0	optional initial HBDP estimate for the center
S0	optional initial HBDP estimate for the covariance matrix
initcontrol	optional control object to be used for computing the initial HBDP estimates

Details

Computes biweight multivariate S-estimator of location and scatter. The computation will be performed by one of the following algorithms:

FAST-S An algorithm similar to the one proposed by Salibián-Barrera and Yohai (2006) for the case of regression

SURREAL Ruppert's SURREAL algorithm when method is set to 'surreal'

BISQUARE Bisquare S-Estimate with method set to 'bisquare'

ROCKE Rocke type S-Estimate with method set to 'rocke'.

Value

An S4 object of class `CovNASest` which is a subclass of the virtual class `CovNARobust`.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>, Matias Salibián-Barrera <matias@stat.ubc.ca> and Victor Yohai <vyohai@dm.uba.ar>. See also the code from Kristel Joossens, K.U. Leuven, Belgium and Ella Roelant, Ghent University, Belgium.

References

- M. Salibián-Barrera and V. Yohai (2006) A fast algorithm for S-regression estimates, *Journal of Computational and Graphical Statistics*, **15**, 414–427.
- R. A. Maronna, D. Martin and V. Yohai (2006). *Robust Statistics: Theory and Methods*. Wiley, New York.
- Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

Examples

```
library(rrcov)
data(bush10)
CovNASest(bush10)

## the following four statements are equivalent
```

```

c0 <- CovNASest(bush10)
c1 <- CovNASest(bush10, bdp = 0.25)
c2 <- CovNASest(bush10, control = CovControlSest(bdp = 0.25))
c3 <- CovNASest(bush10, control = new("CovControlSest", bdp = 0.25))

## direct specification overrides control one:
c4 <- CovNASest(bush10, bdp = 0.40,
                control = CovControlSest(bdp = 0.25))

c1
summary(c1)

## Use the SURREAL algorithm of Ruppert
cr <- CovNASest(bush10, method="surreal")
cr

## Use Bisquare estimation
cr <- CovNASest(bush10, method="bisquare")
cr

## Use Rocke type estimation
cr <- CovNASest(bush10, method="rocke")
cr

```

CovNASest-class

S Estimates of Multivariate Location and Scatter for incomplete data

Description

This class, derived from the virtual class "CovRobust" accomodates S Estimates of multivariate location and scatter computed by the 'Fast S' or 'SURREAL' algorithm.

Objects from the Class

Objects can be created by calls of the form `new("CovSest", ...)`, but the usual way of creating CovSest objects is a call to the function `CovSest` which serves as a constructor.

Slots

`iter`, `crit`, `wt`: from the "CovRobust-class" class.

`call`, `cov`, `center`, `n.obs`, `mah`, `method`, `singularity`, `X`: from the "Cov-class" class.

Extends

Class "CovRobust-class", directly. Class "Cov-class", by class "CovRobust-class".

Methods

No methods defined with class "CovNASest" in the signature.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[CovSest](#), [Cov-class](#), [CovRobust-class](#)

Examples

```
showClass("CovNASest")
```

impNorm

Impute missing multivariate normal data

Description

Draws missing elements of a data matrix under the multivariate normal model and a user-supplied parameter

Usage

```
impNorm(x)
```

Arguments

x the original incomplete data matrix.

Details

This function simply uses `imp.norm` from package `norm`.

Value

a matrix of the same form as `x`, but with all missing values filled in with simulated values drawn from their predictive distribution given the observed data and the specified parameter.

References

See Section 5.4.1 of Schafer (1996).

See Also

[prelim.norm](#), [makeparam.norm](#), and [rngseed](#).

Examples

```
data(bush10)
impNorm(bush10) #impute missing data under the MLE
```

impSeq

Sequential imputation of missing values

Description

Impute missing multivariate data using sequential algorithm

Usage

```
impSeq(x, norm_impute=FALSE, check_data=FALSE, verbose=TRUE)
```

Arguments

x	the original incomplete data matrix.
norm_impute	If there are not enough complete observations and norm_impute=TRUE the data matrix will be imputed using the multivariate normal model, the outlyingness will be computed for the completed matrix and the result will be returned. If norm_impute=FALSE (the default) the algorithm will try to impute sufficient number of observations using the multivariate normal model and then will start the sequential imputation.
check_data	whether to check the variables: only numeric, non discrete, with less than 50% NAs and with non-zero MAD. The default is check_data=FALSE, meaning that no checks will be performed.
verbose	whether to write messages about the checking of the data. By default verbose=TRUE, i.e. messages will be written.

Details

SEQimpute starts from a complete subset of the data set X_c and estimates sequentially the missing values in an incomplete observation, say x^* , by minimizing the determinant of the covariance of the augmented data matrix $X^* = [X_c; x^*]$. Then the observation x^* is added to the complete data matrix and the algorithm continues with the next observation with missing values.

Value

A list containing the following elements:

x	a matrix of the same form as x, but with all missing values filled in sequentially.
colInAnalysis	the column indices of the columns used in the analysis.
namesNotNumeric	the names of the variables which are not numeric.

namesNAcol names of the columns left out due to too many NA's.
 namesDiscrete names of the discrete variables.
 namesZeroScale names of the variables with zero scale.

References

S. Verboven, K. Vanden Branden and P. Goos (2007). Sequential imputation for missing values. *Computational Biology and Chemistry*, **31**, 320–327.

Examples

```
data(bush10)
impSeq(bush10) # impute sequentially missing data
```

impSeqRob	<i>Robust sequential imputation of missing values</i>
-----------	---

Description

Impute missing multivariate data using robust sequential algorithm

Usage

```
impSeqRob(x, alpha=0.9, norm_impute=FALSE, check_data=FALSE, verbose=TRUE)
```

Arguments

x	the original incomplete data matrix.
alpha	the number of regular genes, (1-alpha) measures the fraction of outliers the algorithm should resist. Any value between 0.5 and 1 may be specified. The default is alpha=0.9.
norm_impute	If there are not enough complete observations and norm_impute=TRUE the data matrix will be imputed using the multivariate normal model, the outlyingness will be computed for the completed matrix and the result will be returned. If norm_impute=FALSE (the default) the algorithm will try to impute sufficient number of observations using the multivariate normal model and then will start the sequential imputation.
check_data	whether to check the variables: only numeric, non discrete, with less than 50% NAs and with non-zero MAD. The default is check_data=FALSE, meaning that no checks will be performed.
verbose	whether to write messages about the checking of the data. By default verbose=TRUE, i.e. messages will be written.

Details

The nonrobust version SEQimpute starts from a complete subset of the data set X_c and estimates sequentially the missing values in an incomplete observation, say x^* , by minimizing the determinant of the covariance of the augmented data matrix $X^* = [X_c; x^*]$. Then the observation x^* is added to the complete data matrix and the algorithm continues with the next observation with missing values. Since SEQimpute uses the sample mean and covariance matrix it will be vulnerable to the influence of outliers and it is improved by plugging in robust estimators of location and scatter. One possible solution is to use the outlyingness measure as proposed by Stahel (1981) and Donoho (1982) and successfully used for outlier identification in Hubert et al. (2005). We can compute the outlyingness measure for the complete observations only but once an incomplete observation is imputed (sequentially) we could compute the outlyingness measure for it too and use it to decide if this observation is an outlier or not. If the outlyingness measure does not exceed a predefined threshold the observation is included in the further steps of the algorithm.

Value

A list containing the following elements:

<code>x</code>	a matrix of the same form as <code>x</code> , but with all missing values filled in sequentially.
<code>outl</code>	outlyingness computed for all observations.
<code>flag</code>	flag of outliers.
<code>colInAnalysis</code>	the column indices of the columns used in the analysis.
<code>namesNotNumeric</code>	the names of the variables which are not numeric.
<code>namesNAcol</code>	names of the columns left out due to too many NA's.
<code>namesDiscrete</code>	names of the discrete variables.
<code>namesZeroScale</code>	names of the variables with zero scale.

References

- S. Verboven, K. Vanden Branden and P. Goos (2007). Sequential imputation for missing values. *Computational Biology and Chemistry*, **31**, 320–327.
- K. Vanden Branden and S. Verboven (2009). Robust Data Imputation. *Computational Biology and Chemistry*, **33**, 7–13.

Examples

```
data(bush10)
impSeqRob(bush10) # impute sequentially missing data
```

PcaNA

*Classical or robust Principal Components for incomplete data***Description**

Computes classical and robust principal components for incomplete data using an EM algorithm as described by Serneels and Verdonck (2008)

Usage

```
PcaNA(x, ...)
## Default S3 method:
PcaNA(x, k = ncol(x), kmax = ncol(x), conv=1e-10, maxiter=100,
      method=c("cov", "locantore", "hubert", "grid", "proj", "class"), cov.control=NULL,
      scale = FALSE, signflip = TRUE, crit.pca.distances = 0.975, trace=FALSE, ...)
## S3 method for class 'formula'
PcaNA(formula, data = NULL, subset, na.action, ...)
```

Arguments

formula	a formula with no response variable, referring only to numeric variables.
data	an optional data frame (or similar: see model.frame) containing the variables in the formula formula.
subset	an optional vector used to select rows (observations) of the data matrix x.
na.action	a function which indicates what should happen when the data contain NAs. The default is set by the na.action setting of options , and is na.fail if that is unset. The default is na.omit .
...	arguments passed to or from other methods.
x	a numeric matrix (or data frame) which provides the data for the principal components analysis.
k	number of principal components to compute. If k is missing, or k = 0, the algorithm itself will determine the number of components by finding such k that $l_k/l_1 \geq 10.E - 3$ and $\sum_{j=1}^k l_j / \sum_{j=1}^r l_j \geq 0.8$. It is preferable to investigate the scree plot in order to choose the number of components and then run again. Default is k=ncol(x).
kmax	maximal number of principal components to compute. Default is kmax=10. If k is provided, kmax does not need to be specified, unless k is larger than 10.
conv	convergence criterion for the EM algorithm. Default is conv=1e-10.
maxiter	maximal number of iterations for the EM algorithm. Default is maxiter=100.
method	which PC method to use (classical or robust) - "class" means classical PCA and one of the following "locantore", "hubert", "grid", "proj", "cov" specifies a robust PCA method. If the method is "cov" - i.e. PCA based on a robust covariance matrix - the argument cov.control can specify which method for computing the (robust) covariance matrix will be used. Default is method="locantore".

<code>cov.control</code>	control object in case of robust PCA based on a robust covariance matrix.
<code>scale</code>	a logical value indicating whether the variables should be scaled to have unit variance (only possible if there are no constant variables). As a scale function <code>mad</code> is used but alternatively, a vector of length equal the number of columns of <code>x</code> can be supplied. The value is passed to <code>scale</code> and the result of the scaling is stored in the <code>scale</code> slot. Default is <code>scale = FALSE</code>
<code>signflip</code>	a logical value indicating wheather to try to solve the sign indeterminacy of the loadings - ad hoc approach setting the maximum element in a singular vector to be positive. Default is <code>signflip = FALSE</code>
<code>crit.pca.distances</code>	criterion to use for computing the cutoff values for the orthogonal and score distances. Default is 0.975.
<code>trace</code>	whether to print intermediate results. Default is <code>trace = FALSE</code>

Details

PcaNA, serving as a constructor for objects of class `PcaNA` is a generic function with "formula" and "default" methods. For details see the relevant references.

Value

An S4 object of class `PcaNA` which is a subclass of the virtual class `Pca-class`.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Serneels S & Verdonck T (2008), Principal component analysis for data containing outliers and missing elements. *Computational Statistics and Data Analysis*, **52**(3), 1712–1727 .

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

Examples

```
## 1. With complete data
## PCA of the bushfire data
data(bushfire)
pca <- PcaNA(bushfire)
pca

## Compare with the classical PCA
prcomp(bushfire)

## or
PcaNA(bushfire, method="class")

## If you want to print the scores too, use
```

```

    print(pca, print.x=TRUE)

## Using the formula interface
    PcaNA(~., data=bushfire)

## To plot the results:

    plot(pca)                # distance plot
    pca2 <- PcaNA(bushfire, k=2)
    plot(pca2)               # PCA diagnostic plot (or outlier map)

## Use the standard plots available for for prcomp and princomp
    screeplot(pca)
    biplot(pca)

#####
## 2. Now the same wit incomplete data - bush10
    data(bush10)
    pca <- PcaNA(bush10)
    pca

## Compare with the classical PCA
    PcaNA(bush10, method="class")

## If you want to print the scores too, use
    print(pca, print.x=TRUE)

## Using the formula interface
    PcaNA(~., data=as.data.frame(bush10))

## To plot the results:

    plot(pca)                # distance plot
    pca2 <- PcaNA(bush10, k=2)
    plot(pca2)               # PCA diagnostic plot (or outlier map)

## Use the standard plots available for for prcomp and princomp
    screeplot(pca)
    biplot(pca)

```

PcaNA-class

Class "PcaNA" Principal Components for incomplete data

Description

Contains the results of the computations of classical and robust principal components for incomplete data using an EM algorithm as described by Serneels and Verdonck (2008)

Objects from the Class

Objects can be created by calls of the form `new("PcaNA", ...)` but the usual way of creating PcaNA objects is a call to the function `PcaNA` which serves as a constructor.

Slots

`call`, `center`, `scale`, `loadings`, `eigenvalues`, `scores`, `k`, `sd`, `od`, `cutoff.sd`, `cutoff.od`, `flag`, `n.obs`:
from the "[Pca-class](#)" class.

`Ximp`: the data matrix with imputed missing values

Extends

Class "[Pca-class](#)", directly.

Methods

getQuan signature(`obj` = "PcaNA"): ...

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Serneels S & Verdonck T (2008), Principal component analysis for data containing outliers and missing elements. *Computational Statistics and Data Analysis*, **52**(3), 1712–1727 .

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[PcaRobust-class](#), [Pca-class](#), [PcaClassic](#), [PcaClassic-class](#)

Examples

```
showClass("PcaNA")
```

SummaryCovNA-class	Class " <i>SummaryCovNA</i> " - summary of " <i>CovNA</i> " objects
--------------------	---

Description

The "CovNA" object plus some additional summary information

Objects from the Class

Objects can be created by calls of the form `new("SummaryCovNA", ...)`, but most often by invoking 'summary' on a "CovNA" object. They contain values meant for printing by 'show'.

Slots

No Slots defined with class "SummaryCovNA" in the signature.

Methods

No Methods defined with class "SummaryCovNA" in the signature.

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[CovNA](#)

Examples

```
showClass("SummaryCovNA")
```

SummaryCovNARobust-class

Class "SummaryCovNARobust" - summary of "CovNARobust" objects

Description

Summary information for CovRobust objects means for printing by 'show'

Objects from the Class

Objects can be created by calls of the form `new("SummaryCovNARobust", ...)`, but most often by invoking 'summary' on an "CovNA" object. They contain values meant for printing by 'show'.

Slots

No Slots defined with class "SummaryCovNARobust" in the signature.

Extends

Class "SummaryCovNA", directly.

Methods

```
show signature(object = "SummaryCovNARobust"): ...
```

Author(s)

Valentin Todorov <valentin.todorov@chello.at>

References

Todorov V & Filzmoser P (2009), An Object Oriented Framework for Robust Multivariate Analysis. *Journal of Statistical Software*, **32**(3), 1–47. <doi:10.18637/jss.v032.i03>.

See Also

[CovRobust-class](#), [SummaryCov-class](#)

Examples

```
data(hbk)
hbk.x <- data.matrix(hbk[, 1:3])
cv <- CovMest(hbk.x)
cv
summary(cv)
```

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