Package 'robreg3S'

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robre	Robust regression estimation and inference in the presence of cellwise and casewise contamination
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	ompilation no
License	GPL (>= 2)
Imports	robustbase
Depends	GSE, MASS
Descript	ion Three-step regression and inference for cellwise and casewise contamination.
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Version	0.3
	ree-Step Regression and Inference for Cellwise and Casewise ntamination
Type Pa	ckage

Description

Finds 3S-robust regression estimator using the adaptive consistent filter.

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Usage

```
robreg3S(y, x, dummies=NULL, filter=TRUE, alpha=0.20, K=5, ...)
```

Arguments

y vector of responses.

x matrix of the numerical variables.

dummies matrix of the dummy covariates, i.e., where each column are 0–1 vectors.

filter logical, whether the filtering is used. Default value is TRUE.

alpha 1-alpha upper quantile (and alpha lower quantile) of the covariate distribution

used in tail comparison in the first step. An exponential tail is used as the refer-

ence distribution. Default value is 0.20.

K number of alternating M-S iterations in the estimation of the coefficients of the

dummy covariates. Default value is 5. See Leung et al. for more details.

... optional arguments to be used in the computation of GSE in the second step.

See GSE

Details

This function computes 3S-robust regression as described in Leung et al. (2015).

If the model contains dummy variables (i.e., dummies != NULL), 3S-regression is computed using an iterative algorithm as described in Leung et al. (2015). Briefly, the algorithm first estimates the coefficients of the dummies using an M-estimator of regression and the coefficients of the continuous covariates using the original 3S-regression. See Leung et al. (2015) for more details.

Value

A list with components:

Summary. Table Matrix of information available a	bout the estimator. It contains regression co-
--------------------------------------------------	------------------------------------------------

efficients, and for dummies != NULL, columns for the standard error, t-statistic,

and p-value.

coef vector of regression coefficients.

acov matrix of the asymptotic covariate matrix, only for dummies != NULL.

resid vector of residuals, that is the response minus the fitted values.

sigma.hat the estimated residual standard error.

MD the squared Mahalanobis distances of each observation based on the continuous

covariates to the generalized location S-estimator with respect to the generalized

scatter S-estimator.

xfilter filtered matrix of the numerical variables from Step 1 of the estimator.

ximpute matrix of the numerical variables with filtered cells imputed from Step 2 of the

estimator.

weight vector of the weights used in the estimation of the location generalized S-estimator.

Not meant to be accessed.

Syx estimated generalized S-scatter from Step 2. Not meant to be accessed.

myx estimated generalized S-location from Step 2. Not meant to be accessed.

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Author(s)

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References

Leung, A., Zamar, R.H., and Zhang, H. Robust regression estimation and inference in the presence of cellwise and casewise contamination. arXiv:1509.02564.

See Also

GSE, generate.cellcontam.regress, generate.casecontam.regress, generate.casecontam.regress.dummies, generate.casecontam.regress.dummies

Examples

```
## Boston housing data
data(Boston, package="MASS")
boston <- Boston; rm(Boston)</pre>
boston$crim <- log(boston$crim)</pre>
boston$nox <- boston$nox^2</pre>
boston$rm <- boston$rm^2</pre>
boston$dis <- log(boston$dis)</pre>
boston$lstat <- log(boston$lstat)</pre>
boston$medv <- log(boston$medv)</pre>
boston$black <- boston$black/1000</pre>
boston$age <- boston$age/100
boston$tax <- boston$tax/100</pre>
boston$indus <- boston$indus/100
boston <- subset( boston, select=c(medv, crim, nox, rm, age, dis, tax, ptratio, black, lstat) )</pre>
## LS, MM, 3S
set.seed(100)
fit.LS <- lm(medv ~ ., data=boston)</pre>
fit.MM <- robustbase::lmrob(medv ~ ., data=boston)</pre>
fit.2S <- robreg3S( y=boston$medv, x=as.matrix(subset(boston,select=-medv)), filter = FALSE )</pre>
fit.3S <- robreg3S( y=boston$medv, x=as.matrix(subset(boston, select=-medv)) )</pre>
## Compare estimated coefficients
nrow(boston) * sum(( coef(fit.LS)[-1] - coef(fit.3S)[-1])^2 * apply(boston[,-1], 2, mad)^2)
nrow(boston) * sum(( coef(fit.MM)[-1] - coef(fit.3S)[-1])^2* apply(boston[,-1], 2, mad)^2) \\
nrow(boston) * sum(( coef(fit.2S)[-1] - coef(fit.3S)[-1])^2 * apply(boston[,-1], 2, mad)^2)
## Summary table
summary(fit.3S)
```

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nation		study on cell- and case-wise contami-
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Description

Includes the data generator for the simulation study on cell- and case-wise contamination that appears on Leung et al. (2015).

Usage

```
generate.randbeta(p)
generate.cellcontam.regress(n, p, A, sigma, b, k, cp)
generate.casecontam.regress(n, p, A, sigma, b, l, k, cp)
generate.cellcontam.regress.dummies(n, p, pd, probd, A, sigma, b, k, cp)
generate.casecontam.regress.dummies(n, p, pd, probd, A, sigma, b, l, k, cp)
```

Arguments

n	integer indicating the number of observations to be generated.
р	integer indicating the number of continuous variables to be generated.
pd	integer indicating the number of dummy variables to be generated.
probd	vector of quantiles of length pd. To generate dummy variables pd continuous variables are first generated. Then, the variables are dichotomize at normal quantiles of probd.
A	a correlation matrix. See also generate.randcorr.
sigma	residual standard deviation.
b	vector of regression coefficients.
k	size of cellwise outliers and vertical outliers. See Leung et al. for details.
1	size of leverage outliers. See Leung et al. for details.
ср	proportion of cell- or case-wise contamination. Maximum of 10% for cellwise and 50% for casewise.

Value

A list with components:

x multivariate normal sample with cell- or case-wise contamination.

y vector of responses. dummies vector of dummies. simulation-tools 5

Author(s)

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References

Leung, A., Zamar, R.H., and Zhang, H. Robust regression estimation and inference in the presence of cellwise and casewise contamination. arXiv:1509.02564.

See Also

```
generate.randcorr
```

Examples

```
## Cellwise contaminated data simulation
## (continuous covariates only)
set.seed(10)
b <- 10*generate.randbeta(p=15)</pre>
A <- generate.randcorr(cond=100, p=15)
dat <- generate.cellcontam.regress(n=300, p=15, A=A, sigma=0.5, b=b, k=10, cp=0.05)
## LS
fit.LS <- lm(y \sim x, dat)
mean((coef(fit.LS)[-1] - b)^2)
## MM regression
fit.MM <- robustbase::lmrob( y ~ x, dat)</pre>
mean((coef(fit.MM)[-1] - b)^2)
## 3S regression
fit.3S <- robreg3S( y=dat$y, x=dat$x, init="imputed")</pre>
mean((coef(fit.3S)[-1] - b)^2)
## Casewise contaminated data simulation
## (continuous covariates only)
set.seed(10)
b <- 10*generate.randbeta(p=10)</pre>
A <- generate.randcorr(cond=100, p=10)
dat <- generate.casecontam.regress(n=200, p=10, A=A, sigma=0.5, b=b, l=8, k=10, cp=0.10)
## LS
fit.LS <- lm(y \sim x, dat)
mean((coef(fit.LS)[-1] - b)^2)
## MM regression
fit.MM <- robustbase::lmrob( y ~ x, dat)</pre>
mean((coef(fit.MM)[-1] - b)^2)
```

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```
## 3S regression
fit.3S <- robreg3S( y=dat$y, x=dat$x, init="imputed")</pre>
mean((coef(fit.3S)[-1] - b)^2)
## Not run:
## Cellwise contaminated data simulation
## (continuous and dummies covariates)
set.seed(10)
b <- 10*generate.randbeta(p=15)</pre>
A <- generate.randcorr(cond=100, p=15)
dat <- generate.cellcontam.regress.dummies(n=300, p=12, pd=3,</pre>
  probd=c(1/2,1/3,1/4), A=A, sigma=0.5, b=b, k=10, cp=0.05)
fit.LS <- lm( dat y ~ dat x + dat dummies)
mean((coef(fit.LS)[-1] - b)^2)
## MM regression
fit.MM <- robustbase::lmrob( dat$y ~ dat$x + dat$dummies)</pre>
mean((coef(fit.MM)[-1] - b)^2)
## 3S regression
fit.3S <- robreg3S( y=dat$y, x=dat$x, dummies=dat$dummies, init="imputed")</pre>
mean((coef(fit.3S)[-1] - b)^2)
## Casewise contaminated data simulation
## (continuous and dummies covariates)
set.seed(10)
b <- 10*generate.randbeta(p=15)</pre>
A <- generate.randcorr(cond=100, p=15)
dat <- generate.casecontam.regress.dummies(n=300, p=12, pd=3,</pre>
  probd=c(1/2,1/3,1/4), A=A, sigma=0.5, b=b, l=7, k=10, cp=0.10)
## 15
fit.LS <- lm( dat y \sim dat x + dat dummies)
mean((coef(fit.LS)[-1] - b)^2)
## MM regression
fit.MM <- robustbase::lmrob( dat$y ~ dat$x + dat$dummies)</pre>
mean((coef(fit.MM)[-1] - b)^2)
## 3S regression
fit.3S <- robreg3S( y=dat$y, x=dat$x, dummies=dat$dummies, init="imputed")</pre>
mean((coef(fit.3S)[-1] - b)^2)
## End(Not run)
```

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