Package 'RobStatTM'

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Version 1.0.5 Date 2022-04-18 Title Robust Statistics: Theory and Methods Description Companion package for the book: "Robust Statistics: Theory and Methods, second edition", http://www.wiley.com/go/maronna/robust. This package contains code that implements the robust estimators discussed in the recent second edition of the book above, as well as the scripts reproducing all the examples in the book. **Depends** R (>= 3.5.0) **Imports** stats, graphics, utils, methods, pyinit, rrcov, robustbase Suggests R.rsp LazyData yes **License** GPL (>= 3) RoxygenNote 7.1.1 **Encoding UTF-8** VignetteBuilder R.rsp **NeedsCompilation** yes Author Matias Salibian-Barrera [cre], Victor Yohai [aut], Ricardo Maronna [aut], Doug Martin [aut], Gregory Brownson [aut] (ShinyUI), Kjell Konis [aut], Kjell Konis [cph] (erfi), Christophe Croux [ctb] (WBYlogreg, BYlogreg), Gentiane Haesbroeck [ctb] (WBYlogreg, BYlogreg), Martin Maechler [cph] (lmrob.fit, lmrob..M..fit, lmrob.S), Manuel Koller [cph] (lmrob.fit, .vcov.avar1, lmrob.S, lmrob.lar), Matias Salibian-Barrera [aut] Maintainer Matias Salibian-Barrera <matias@stat.ubc.ca> Repository CRAN

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Description

This data set contains physicochemical characteristics of 44 aliphatic alcohols. The aim of the experiment was the prediction of the solubility on the basis of molecular descriptors.

Usage

data(alcohol)

Format

An object of class "data.frame".

Details

Format: 44 cases and 7 continuous variables. The columns are: 1. SAG=solvent accessible surface-bounded molecular volume 2. V=volume 3. log PC (PC=octanol-water partitions coefficient) 4. P=polarizability 5. RM=molar refractivity 6. Mass 7. log(Solubility) (response)

Source

Romanelli, G.P., Martino, C.M. and Castro, E.A. (2001), Modeling the solubility of aliphatic alcohols via molecular descriptors, Journal of the Chemical Society of Pakistan, 23, 195-199.

Examples

data(alcohol)

4 algae

algae

Algae data

Description

Each row of the data set is a set of 90 measurements at a river in some place in Europe. There are 11 predictors. The response is the logarithm of the abundance of a certain class of algae. Description: The columns are: 1. season, categorical (1,2,3,4 for winter, spring, summer and autumn) 2. river size (categorical) (1,2,3 for small, medium and large) 3. fluid velocity (categorical) (1,2,3 for low, medium and high) 4-11 (numerci): content of nitrogen in the form of nitrates, nitrites and ammonia, and other chemical compounds. Col. 12 ia the response: abundance of a type of algae (type 6 in the complete file). For simplicity we deleted the rows with missing values and took the logarithm of the response.

Usage

data(algae)

Format

An object of class "data.frame".

Details

Format 90 rows, 12 columns (3 categorical, 9 numeric)

Source

Hettich, S. and Bay, S.D. (1999), The UCI KDD Archive http://kdd.ics.uci.edu. Irvine, CA: University of California, Department of Information and Computer Science.

References

References go here.

Examples

data(algae)

biochem 5

biochem

Biochem data

Description

Two biochemical measurements on 12 men with similar weights.

Usage

```
data(biochem)
```

Format

An object of class "data.frame".

Details

Format: Numeric, 12 rows, two columns

Source

Seber, G.A.F. (1984), Multivariate Observations. New York: John Wiley.

Examples

data(biochem)

bisquare

Tuning parameter the rho loss functions

Description

This function computes the tuning constant that yields an MM-regression estimator with a desired asymptotic efficiency when computed with a rho function in the corresponding family. The output of this function can be passed to the functions lmrobdet.control, mscale and rho.

Usage

bisquare(e)

Arguments

e

the desired efficiency of the corresponding regression estimator for Gaussian errors

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Value

A length-1 vector with the corresponding tuning constant.

Author(s)

Kjell Konis

Examples

Tuning parameters for an 85%-efficient M-estimator at a Gaussian model bisquare(.85)

breslow.dat

Breslow Data

Description

Patients suffering from simple or complex partial seizures were randomized to receive either the antiepileptic drug progabide or a placebo. At each of four successive post randomization clinic visits, the number of seizures occuring over the previous two weeks was reported.

Usage

data(breslow.dat)

Format

An object of class "data.frame".

Details

Description: A data frame with 59 observations on the following 12 variables: ID: an integer value specifying the patient identification number; Y1: an integer value, the number of seizures during the first two week period; Y2: an integer value, the number of seizures during the second two week period; Y3: an integer value, the number of seizures during the third two week period. Y4: an integer value, the number of seizures during the fourth two week period; Base: an integer value giving the eight-week baseline seizure count; Age: an integer value giving the age of the parient in years; Trt: the treatment: a factor with levels placebo and progabide; Ysum: an integer value, the sum of Y1, Y2, Y3 and Y4; sumY: an integer value, the sum of Y1, Y2, Y3 and Y4; Age10: a numeric value, Age divided by 10; Base4: a numeric value, Base divided by 4.

Format: Numeric, 59 rows and 12 columns.

Source

Breslow, N. E., and Clayton, D. G. (1993), "Approximate Inference in Generalized Linear Mixed Models," Journal of the American Statistical Association, Vol. 88, No. 421, pp. 9-25.

Thrall, P. F., and Vail, S. C. (1990), "Some Covariance Models for Longitudinal Count Data With Overdispersion," Biometrics, Vol. 46, pp. 657-671.

bus 7

Examples

data(breslow.dat)

bus Bus data

Description

This data set corresponds to a study in automatic vehicle recognition. Each of the 218 rows corresponds to a view of a bus silhouette, and contains 18 attributes of the image. It was decided to exclude variable 9 and divide the remaining variables by their MADN's.

Usage

data(bus)

Format

An object of class "data.frame".

Details

Description: The following features were extracted from the silhouettes. 1. compactness 2. circularity 3. distance circularity 4. radius ratio 5. principal axis aspect ratio 6. maximum length aspect ratio 7. scatter ratio 8. elongatedness 9. principal axis rectangularity 10. maximum length rectangularity 11. scaled variance along major axis 12. scaled variance along minor axis 13. scaled radius of gyration 14. skewness about major axis 15. skewness about minor axis 16. kurtosis about minor axis 17. kurtosis about major axis 18. hollows ratio

Format: Numeric, 218 rows and 18 columns.

Source

Hettich, S. and Bay, S.D. (1999), The UCI KDD Archive http://kdd.ics.uci.edu. Irvine, CA: University of California, Department of Information and Computer Science.

Examples

data(bus)

8 covClassic

cov.dcml	Approximate covariance matrix of the DCML regression estimator.

Description

The estimated covariance matrix of the DCML regression estimator. This function is used internally and not meant to be used directly.

Usage

```
cov.dcml(res.LS, res.R, CC, sig.R, t0, p, n, control)
```

Arguments

res.LS	vector of residuals from the least squares fit
res.R	vector of residuals from the robust regression fit
CC	estimated covariance matrix of the robust regression estimator
sig.R	robust estimate of the scale of the residuals
t0	mixing parameter
p, n	the dimensions of the problem, needed for the finite sample correction of the tuning constant of the M-scale
control	a list of control parameters as returned by lmrobdet.control

Value

The covariance matrix estimate.

Author(s)

Victor Yohai, <victoryohai@gmail.com>

covClassic	Classical Covariance Estimation	

Description

Compute an estimate of the covariance/correlation matrix and location vector using classical methods.

covClassic 9

Usage

```
covClassic(
  data,
  corr = FALSE,
  center = TRUE,
  distance = TRUE,
  na.action = na.fail,
  unbiased = TRUE
)
```

Arguments

data a numeric matrix or data frame containing the data.

corr a logical flag. If corr = TRUE then the estimated correlation matrix is computed.

center a logical flag or a numeric vector of length p (where p is the number of columns

of x) specifying the center. If center = TRUE then the center is estimated. Oth-

erwise the center is taken to be 0.

distance a logical flag. If distance = TRUE the Mahalanobis distances are computed.

na.action a function to filter missing data. The default na.fail produces an error if miss-

ing values are present. An alternative is na.omit which deletes observations

that contain one or more missing values.

unbiased a logical flag. If TRUE the unbiased estimator is returned (computed with de-

nominator equal to n-1), else the MLE (computed with denominator equal to n)

is returned.

Details

Its main intention is to return an object compatible to that produced by covRob, but fit using classical methods.

Value

a list with class "covClassic" containing the following elements:

center a numeric vector containing the estimate of the location vector.

cov a numeric matrix containing the estimate of the covariance matrix.

cor a numeric matrix containing the estimate of the correlation matrix if the argu-

ment corr = TRUE. Otherwise it is set to NULL.

dist a numeric vector containing the squared Mahalanobis distances. Only present if

distance = TRUE in the call.

call an image of the call that produced the object with all the arguments named. The

matched call.

Note

Originally, and in S-PLUS, this function was called cov; it has been renamed, as that did mask the function in the standard package **stats**.

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Examples

```
data(wine)
round( covClassic(wine)$cov, 2)
```

covRob

Robust multivariate location and scatter estimators

Description

This function computes robust estimators for multivariate location and scatter.

Usage

```
covRob(X, type = "auto", maxit = 50, tol = 1e-04, corr = FALSE)
```

Arguments

X a data matrix with observations in rows.

type a string indicating which estimator to compute. Valid options are "Rocke" for

Rocke's S-estimator, "MM" for an MM-estimator with a SHR rho function, or "auto" (default) which selects "Rocke" if the number of variables is greater than

or equal to 10, and "MM" otherwise.

maxit Maximum number of iterations, defaults to 50.

tol Tolerance for convergence, defaults to 1e-4.

corr A logical value. If TRUE a correlation matrix is included in the element cor of

the returned object. Defaults to FALSE.

Details

This function computes robust estimators for multivariate location and scatter. The default behaviour (type = "auto") computes a "Rocke" estimator (as implemented in covRobRocke) if the number of variables is greater than or equal to 10, and an MM-estimator with a SHR rho function (as implemented in covRobMM) otherwise.

Value

A list with class "covClassic" with the following components:

center	The location estimate.
cov	The scatter matrix estimate, scaled for consistency at the normal distribution.
cor	The correlation matrix estimate, if the argument cor equals TRUE. Otherwise it is set to NULL.

dist Robust Mahalanobis distances

wts weights

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call an image of the call that produced the object with all the arguments named. The

matched call.

mu The location estimate. Same as center above.

V The scatter matrix estimate, scaled for consistency at the normal distribution.

Same as cov above.

Author(s)

Ricardo Maronna, <rmaronna@retina.ar>

References

```
http://www.wiley.com/go/maronna/robust
```

See Also

```
covRobRocke, covRobMM
```

Examples

```
data(bus)
X0 <- as.matrix(bus)
X1 <- X0[,-9]
tmp <- covRob(X1)
round(tmp$cov[1:10, 1:10], 3)
tmp$mu</pre>
```

covRobMM

MM robust multivariate location and scatter estimator

Description

This function computes an MM robust estimator for multivariate location and scatter with the "SHR" loss function.

Usage

```
covRobMM(X, maxit = 50, tolpar = 1e-04, corr = FALSE)
```

Arguments

X a data matrix with observations in rows.

maxit Maximum number of iterations.
tolpar Tolerance to decide converngence.

corr A logical value. If TRUE a correlation matrix is included in the element cor of

the returned object. Defaults to FALSE.

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Details

This function computes an MM robust estimator for multivariate location and scatter with the "SHR" loss function.

Value

A list with class "covRob" containing the following elements

center	The location estimate.
cov	The scatter matrix estimate, scaled for consistency at the normal distribution. Same as $\mbox{\sf V}$ above.
cor	The correlation matrix estimate, if the argument cor equals TRUE. Otherwise it is set to $\ensuremath{NULL}.$
dist	Robust Mahalanobis distances
wts	weights
call	an image of the call that produced the object with all the arguments named. The matched call.
mu	The location estimate. Same as center above.
V	The scatter or correlation matrix estimate, scaled for consistency at the normal distribution

Author(s)

References

http://www.wiley.com/go/maronna/robust

Examples

```
data(bus)
X0 <- as.matrix(bus)
X1 <- X0[,-9]
tmp <- covRobMM(X1)
round(tmp$cov[1:10, 1:10], 3)
tmp$mu</pre>
```

covRobRocke 13

covRobRocke Rocke's robust multivariate location and scatter estimator
--

Description

This function computes Rocke's robust estimator for multivariate location and scatter.

Usage

```
covRobRocke(
   X,
   initial = "K",
   maxsteps = 5,
   propmin = 2,
   qs = 2,
   maxit = 50,
   tol = 1e-04,
   corr = FALSE
)
```

Arguments

Χ	a data matrix with observations in rows.
initial	A character indicating the initial estimator. Valid options are 'K' (default) for the Pena-Prieto 'KSD' estimate, and 'mve' for the Minimum Volume Ellipsoid.
maxsteps	Maximum number of steps for the line search section of the algorithm.
propmin	Regulates the proportion of weights computed from the initial estimator that will be different from zero. The number of observations with initial non-zero weights will be at least p (the number of columns of X) times propmin.
qs	Tuning paramater for Rocke's loss functions.
maxit	Maximum number of iterations.
tol	Tolerance to decide converngence.
corr	A logical value. If TRUE a correlation matrix is included in the element cor of the returned object. Defaults to FALSE.

Details

This function computes Rocke's robust estimator for multivariate location and scatter.

Value

A list with class "covRob" containing the following elements:

center The location estimate.

cov The scatter matrix estimate, scaled for consistency at the normal distribution.

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cor	The correlation matrix estimate, if the argument cor equals TRUE. Otherwise it is set to NULL.
dist	Robust Mahalanobis distances.
wts	weights
call	an image of the call that produced the object with all the arguments named. The matched call.
mu	The location estimate. Same as center above.
٧	The scatter (or correlation) matrix estimate, scaled for consistency at the normal distribution. Same as cov above.
sig	sig
gamma	Final value of the constant gamma that regulates the efficiency.

Author(s)

Ricardo Maronna, <rmaronna@retina.ar>

References

http://www.wiley.com/go/maronna/robust

Examples

```
data(bus)
X0 <- as.matrix(bus)
X1 <- X0[,-9]
tmp <- covRobRocke(X1)
round(tmp$cov[1:10, 1:10], 3)
tmp$mu</pre>
```

DCML

DCML regression estimator

Description

This function computes the DCML regression estimator. This function is used internally by lmrobdetDCML, and not meant to be used directly.

```
DCML(x, y, z, z0, control)
```

drop1.lmrobdetMM

Arguments

X	design matrix
у	response vector
z	robust fit as returned by MMPY or SMPY
z0	least squares fit as returned by lm.fit
control	a list of control parameters as returned by lmrobdet.control

Value

a list with the following components

coefficients the vector of regression coefficients

cov the estimated covariance matrix of the DCML regression estimator

residuals the vector of regression residuals from the DCML fit

scale a robust residual (M-)scale estimate

to the mixing proportion between the least squares and robust regression estimators

Author(s)

Victor Yohai, <victoryohai@gmail.com>, Matias Salibian-Barrera, <matias@stat.ubc.ca>

References

```
http://www.wiley.com/go/maronna/robust
```

See Also

```
DCML, MMPY, SMPY
```

drop1.lmrobdetMM R	RFPE of submodels of an 1mrobdetMM fit
--------------------	--

Description

This function computes the RFPE for the MM-estimators obtained with lmrobdetMM by recomputing it, successively removing each of a number of specified terms. It is used internally by step.lmrobdetMM and not meant to be used directly.

```
## S3 method for class 'lmrobdetMM'
drop1(object, scope, scale, keep, ...)
```

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Arguments

object	the MM element (of class lmrob) in an object of class lmrobdetMM.
scope	an optional formula giving the terms to be considered for dropping. Typically this argument is omitted, in which case all possible terms are dropped (without breaking hierarchy rules). The scope can also be a character vector of term labels. If the argument is supplied as a formula, any . is interpreted relative to the formula implied by the object argument.
scale	an optional residual scale estimate. If missing the residual scale estimate in object is used.
keep	a character vector of names of components that should be saved for each subset model. Only names from the set "coefficients", "fitted" and "residuals" are allowed. If keep == TRUE, the complete set is saved. The default behavior is not to keep anything.
	additional parameters to match generic method drop1

Value

An anova object consisting of the term labels, the degrees of freedom, and Robust Final Prediction Errors (RFPE) for each subset model. If keep is missing, the anova object is returned. If keep is present, a list with components "anova" and "keep" is returned. In this case, the "keep" component is a matrix of mode "list", with a column for each subset model, and a row for each component kept.

Author(s)

Victor Yohai, <victoryohai@gmail.com>, Matias Salibian-Barrera, <matias@stat.ubc.ca>

References

http://www.wiley.com/go/maronna/robust

See Also

1mrobdetMM

fastmve	Minimum Volume Ellipsoid covariance estimator	

Description

This function uses a fast algorithm to compute the Minimum Volume Ellipsoid (MVE) for multivariate location and scatter.

```
fastmve(x, nsamp = 500)
```

fastmve 17

Arguments

x data matrix (n x p) with cases stored in rows.

nsamp number of random starts for the iterative algorithm, these are constructed using

subsamples of the data.

Details

This function computes the Minimum Volume Ellipsoid (MVE) for multivariate location and scatter, using a fast algorithm related to the fast algorithm for S-regression estimators (see lmrob).

Value

A list with the following components:

center a vector with the robust multivariate location estimate

cov a matrix with the robust covariance / scatter matrix estimate

scale A scalar that equals the median of the mahalanobis distances of the data to the center, multiplied by the determinant of the covariance matrix to the power 1/p

best Indices of the observations that correspond to the MVE estimator

nsamp Number of random starts used for the iterative algorithm

Number of random subsamples (among the nsamp attempted) that failed (result-

ing in singular initial values)

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>

References

http://www.wiley.com/go/maronna/robust

Examples

```
data(bus)
X0 <- as.matrix(bus)
X1 <- X0[,-9]
tmp <- fastmve(X1)
round(tmp$cov[1:10, 1:10], 3)
tmp$center</pre>
```

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flour

Flour data

Description

Determinations of the copper content in wholemeal flour (in parts per million), sorted in ascending order. Format: numeric vector of size 24.

Usage

```
data(flour)
```

Format

An object of class "data.frame".

Source

Analytical Methods Committee (1989), Robust statistics-How not to reject outliers, Analyst, 114, 1693-1702.

References

References go here.

Examples

data(flour)

glass

Glass data

Description

Measurements of the presence of seven chemical constituents in 76 pieces of glass from nonfloat car windows.

Usage

```
data(glass)
```

Format

An object of class "data.frame".

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Details

Format: 76 cases and 7 continuous variables. Description: The columns are: 1. RI refractive index 2. Na2O sodium oxide (unit measurement: weight percent in corresponding oxide, as are the rest of attributes) 3. MgO magnesium oxide 4. Al2O3 aluminum oxide 5. SiO2 silcon oxide 6. K2O potassium oxide 7. CaO calcium oxide

Source

Hettich, S. and Bay, S.D. (1999), The UCI KDD Archive http://kdd.ics.uci.edu, Irvine, CA: University of California, Department of Information and Computer Science.

Examples

data(glass)

hearing

Hearing data

Description

Prevalence rates in percent for men aged 55–64 with hearing levels 16 decibels or more above the audiometric zero.

Usage

data(hearing)

Format

An object of class "data.frame".

Details

Format: Two-way ANOVA. Description: The rows correspond to different frequencies and to normal speech. 1. 500 hertz 2. 1000 hertz 3. 2000 hertz 4. 3000 hertz 5. 4000 hertz 6. 6000 hertz 7. Normal speech The columns classify the data in seven occupational groups: 1. professional–managerial 2. farm 3. clerical sales 4. craftsmen 5. operatives 6. service 7. laborers

Source

Roberts, J. and Cohrssen, J. (1968), Hearing levels of adults, US National Center for Health Statistics Publications, Series 11, No. 31

Examples

data(hearing)

20 image

huber

Tuning parameter the rho loss functions

Description

This function computes the tuning constant that yields an MM-regression estimator with a desired asymptotic efficiency when computed with a rho function in the corresponding family. The output of this function can be passed to the functions lmrobdet.control, mscale and rho.

Usage

huber(e)

Arguments

е

the desired efficiency of the corresponding regression estimator for Gaussian errors

Value

A length-1 vector with the corresponding tuning constant.

Author(s)

Kjell Konis

Examples

```
# Tuning parameters for an 85%-efficient M-estimator at a Gaussian model huber(.95)
```

image

Image data

Description

These data are part of a synthetic aperture satellite radar image corresponding to a suburb of Munich, and contain the values corresponding to three frequency bands for each of 1573 pixels of a radar image.

```
data(image)
```

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Format

An object of class "data.frame".

Details

Format: 1573 cases and 3 variables.

Source

Source: Frery, A. (2005), Personal communication.

Examples

data(image)

initPP

Robust multivariate location and scatter estimators

Description

This function computes robust multivariate location and scatter estimators using both random and deterministic starting points.

Usage

```
initPP(X, muldirand = 20, muldifix = 10, dirmin = 1000)
```

Arguments

X a data matrix with observations in rows.

muldirand used to determine the number of random directions (candidates), which is max(p*muldirand,

dirmin), where p is the number of columns in X.

muldifix used to determine the number of random directions (candidates), which is min(n,

2*muldifix*p).

dirmin minimum number of random directions

Details

This function computes robust multivariate location and scatter using both Pen~a-Prieto and random candidates.

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Value

A list with the following components:

idx A zero/one vector with ones in the positions of the suspected outliers

disma Robust squared Mahalanobis distances

center Robust mean estimate

cova Robust covariance matrix estimate

t Outlyingness of data points

Author(s)

Ricardo Maronna, <rmaronna@retina.ar>, based on original code by D. Pen~a and J. Prieto

References

```
http://www.wiley.com/go/maronna/robust
```

Examples

```
data(bus)
X0 <- as.matrix(bus)
X1 <- X0[,-9]
tmp <- initPP(X1)
round(tmp$cov[1:10, 1:10], 3)
tmp$center</pre>
```

INVTR2

Robust R^2 coefficient of determination

Description

This function computes a robust version of the R^2 coefficient of determination. It is used internally by lmrobdetMM, and not meant to be used directly.

Usage

```
INVTR2(RR2, family, cc)
```

Arguments

RR2	the proportional difference in loss functions (a naive robust R^2 coefficient).
family	family string specifying the name of the family of loss function to be used (current valid options are "bisquare", "opt" and "mopt").
СС	tuning parameters to be computed according to efficiency and / or breakdown considerations. See lmrobdet.control, bisquare, mopt and opt.

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Details

This function computes a robust version of the R^2 coefficient. It is used internally by 1mrobdetMM, and not meant to be used directly.

Value

An unbiased version of the robust R^2 coefficient of determination.

Author(s)

Victor Yohai, <victoryohai@gmail.com>

References

http://www.wiley.com/go/maronna/robust

leuk.dat

Leukemia Data

Description

Records for 33 leukemia patients.

Usage

data(leuk.dat)

Format

An object of class "data.frame".

Details

Description: The following features are present: wbc: white blood cell count; ag: presence or absence of a certain morphological characteristic in the white cells; and y: binary response variable, equals 1 if the patient survives more than 52 weeks, 0 otherwise.

Format: Numeric, 33 rows and 3 columns.

Source

Cook, R.D. and Weisberg, S. (1982). Residuals and Influence in Regression, Chapman and Hall; Johnson, W. (1985), Influence measures for logistic regression: another point of view, Biometrika, 72, 59-65.

Examples

```
data(leuk.dat)
```

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lmrobdet.control

Tuning parameters for lmrobdetMM and lmrobdetDCML

Description

This function sets tuning parameters for the MM estimator implemented in lmrobdetMM and the Distance Constrained Maximum Likelihood regression estimators computed by lmrobdetDCML.

Usage

```
lmrobdet.control(
 bb = 0.5,
  efficiency = 0.95,
  family = "mopt",
  tuning.psi,
  tuning.chi,
  compute.rd = FALSE,
  corr.b = TRUE,
  split.type = "f",
  initial = "S",
 max.it = 100,
  refine.tol = 1e-07,
  rel.tol = 1e-07,
  refine.PY = 10,
  solve.tol = 1e-07,
  trace.lev = 0,
  psc_keep = 0.5,
  resid_keep_method = "threshold",
  resid_keep_thresh = 2,
  resid_keep_prop = 0.2,
  py_maxit = 20,
  py_{eps} = 1e-05,
 mscale_maxit = 50,
 mscale_tol = 1e-06,
 mscale_rho_fun = "bisquare"
)
```

Arguments

bb	tuning constant (between 0 and 1/2) for the M-scale used to compute the initial S-estimator. It determines the robusness (breakdown point) of the resulting MM-estimator, which is bb. Defaults to 0.5.
efficiency	desired asymptotic efficiency of the final regression M-estimator. Defaults to 0.95.
family	string specifying the name of the family of loss function to be used (current valid options are "bisquare", "opt" and "mopt"). Incomplete entries will be matched to the current valid options. Defaults to "mopt".

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tuning.psi	tuning parameters for the regression M-estimator computed with a rho function as specified with argument family. If missing, it is computed inside lmrobdet.control to match the value of efficiency according to the family of rho functions specified in family. Appropriate values for tuning.psi for a given desired efficiency for Gaussian errors can be constructed using the functions bisquare, mopt and opt.
tuning.chi	tuning constant for the function used to compute the M-scale used for the initial S-estimator. If missing, it is computed inside lmrobdet.control to match the value of bb according to the family of rho functions specified in family.
compute.rd	logical value indicating whether robust leverage distances need to be computed.
corr.b	logical value indicating whether a finite-sample correction should be applied to the M-scale parameter bb.
split.type	determines how categorical and continuous variables are split. See splitFrame.
initial	string specifying the initial value for the M-step of the MM-estimator. Valid options are 'S', for an S-estimator and 'MS' for an M-S estimator which is appropriate when there are categorical explanatory variables in the model.
max.it	maximum number of IRWLS iterations for the MM-estimator
refine.tol	relative covergence tolerance for the S-estimator
rel.tol	relative covergence tolerance for the IRWLS iterations for the MM-estimator
refine.PY	number of refinement steps for the Pen~a-Yohai candidates
solve.tol	(for the S algorithm): relative tolerance for matrix inversion. Hence, this corresponds to solve.default's tol.
trace.lev	positive values (increasingly) provide details on the progress of the MM-algorithm
psc_keep	For pyinit, proportion of observations to remove based on PSCs. The effective proportion of removed observations is adjusted according to the sample size to be prosac*(1-p/n). See pyinit.
resid_keep_meth	
	For pyinit, how to clean the data based on large residuals. If "threshold", all observations with scaled residuals larger than C.res will be removed, if "proportion", observations with the largest prop residuals will be removed. See pyinit.
resid_keep_thre	
	See parameter resid_keep_method above. See pyinit.
resid_keep_prop	See parameter resid_keep_method above. See pyinit.
py_maxit	Maximum number of iterations. See pyinit.
py_eps	Relative tolerance for convergence. See pyinit.
mscale_maxit	Maximum number of iterations for the M-scale algorithm. See pyinit and mscale.
mscale_tol	Convergence tolerance for the M-scale algorithm. See mscale and mscale.
mscale_rho_fun	String indicating the loss function used for the M-scale. See pyinit.

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Details

The argument family specifies the name of the family of loss function to be used. Current valid options are "bisquare", "opt" and "mopt"—"opt" refers to the optimal psi function defined in Section 5.8.1. of the book Robust Statistics: Theory and Methods (with R) by Maronna, Martin, Yohai and Salibian-Barrera, "mopt" is a modified version of the optimal psi function to make it strictly increasing close to 0, and to make the corresponding weight function non-increasing near 0.

Value

A list with the necessary tuning parameters.

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>

See Also

```
pyinit, mscale.
```

Examples

```
data(coleman, package='robustbase') m2 <- lmrobdetMM(Y \sim ., data=coleman, control=lmrobdet.control(refine.PY=50)) m2 summary(m2)
```

lmrobdetDCML

Robust Distance Constrained Maximum Likelihood estimators for linear regression

Description

This function computes robust Distance Constrained Maximum Likelihood estimators for linear models.

```
lmrobdetDCML(
  formula,
  data,
  subset,
  weights,
  na.action,
  model = TRUE,
  x = !control$compute.rd,
  y = FALSE,
  singular.ok = TRUE,
```

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```
contrasts = NULL,
  offset = NULL,
  control = lmrobdet.control()
)
```

Arguments

formula a symbolic description of the model to be fit.

data an optional data frame, list or environment containing the variables in the model.

If not found in data, model variables are taken from environment(formula),

which usually is the root environment of the current R session.

subset an optional vector specifying a subset of observations to be used. weights an optional vector of weights to be used in the fitting process.

na.action a function to 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.

model logical value indicating whether to return the model frame

x logical value indicating whether to return the model matrix

y logical value indicating whether to return the vector of responses

singular.ok logical value. If FALSE a singular fit produces an error.

 $contrasts \qquad \qquad an \ optional \ list. \ See \ the \ contrasts.arg \ of \ \underline{model.matrix.default}.$

offset this can be used to specify an a priori known component to be included in the

linear predictor during fitting. An offset term can be included in the formula

instead or as well, and if both are specified their sum is used.

control a list specifying control parameters as returned by the function lmrobdet.control.

Details

This function computes Distance Constrained Maximum Likelihood regression estimators computed using an MM-regression estimator based on Pen~a-Yohai candidates (instead of subsampling ones). This function makes use of the functions lmrob.fit, lmrob.M..fit, .vcov.avar1, lmrob.S and lmrob.lar, from robustbase, along with utility functions used by these functions, modified so as to include use of the analytic form of the optimal psi and rho functions (for the optimal psi function, see Section 5.8.1 of Maronna, Martin, Yohai and Salibian Barrera, 2019)

Value

A list with the following components:

coefficients The estimated vector of regression coefficients

scale The estimated scale of the residuals

residuals The vector of residuals associated with the robust fit

converged Logical value indicating whether IRWLS iterations for the MM-estimator have

converged

iter Number of IRWLS iterations for the MM-estimator

rweightsMM Robustness weights for the MM-estimator

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fitted.values Fitted values associated with the robust fit rank Numeric rank of the fitted linear model

cov The estimated covariance matrix of the regression estimates

df.residual The residual degrees of freedom

contrasts (only where relevant) the contrasts used

xlevels (only where relevant) a record of the levels of the factors used in fitting

call the matched call

model if requested, the model frame used
x if requested, the model matrix used
y if requested, the response vector used

na.action (where relevant) information returned by model.frame on the special handling

of NAs

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>, based on 1mrob

References

```
http://www.wiley.com/go/maronna/robust
```

See Also

```
DCML, MMPY, SMPY
```

Examples

```
data(coleman, package='robustbase')
m1 <- lmrobdetDCML(Y ~ ., data=coleman)
m1
summary(m1)</pre>
```

lmrobdetLinTest

Robust likelihood ratio test for linear hypotheses

Description

This function computes a robust likelihood ratio test for linear hypotheses.

```
lmrobdetLinTest(object1, object2)
```

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Arguments

object1 an 1mrobdetMM or 1mrobM object with the fit corresponding to the complete

model

object2 an 1mrobdetMM or 1mrobM object with the fit corresponding to the model re-

stricted under the null linear hypothesis.

Value

A list with the following components: c("test", "chisq.pvalue", "f.pvalue", "df")

test The value of the F-statistic

f. pvalue p-value based on the F distribution

chisq.pvalue p-value based on the chi-squared distribution

df degrees of freedom

Author(s)

Victor Yohai, <vyohai@gmail.com>

References

http://www.wiley.com/go/maronna/robust

Examples

```
data(oats)
cont <- lmrobdet.control(bb = 0.5, efficiency = 0.85, family = "bisquare")
oats1M <- lmrobM(response1 ~ variety+block, control=cont, data=oats)
oats1M_var <- lmrobM(response1 ~ block, control=cont, data=oats)
( anov1M_var <- rob.linear.test(oats1M, oats1M_var) )</pre>
```

1mrobdetMM

Robust linear regression estimators

Description

This function computes an MM-regression estimators for linear models using deterministic starting points.

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Usage

```
lmrobdetMM(
  formula,
  data,
  subset,
  weights,
  na.action,
  model = TRUE,
  x = !control$compute.rd,
  y = FALSE,
  singular.ok = TRUE,
  contrasts = NULL,
  offset = NULL,
  control = lmrobdet.control()
)
```

Arguments

formula	a symbolic description of the model to be fit.
data	an optional data frame, list or environment containing the variables in the model. If not found in data, model variables are taken from environment(formula), which usually is the root environment of the current R session.
subset	an optional vector specifying a subset of observations to be used.
weights	an optional vector of weights to be used in the fitting process.
na.action	a function to 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.
model	logical value indicating whether to return the model frame
x	logical value indicating whether to return the model matrix
У	logical value indicating whether to return the vector of responses
singular.ok	logical value. If FALSE a singular fit produces an error.
contrasts	an optional list. See the contrasts.arg of model.matrix.default.
offset	this can be used to specify an a priori known component to be included in the linear predictor during fitting. An offset term can be included in the formula instead or as well, and if both are specified their sum is used.
control	a list specifying control parameters as returned by the function lmrobdet.control.

Details

This function computes MM-regression estimators computed using Pen~a-Yohai candidates (instead of subsampling ones). This function makes use of the functions lmrob.fit, lmrob.M..fit, .vcov.avar1, lmrob.S and lmrob.lar, from robustbase, along with utility functions used by these functions, modified so as to include use of the analytic form of the optimal psi and rho functions (for the optimal psi function, see Section 5.8.1 of Maronna, Martin, Yohai and Salibian Barrera, 2019)

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Value

A list with the following components:

coefficients The estimated vector of regression coefficients

scale The robust residual M-scale estimate using the final residuals from the con-

verged iterated weighted least square (IRWLS) algorithm final estimate

residuals The vector of residuals associated with the robust fit

loss Value of the objective function at the final MM-estimator

converged Logical value indicating whether IRWLS iterations for the MM-estimator have

converged

iter Number of IRWLS iterations for the MM-estimator

rweights Robustness weights for the MM-estimator fitted.values Fitted values associated with the robust fit Numeric rank of the fitted linear model

cov The estimated covariance matrix of the regression estimates

df.residual The residual degrees of freedom degree.freedom The residual degrees of freedom

scale. S Minimum robust scale associated with the preliminary highly robust but ineffi-

cient S-estimator.

r. squared The robust multiple correlation coefficient

adj.r.squared The adjusted robust multiple correlation coefficient taking into account the de-

grees of freedom of each term

contrasts (only where relevant) the contrasts used

xlevels (only where relevant) a record of the levels of the factors used in fitting

call the matched call

model if requested, the model frame used

x if requested, the model matrix used

y if requested, the response vector used

terms The terms object used.

assign Used to separate continuous from categorical columns in the design matrix

na.action (where relevant) information returned by model.frame on the special handling

of NAs

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>, based on 1mrob from package robustbase

References

http://www.wiley.com/go/maronna/robust

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See Also

```
DCML, MMPY, SMPY
```

Examples

```
data(coleman, package='robustbase')
m2 <- lmrobdetMM(Y ~ ., data=coleman)
m2
summary(m2)</pre>
```

1mrobdetMM.RFPE

Robust Final Prediction Error

Description

This function computes the robust Final Prediction Errors (RFPE) for a robust regression fit using M-estimates.

Usage

```
lmrobdetMM.RFPE(object, scale = NULL, bothVals = FALSE)
```

Arguments

object the MM element (of class lmrob) in an object of class lmrobdetMM.

scale a numeric value specifying the scale estimate used to compute the RFPE. Usu-

ally this should be the scale estimate from an encompassing model. If NULL, the

scale estimate in object is used.

bothVals a logical value: if TRUE the function returns the two terms of the RFPE expres-

sion separately (equation (5.39) in the reference book); otherwise, the value of

RFPE is returned.

Value

If the argument bothVals is FALSE, the robust final prediction error (numeric). Otherwise, the two terms of the RFPE expression in equation (5.39), Section 5.6.2 of Maronna et al. (2019), http://www.wiley.com/go/maronna/robust, are returned separately in a list with components named minRhoMM.C and penaltyRFPE

Author(s)

Victor Yohai, <victoryohai@gmail.com>, Matias Salibian-Barrera, <matias@stat.ubc.ca>

References

http://www.wiley.com/go/maronna/robust

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See Also

lmrobdetMM

Examples

```
data(coleman, package='robustbase')
m2 <- lmrobdetMM(Y ~ ., data=coleman)
lmrobdetMM.RFPE(m2)</pre>
```

1mrobM

Robust estimators for linear regression with fixed designs

Description

This function computes a robust regression estimator for a linear models with fixed designs.

Usage

```
lmrobM(
  formula,
  data,
  subset,
  weights,
  na.action,
  model = TRUE,
  x = FALSE,
  y = FALSE,
  singular.ok = TRUE,
  contrasts = NULL,
  offset = NULL,
  control = lmrobM.control()
)
```

Arguments

formula	a symbolic description of the model to be fit.
data	an optional data frame, list or environment containing the variables in the model. If not found in data, model variables are taken from $environment(formula)$, which usually is the root environment of the current R session.
subset	an optional vector specifying a subset of observations to be used.
weights	an optional vector of weights to be used in the fitting process.
na.action	a function to 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.
model	logical value indicating whether to return the model frame

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x logical value indicating whether to return the model matrix
y logical value indicating whether to return the vector of responses

singular.ok logical value. If FALSE a singular fit produces an error.

contrasts an optional list. See the contrasts.arg of model.matrix.default.

offset this can be used to specify an a priori known component to be included in the

linear predictor during fitting. An offset term can be included in the formula

instead or as well, and if both are specified their sum is used.

control a list specifying control parameters as returned by the function lmrobM.control.

Details

This function computes robust regression estimators for linear models with fixed designs. It computes an L1 estimator, and uses it as a starting point to find a minimum of a re-descending M estimator. The scale is set to a quantile of the absolute residuals from the L1 estimator. This function makes use of the functions lmrob.fit, lmrob..M..fit, .vcov.avar1, lmrob.S and lmrob.lar, from robustbase, along with utility functions used by these functions, modified so as to include use of the analytic form of the optimal psi and rho functions (for the optimal psi function, see Section 5.8.1 of Maronna, Martin, Yohai and Salibian Barrera, 2019)

Value

A list with the following components:

coefficients The estimated vector of regression coefficients

scale The estimated scale of the residuals

residuals The vector of residuals associated with the robust fit

converged Logical value indicating whether IRWLS iterations for the MM-estimator have

converged

iter Number of IRWLS iterations for the MM-estimator

rweights Robustness weights for the MM-estimator fitted.values Fitted values associated with the robust fit Numeric rank of the fitted linear model

cov The estimated covariance matrix of the regression estimates

df.residual The residual degrees of freedom

contrasts (only where relevant) the contrasts used

xlevels (only where relevant) a record of the levels of the factors used in fitting

call the matched call

model if requested, the model frame used
x if requested, the model matrix used
y if requested, the response vector used

na.action (where relevant) information returned by model.frame on the special handling

of NAs

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Author(s)

Victor Yohai, <vyohai@gmail.com>, based on lmrob

References

```
http://www.wiley.com/go/maronna/robust
```

Examples

```
data(shock)
cont <- lmrobM.control(bb = 0.5, efficiency = 0.85, family = "bisquare")
shockrob <- lmrobM(time ~ n.shocks, data = shock, control=cont)
shockrob
summary(shockrob)</pre>
```

lmrobM.control

Tuning parameters for lmrobM

Description

This function sets tuning parameters for the M estimators of regression implemented in 1mrobM.

Usage

```
lmrobM.control(
  bb = 0.5,
  efficiency = 0.99,
  family = "opt",
  tuning.chi,
  tuning.psi,
  max.it = 100,
  rel.tol = 1e-07,
  mscale_tol = 1e-06,
  mscale_maxit = 50,
  trace.lev = 0
)
```

Arguments

bb tuning constant (between 0 and 1/2) for the M-scale used to compute the residual

scale estimator. Defaults to 0.5.

efficiency desired asymptotic efficiency of the final regression M-estimator. Defaults to

0.85.

family string specifying the name of the family of loss function to be used (current valid

options are "bisquare", "opt" and "mopt"). Incomplete entries will be matched

to the current valid options.

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tuning.chi	tuning constant for the function used to compute the M-scale used for the residual scale estimator. If missing, it is computed inside lmrobdet.control to match the value of bb according to the family of rho functions specified in family.
tuning.psi	tuning parameters for the regression M-estimator computed with a rho function as specified with argument family. If missing, it is computed inside lmrobdet.control to match the value of efficiency according to the family of rho functions specified in family. Appropriate values for tuning.psi for a given desired efficiency for Gaussian errors can be constructed using the functions bisquare, mopt and opt.
max.it	maximum number of IRWLS iterations for the M-estimator
rel.tol	relative covergence tolerance for the IRWLS iterations for the M-estimator
mscale_tol	Convergence tolerance for the M-scale algorithm. See mscale.
mscale_maxit	Maximum number of iterations for the M-scale algorithm. See mscale.
trace.lev	positive values (increasingly) provide details on the progress of the M-algorithm

Value

A list with the necessary tuning parameters.

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>

Examples

```
data(coleman, package='robustbase')
m2 <- lmrobM(Y ~ ., data=coleman, control=lmrobM.control())
m2
summary(m2)</pre>
```

locScaleM

Robust univariate location and scale M-estimators

Description

This function computes M-estimators for location and scale.

```
locScaleM(x, psi = "mopt", eff = 0.95, maxit = 50, tol = 1e-04, na.rm = FALSE)
```

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Arguments

X	a vector of univariate observations
psi	a string indicating which score function to use. Valid options are "bisquare", "huber", "opt" and "mopt".
eff	desired asymptotic efficiency. Valid options are 0.85 , 0.9 and 0.95 (default) when $psi =$ "bisquare" or "huber", and 0.85 , 0.9 , 0.95 (default) and 0.99 when $psi =$ "opt" or "mopt".
maxit	maximum number of iterations allowed.
tol	tolerance to decide convergence of the iterative algorithm.
na.rm	a logical value indicating whether NA values should be stripped before the computation proceeds. Defaults to FALSE

Details

This function computes M-estimators for location and scale.

Value

A list with the following components:

mu The location estimate
std.mu Estimated standard deviation of the location estimator mu
disper M-scale/dispersion estimate

Author(s)

References

http://www.wiley.com/go/maronna/robust

```
set.seed(123)
r <- rnorm(150, sd=1.5)
locScaleM(r)
# 10% of outliers, sd of good points is 1.5
set.seed(123)
r2 <- c(rnorm(135, sd=1.5), rnorm(15, mean=-10, sd=.5))
locScaleM(r2)</pre>
```

38 logregBY

logregBY	Bianco and Yohai estimator for logistic regression
0 0	, ,

Description

This function computes the M-estimator proposed by Bianco and Yohai for logistic regression. By default, an intercept term is included and p parameters are estimated. Modified by Yohai (2018) to take as initial estimator a weighted ML estimator with weights derived from the MCD estimator. For more details we refer to Croux, C., and Haesbroeck, G. (2002), "Implementing the Bianco and Yohai estimator for Logistic Regression"

Usage

```
logregBY(x0, y, intercept = 1, const = 0.5, kmax = 1000, maxhalf = 10)
```

Arguments

x0 matrix of explanatory variables;

y vector of binomial responses (0 or 1);

intercept 1 or 0 indicating if an intercept is included or or not

const tuning constant used in the computation of the estimator (default=0.5); kmax maximum number of iterations before convergence (default=1000);

maxhalf max number of step-halving (default=10).

Value

A list with the following components:

coefficients estimates for the regression coefficients

standard.deviation

standard deviations of the coefficients

fitted.values fitted values

residual.deviances

residual deviances

components logical value indicating whether convergence was achieved

objective value of the objective function at the minimum

Author(s)

Christophe Croux, Gentiane Haesbroeck, Victor Yohai

References

http://www.wiley.com/go/maronna/robust

logregWBY 39

Examples

```
data(skin)
Xskin <- as.matrix( skin[, 1:2] )
yskin <- skin$vasoconst
skinBY <- logregBY(Xskin, yskin, intercept=1)
skinBY$coeff
skinBY$standard.deviation</pre>
```

logregWBY

Bianco and Yohai estimator for logistic regression

Description

This function computes the weighted M-estimator of Bianco and Yohai in logistic regression. By default, an intercept term is included and p parameters are estimated. Modified by Yohai (2018) to take as initial estimator a weighted ML estimator computed with weights derived from the MCD estimator of the continuous explanatory variables. The same weights are used to compute the final weighted M-estimator. For more details we refer to Croux, C., and Haesbroeck, G. (2002), "Implementing the Bianco and Yohai estimator for Logistic Regression"

Usage

```
logregWBY(x0, y, intercept = 1, const = 0.5, kmax = 1000, maxhalf = 10)
```

Arguments

x0 matrix of explanatory variables;

y vector of binomial responses (0 or 1);

intercept 1 or 0 indicating if an intercept is included or or not

const tuning constant used in the computation of the estimator (default=0.5); kmax maximum number of iterations before convergence (default=1000);

maxhalf max number of step-halving (default=10).

Value

A list with the following components:

coefficients estimates for the regression coefficients

standard.deviation

standard deviations of the coefficients

fitted.values fitted values

residual.deviances

residual deviances

components logical value indicating whether convergence was achieved

objective value of the objective function at the minimum

40 logregWML

Author(s)

Christophe Croux, Gentiane Haesbroeck, Victor Yohai

References

```
http://www.wiley.com/go/maronna/robust
```

Examples

```
data(skin)
Xskin <- as.matrix( skin[, 1:2] )
yskin <- skin$vasoconst
skinWBY <- logregWBY(Xskin, yskin, intercept=1)
skinWBY$coeff
skinWBY$standard.deviation</pre>
```

logregWML

Weighted likelihood estimator for the logistic model

Description

This function computes a weighted likelihood estimator for the logistic model, where the weights penalize high leverage observations. In this version the weights are zero or one.

Usage

```
logregWML(x0, y, intercept = 1)
```

Arguments

x0 p x n matrix of explanatory variables, p is the number of explanatory variables,

n is the number of observations

y response vector

intercept 1 or 0 indicating if an intercept is included or or not

Value

A list with the following components:

```
coefficients vector of regression coefficients
standard.deviation
standard deviations of the regression coefficient estimators
fitted.values vector with the probabilities of success
residual.deviances
residual deviances
cov covariance matrix of the regression estimates
```

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objective value of the objective function at the minimum

xweights vector of zeros and ones used to compute the weighted maimum likelihood esti-

mator

Author(s)

Victor Yohai

References

http://www.wiley.com/go/maronna/robust

Examples

```
data(skin)
Xskin <- as.matrix( skin[, 1:2] )
yskin <- skin$vasoconst
skinWML <- logregWML(Xskin, yskin, intercept=1)
skinWML$coeff
skinWML$standard.deviation</pre>
```

mineral

Mineral data

Description

Contents (in parts per million) of 22 chemical elements in 53 samples of rocks in Western Australia. Two columns (8 and 9) were selected for use in this book.

Usage

```
data(mineral)
```

Format

An object of class "data.frame".

Details

Format: Numeric with 53 rows and 2 columns:

Source

Smith, R.E., Campbell, N.A. and Lichfield, A. (1984), Multivariate statistical techniques applied to pisolitic laterite geochemistry at Golden Grove, Western Australia, Journal of Geochemical Exploration, 22, 193-216.

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Examples

```
data(mineral)
```

MMPY

MM regression estimator using Pen~a-Yohai candidates

Description

This function computes MM-regression estimator using Pen~a-Yohai candidates for the initial Sestimator. This function is used internally by lmrobdetMM, and not meant to be used directly.

Usage

```
MMPY(X, y, control, mf)
```

Arguments

X design matrix
y response vector

control a list of control parameters as returned by lmrobdet.control

mf model frame

Value

an 1mrob object with the M-estimator obtained starting from the S-estimator computed with the Pen~a-Yohai initial candidates. The properties of the final estimator (efficiency, etc.) are determined by the tuning constants in the argument control.

Author(s)

Victor Yohai, <victoryohai@gmail.com>, Matias Salibian-Barrera, <matias@stat.ubc.ca>

References

```
http://www.wiley.com/go/maronna/robust
```

See Also

```
DCML, MMPY, SMPY
```

mopt 43

mopt	Tuning parameter for a rho function in the modified (asymptotic bias-) optimal family

Description

This function computes the tuning constant that yields an MM-regression estimator with a desired asymptotic efficiency when computed with a rho function in the corresponding family. The output of this function can be passed to the functions lmrobdet.control, mscale and rho.

Usage

mopt(e)

Arguments

е

the desired efficiency of the corresponding regression estimator for Gaussian errors

Value

A vector with named elements containing the corresponding tuning parameters.

Author(s)

Kjell Konis

Examples

```
\# Tuning parameters for an 85%-efficient M-estimator at a Gaussian model mopt(.85)
```

neuralgia

Neuralgia data

Description

Neuralgia data. More details here.

Usage

data(neuralgia)

Format

An object of class "data.frame".

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Source

Source goes here.

References

References go here.

Examples

data(neuralgia)

oats

Oats data

Description

Yield of grain for eight varieties of oats in five replications of a randomized-block experiment

Usage

data(oats)

Format

An object of class "data.frame".

Details

Format: Two-way ANOVA table with 8 rows and 5 columns.

Source

Scheffe, H. (1959), Analysis of Variance. New York: John Wiley.

References

References go here.

Examples

data(oats)

opt 45

opt	Tuning parameter for a rho function in the (asymptotic bias-) optimal family

Description

This function computes the tuning constant that yields an MM-regression estimator with a desired asymptotic efficiency when computed with a rho function in the corresponding family. The output of this function can be passed to the functions lmrobdet.control, mscale and rho.

Usage

opt(e)

Arguments

e the desired efficiency of the corresponding regression estimator for Gaussian errors

Value

A vector with named elements containing the corresponding tuning parameters.

Author(s)

Kjell Konis

Examples

```
# Tuning parameters for an 85%-efficient M-estimator at a Gaussian model opt(.85)
```

pcaRobS

Robust principal components

Description

This function computes robust principal components based on the minimization of the "residual" M-scale.

Usage

```
pcaRobS(X, ncomp, desprop = 0.9, deltasca = 0.5, maxit = 100)
```

46 pcaRobS

Arguments

X a data matrix with observations in rows.

ncomp desired (maximum) number of components

desprop desired (minimum) proportion of explained variability (default = 0.9)

deltasca "delta" parameter of the scale M-estimator (default=0.5)

maxit maximum number of iterations (default= 100)

Value

A list with the following components:

q The actual number of principal components

propex The actual proportion of unexplained variability

eigvec Eigenvectors, in a p x q matrix

fit an n x p matrix with the rank-q approximation to X

repre An n x q matrix with representation of data in R^q (scores)

propSPC A vector of length p with the cumulative explained variance from initial SPC

Author(s)

Ricardo Maronna, <rmaronna@retina.ar>, based on original code by D. Pen~a and J. Prieto

References

http://www.wiley.com/go/maronna/robust

```
data(bus)
X0 <- as.matrix(bus)
X1 <- X0[,-9]
ss <- apply(X1, 2, mad)
mu <- apply(X1, 2, median)
X <- scale(X1, center=mu, scale=ss)
q <- 3  #compute three components
rr <- pcaRobS(X, q, 0.99)
round(rr$eigvec, 3)</pre>
```

prcompRob 47

prcompRob	Robust Principal Components Cont'd

Description

This function uses the pcaRobS function to compute all principal components while behaving similarly to the prcomp function

Usage

```
prcompRob(x, rank. = NULL, delta.scale = 0.5, max.iter = 100L)
```

Arguments

X	data matrix with observations in rows
rank.	Maximal number of principal components to be used (optional)
delta.scale	"delta" parameter of the scale M-estimator (default = 0.5)

max.iter maximum number of iterations (default = 100)

Value

sdev the standard deviation of the principal components

rotation matrix containing the factor loadings x matrix containing the rotated data

center the centering used

Author(s)

```
Gregory Brownson, <gregory.brownson@gmail.com>
```

```
data(wine)
p.wine <- prcompRob(wine)
summary(p.wine)

## Choose only 5
p5.wine <- prcompRob(wine, rank. = 5)
summary(p5.wine)</pre>
```

48 refine.sm

refine.sm

IRWLS iterations for S- or M-estimators

Description

This function performs iterative improvements for S- or M-estimators.

Usage

```
refine.sm(
   x,
   y,
   initial.beta,
   initial.scale,
   k = 50,
   conv = 1,
   b,
   cc,
   family,
   step = "M"
)
```

Arguments

X	design matrix
у	vector of responses
initial.beta	vector of initial regression estimates
initial.scale	initial residual scale estimate. If missing the (scaled) median of the absolute residuals is used.
k	maximum number of refining steps to be performed
conv	an integer indicating whether to check for convergence (1) at each step, or to force running k steps (0)
b	tuning constant for the M-scale estimator, used if iterations are for an S-estimator.
СС	tuning constant for the rho function.
family	string specifying the name of the family of loss function to be used (current valid options are "bisquare", "opt" and "mopt")
step	a string indicating whether the iterations are to compute an S-estiamator ('S') or an M-estimator ('M')

Details

This function performs iterative improvements for S- or M-estimators. Both iterations are formally the same, the only difference is that for M-iterations the residual scale estimate remains fixed, while for S-iterations it is updated at each step. In this case, we follow the Fast-S algorithm of Salibian-Barrera and Yohai an use one step updates for the M-scale, as opposed to a full computation. This as internal function.

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Value

A list with the following components:

beta.rw The updated vector of regression coefficients scale.rw The corresponding estimated residual scale

converged A logical value indicating whether the algorithm converged

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>.

resex Resex data

Description

A monthly series of inward movement of residential telephone extensions in a fixed geographic area from January 1966 to May 1973.

Usage

data(resex)

Format

An object of class "data.frame".

Details

Format: numeric vector of size 89.

Source

Source Engineering, 2nd. Edition, New York, John Wiley.

References

Brubacher. S.R. (1974), Time series outlier detection and modeling with interpolation, Bell Laboratories Technical Memo.

Examples

data(resex)

50 rho

rho	Rho functions

Description

This function returns the value of the "rho" loss function used to compute either an M-scale estimator or a robust regression estimator. It currently can be used to compute the bisquare, optimal and modified optimal loss functions.

Usage

```
rho(u, family = " bisquare", cc, standardize = TRUE)
```

Arguments

u	point or vector at which rho is to be evaluated
family	family string specifying the name of the family of loss function to be used (current valid options are "bisquare", "opt" and "mopt").
СС	tuning parameters to be computed according to efficiency and / or breakdown considerations. See lmrobdet.control, bisquare, mopt and opt.
standardize	logical value determining whether the rho function is to be standardized so that its maximum value is 1. See Mpsi.

Value

The value(s) of rho at u

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>

```
# Evaluate rho tuned for 85% efficiency
rho(u=1.1, family='bisquare', cc=bisquare(.85))
# Evaluate rho tuned for 50% breakdown
rho(u=1.1, family='opt', cc=lmrobdet.control(bb=.5, family='opt')$tuning.chi)
```

rhoprime 51

rhoprime The first derivative of the rho function

Description

The first derivative of the rho function

Usage

```
rhoprime(u, family, cc, standardize = FALSE)
```

Arguments

u	point or vector at which rho is to be evaluated
family	family string specifying the name of the family of loss function to be used (current valid options are "bisquare", "opt" and "mopt").
СС	tuning parameters to be computed according to efficiency and / or breakdown considerations. See lmrobdet.control , bisquare, mopt and opt.
standardize	logical value determining whether the rho function is to be standardized so that its maximum value is 1. See Mpsi.

Value

The value of the first derivative rho evaluated at u

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>

```
# Evaluate the derivative of a rho function tuned for 85% efficiency
rhoprime(u=1.1, family='bisquare', cc=bisquare(.85))
# Evaluate the derivative of a rho function tuned for 50% breakdown
rhoprime(u=1.1, family='opt', cc=lmrobdet.control(bb=.5, family='opt')$tuning.chi)
```

52 rhoprime2

rhoprime2	The second derivative of the rho function	
-----------	---	--

Description

The second derivative of the rho function

Usage

```
rhoprime2(u, family, cc, standardize = FALSE)
```

Arguments

u	point or vector at which rho is to be evaluated
family	family string specifying the name of the family of loss function to be used (current valid options are "bisquare", "opt" and "mopt").
СС	tuning parameters to be computed according to efficiency and / or breakdown considerations. See lmrobdet.control, bisquare, mopt and opt.
standardize	logical value determining whether the rho function is to be standardized so that its maximum value is 1. See Mpsi.

Value

The value of the second derivative of rho evaluated at u

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>

```
# Evaluate the 2nd derivative of a rho function tuned for 85% efficiency
rhoprime2(u=1.1, family='bisquare', cc=bisquare(.85))
# Evaluate the 2nd derivative of a rho function tuned for 50% breakdown
rhoprime2(u=1.1, family='opt', cc=lmrobdet.control(bb=.5, family='opt')$tuning.chi)
```

scaleM 53

|--|

Description

This function computes an M-scale, which is a robust scale (spread) estimator. M-estimators of scale are a robust alternative to the sample standard deviation. Given a vector of residuals r, the M-scale estimator s solves the non-linear equation mean(rho(r/s, cc))=sb, where sb and sc are user-chosen tuning constants. In this package the function sh one of Tukey's bisquare family. The breakdown point of the estimator is min(s, 1-s), so the optimal choice for sh is 0.5. To obtain a consistent estimator the constant sh constant sh where sh is a standard normal random variable.

Usage

```
scaleM(
    u,
    delta = 0.5,
    tuning.chi = 1.547645,
    family = "bisquare",
    max.it = 100,
    tol = 1e-06,
    tolerancezero = .Machine$double.eps
)
```

Arguments

u	vector of residuals
delta	the right hand side of the M-scale equation
tuning.chi	the tuning object for the rho function as returned by lmrobdet.control, bisquare, mopt or opt. It should correspond to the family of rho functions specified in the argument family.
family	string specifying the name of the family of loss function to be used (current valid options are "bisquare", "opt" and "mopt").
max.it	maximum number of iterations allowed
tol	relative tolerance for convergence
tolerancezero	smallest (in absolute value) non-zero value accepted as a scale. Defaults to .Machine\$double.eps

Details

The iterative algorithm starts from the scaled median of the absolute values of the input vector, and then cycles through the equation $s^2 = s^2 * mean(rho(r/s, cc)) / b$.

Value

The scale estimate value at the last iteration or at convergence.

54 shock

Author(s)

Matias Salibian-Barrera, <matias@stat.ubc.ca>

Examples

```
set.seed(123)
r <- rnorm(150, sd=1.5)
mscale(r)
sd(r)
# 10% of outliers, sd of good points is 1.5
set.seed(123)
r2 <- c(rnorm(135, sd=1.5), rnorm(15, mean=-5, sd=.5))
mscale(r2)
sd(r2)</pre>
```

shock

Shock data

Description

Times recorded for a rat to go through a shuttlebox in successive attempts. If the time exceeded 5 seconds, the rat received an electric shock for the duration of the next attempt. The data are the number of shocks received and the average time for all attempts between shocks.

Usage

```
data(shock)
```

Format

An object of class "data.frame".

Details

Format: Numeric matrix with 16 rows and 2 columns

Source

Bond, N.W. (1979), Impairment of shuttlebox avoidance-learning following repeated alcohol withdrawal episodes in rats, Pharmacology, Biochemistry and Behavior, 11, 589-591.

References

References go here.

```
data(shock)
```

skin 55

skin Skin data

Description

These data correspond to a study of the relationship between air inspiration and blood circulation in the skin.

Usage

data(skin)

Format

An object of class "data.frame".

Details

Description: The covariates are the logarithms of the volume of air inspired (log VOL) and of the inspiration rate (log RATE). The response (column 3) is the presence or absence of vasoconstriction of the skin of the digits after air inspiration. Format Numeric, 23 rows and 3 columns.

Source

Finney, D.J. (1947), The estimation from individual records of the relationship between dose and quantal response, Biometrika, 34, 320-334.

Examples

data(skin)

SMPY

SM regression estimator using Pen~a-Yohai candidates

Description

This function computes a robust regression estimator when there are categorical / dummy explanatory variables. It uses Pen~a-Yohai candidates for the S-estimator. This function is used internally by lmrobdetMM, and not meant to be used directly.

Usage

```
SMPY(mf, y, control, split)
```

56 stackloss

Arguments

mf	model frame
у	response vector
control	a list of control parameters as returned by lmrobdet.control
split	a list as returned by splitFrame containing the continuous and dummy compo-

nents of the design matrix

Value

an 1mrob object with the M-estimator obtained starting from the MS-estimator computed with the Pen~a-Yohai initial candidates. The properties of the final estimator (efficiency, etc.) are determined by the tuning constants in the argument control.

Author(s)

Victor Yohai, <victoryohai@gmail.com>, Matias Salibian-Barrera, <matias@stat.ubc.ca>

References

```
http://www.wiley.com/go/maronna/robust
```

See Also

```
DCML, MMPY, SMPY
```

|--|

Description

Observations from 21 days operation of a plant for the oxidation of ammonia as a stage in the production of nitric acid.

Usage

```
data(stackloss)
```

Format

An object of class "data.frame".

Details

Format: 21 cases and 4 continuous variables. Description: The columns are: 1. air flow 2. cooling water inlet temperature (C) 3. acid concentration (4. Stack loss, defined as the percentage of ingoing ammonia that escapes unabsorbed (response)

step.lmrobdetMM 57

Source

Brownlee, K.A. (1965), Statistical Theory and Methodology in Science and Engineering, 2nd Edition, New York: John Wiley & Sons, Inc.

Examples

```
data(stackloss)
```

step.lmrobdetMM

Robust stepwise using RFPE

Description

This function performs stepwise model selection on a robustly fitted linear model using the RFPE criterion and the robust regression estimators computed with lmrobdetMM. Only backwards stepwise is currently implemented.

Usage

```
step.lmrobdetMM(
  object,
  scope,
  direction = c("both", "backward", "forward"),
  trace = TRUE,
  keep = NULL,
  steps = 1000,
  whole.path = FALSE
)
```

Arguments

object	a robust fit as returned by lmrobdetMM
scope	either a formula or a list with elements lower and upper each of which is a formula. The terms in the right-hand-side of lower are always included in the model and the additional terms in the right-hand-side of upper are the candidates for inclusion/exclusion from the model. If a single formula is given, it is taken to be upper, and lower is set to the empty model. The . operator is interpreted in the context of the formula in object.
direction	the direction of stepwise search. Currenly only backward stepwise searches are implemented.

trace logical. If TRUE information about each step is printed on the screen.

keep a filter function whose input is a fitted model object and the associated AIC

statistic, and whose output is arbitrary. Typically keep will select a subset of the components of the object and return them. The default is not to keep anything.

steps maximum number of steps to be performed. Defaults to 1000, which should

mean as many as needed.

58 step.ImrobdetMM

whole.path

if FALSE (default) variables are dropped until the RFPE fails to improve. If TRUE the best variable to be dropped is removed, even if this does not improve the RFPE.

Details

Presently only backward stepwise selection is supported. During each step the Robust Final Prediction Error (as computed by the function lmrobdetMM.RFPE) is calculated for the current model and for each sub-model achievable by deleting a single term. If the argument whole path is FALSE, the function steps to the sub-model with the lowest Robust Final Prediction Error or, if the current model has the lowest Robust Final Prediction Error, terminates. If the argument whole path is TRUE, the function steps through all smaller submodels removing, at each step, the variable that most reduces the Robust Final Prediction Error. The scale estimate from object is used to compute the Robust Final Prediction Error throughout the procedure.

Value

If whole.path == FALSE the function returns the robust fit as obtained by lmrobdetMM using the final model. If whole.path == TRUE a list is returned containing the RFPE of each model on the sequence of submodels. The names of the components of this list are the formulas that correspods to each model.

Author(s)

Victor Yohai, <victoryohai@gmail.com>, Matias Salibian-Barrera, <matias@stat.ubc.ca>

References

http://www.wiley.com/go/maronna/robust

See Also

```
DCML, MMPY, SMPY
```

```
cont <- lmrobdet.control(bb = 0.5, efficiency = 0.85, family = "bisquare")
set.seed(300)
X <- matrix(rnorm(50*6), 50, 6)
beta <- c(1,1,1,0,0,0)
y <- as.vector(X %*% beta) + 1 + rnorm(50)
y[1:6] <- seq(30, 55, 5)
for (i in 1:6) X[i,] <- c(X[i,1:3],i/2,i/2,i/2)
Z <- cbind(y,X)
Z <- as.data.frame(Z)
obj <- lmrobdetMM(y ~ ., data=Z, control=cont)
out <- step.lmrobdetMM(obj)</pre>
```

vehicle 59

vehicle

Vehicle data

Description

The original data set contains an ensemble of shape feature extractors to the 2D silhouettes of different vehicles. The purpose is to classify a given silhouette as one of four types of vehicle, using a set of 18 features extracted from the silhouette. Here we deal with the "van" type, which has 217 cases. Description; The following features were extracted from the silhouettes. 1. compactness 2. circularity 3. distance circularity 4. radius ratio 5. principal axis aspect ratio 6. maximum length aspect ratio 7. scatter ratio 8. elongatedness 9. principal axis rectangularity 10. maximum length rectangularity 11. scaled variance along major axis 12. scaled variance along minor axis 13. scaled radius of gyration 14. skewness about major axis 15. skewness about minor axis 16. kurtosis about minor axis 17. kurtosis about major axis 18. hollows ratio

Usage

data(vehicle)

Format

An object of class "data.frame".

Details

Format: Numeric, 217 rows and 18 columns.

Source

Turing Institute, Glasgow, and are available at https://archive.ics.uci.edu/ml/datasets/Statlog+(Vehicle+Silhouettes).

Examples

data(vehicle)

waste

Waste data

Description

Waste data. The original data are the result of a study on production waste and land use by Golueke and McGauhey (1970), and contain nine variables, of which we consider six.

Usage

data(waste)

60 wine

Format

An object of class "data.frame".

Details

Format: 40 cases and 6 continuous variables. Description: The columns are 1. industrial land (acres) 2. fabricated metals (acres) 3. trucking and wholesale trade (acres) 4. retail trade (acres) 5. restaurants and hotels (acres) 6. solid waste (millions of tons), response

Source

Golueke, C.G. and McGauhey, P.H. (1970), Comprehensive Studies of Solid Waste Management, US Department of Health, Education and Welfare, Public Health Services Publication No. 2039.

References

Golueke, C.G. and McGauhey, P.H. (1970), Comprehensive Studies of Solid Waste Management, US Department of Health, Education and Welfare, Public Health Services Publication No. 2039.

Examples

data(waste)

wine

Wine data

Description

It contains, for each of 59 wines grown in the same region in Italy, the quantities of 13 constituents. The original purpose of the analysis was to classify wines from different cultivars by means of these measurements. In this example we treat cultivar one.

Usage

data(wine)

Format

An object of class "data.frame".

Details

Format: Numeric, 59 rows and 13 columns. Description: The attributes are: 1. Alcohol 2. Malic acid 3. Ash 4. Alcalinity of ash 5. Magnesium 6. Total phenols 7. Flavanoids 8. Nonflavanoid phenols 9. Proanthocyanins 10. Color intensity 11. Hue 12. OD280/OD315 of diluted wines 13. Proline

wine 61

Source

Hettich, S. and Bay, S.D. (1999), The UCI KDD Archive http://kdd.ics.uci.edu. Irvine, CA: University of California, Department of Information and Computer Science.

Examples

data(wine)

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