Package 'subDebiased'

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Title Sharp Inference on Selected Subgroup in Observational Studies
Version 0.0.0.9000
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Description This package implements bootstrap-assisted desparsified Lasso and bootstrap-assisted R-split estimators on selected subgroup's treatment effect estimation. The implemented estimators remove the subgroup selection bias and the regularization bias induced by high-dimensional covariates.
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BSciCoverfun BSDesparseLasso BSSplitLasso cvDesparse cvSplit IFvarestbiascorr sigmaMatNew Zmatrix
Index

2 BSDesparseLasso

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RVCI	COVE	erfun

Compute CI for bootstrap-calibrated methods

Description

Compute CI for bootstrap-calibrated methods

Usage

```
BSciCoverfun(beta, TB = NULL, G = NULL, alpha = 0.95)
```

Arguments

beta estimated betas

TB recalibrated bootstrap statistics

G indices of subgroups alpha: confidence level

Value

Lower Bound Lower confidence bound Upper Bound Upper confidence bound

betaMax debiased maximum beta estimate

BSDesparseLasso

Bootstrap-calibrated Desparsified Lasso

Description

This method first constructs the debiased estimator of β via the desparsified Lasso procedure. Then it calculates the calibration term $\hat{b}_{max} = (1 - n^{r-0.5})(\hat{\beta}_{max} - \hat{\beta}_{j,lasso})$. Through B bootstrap iterations, it recalibrates the bootstrap statistic T_b . The bias-reduced estimate is computed as: $\hat{b}_{max} - \frac{1}{B} \sum_{b=1}^{B} T_b$.

Usage

```
BSDesparseLasso(y, x, r = NULL, G = NULL, B = NULL, alpha = 0.95, fold = 3)
```

Arguments

alpha

У	response
Х	design matrix
r	tuning parameter
G	subgroup indicator
В	bootstrap iterations

level of CI

3 BSSplitLasso

Value

lower confidence bound LowerBound upper confidence bound UpperBound

op optimal tuning

BSSplitLasso

Bootstrap-calibrated R-split method

Description

This method first obtains the estimate of β via repetitive splitting procedure (R-Split) through BB iterations. Then it calculates the calibration term $\tilde{b}_{max} = (1 - n^{r-0.5})(\tilde{\beta}_{max} - \tilde{\beta}_j)$. Through B iterations, it recalibrates the bootstrap statistic T_b . The bias-reduced estimate is computed as: $\tilde{b}_{max} - \frac{1}{B} \sum_{b=1}^{B} T_b$.

Usage

```
BSSplitLasso(
  у,
  r = NULL,
  G = NULL,
  B = NULL,
  BB = NULL,
  alpha = 0.95,
  splitRatio = 0.6,
  fold = 2
)
```

Arguments

У	response
x	design matrix
r	tuning parameter
G	subgroup indicator
В	bootstrap number
BB	split number
alpha	level ## change other places
splitRatio	split ratio
fold	cross validation fold

Value

LowerBound lower confidence bound UpperBound upper confidence bound

optimal tuning ор

cvSplit cvSplit

Select the optimal tuning for bootstrap-calibrated desparsified Lasso

Description

Select the optimal tuning for bootstrap-calibrated desparsified Lasso

Usage

```
cvDesparse(y, x, r = NULL, G = NULL, B = NULL, fold = 3)
```

Arguments

У	response
X	design matrix
r	candidate tuning parameters
G	indices of subgroups
В	bootstrap repetitions
fold	number of folds in cross-validation

Value

ор	optimal tuning parameter	r

cvSplit	Select the optimal tuning for bootstrap-calibrated R-Split through
	cross-validation

Description

Select the optimal tuning for bootstrap-calibrated R-Split through cross-validation

Usage

```
cvSplit(y, x, r = NULL, G = NULL, B = NULL, BB = NULL, ratio = NULL, fold = 2)
```

Arguments

у	response
X	design matrix
r	candidate tuning parameter
G	subgroup indicator
В	bootstrap iterations
BB	bootstrap iterations for repetitive splitting
ratio	ratio of data splitting
fold	number of folds in cross-validation

Value

op optimal tuning parameter

IFvarestbiascorr 5

IFvarestbiascorr

Cross-validation metric

Description

Cross-validation metric

Usage

```
IFvarestbiascorr(Ycount, alphaEst, n = NULL, splitSize = NULL)
```

Arguments

Ycount Y

alphaEst estimated values
n sample size
splitSize size of each split

Value

mean squared error

sigmaMatNew

Generate different types of covariance matrices

Description

Generate different types of covariance matrices

Usage

```
sigmaMatNew(p, type = NULL)
```

Arguments

p dimension of confounders

type type of matrix

Value

Sigma A covariance matrix

6 Zmatrix

Zmatrix

Generate the nodewise Lasso matrix used in desparsified Lasso

Description

Generate the nodewise Lasso matrix used in desparsified Lasso

Usage

```
Zmatrix(x, G = NULL)
```

Arguments

x nodewise confounder matrix

G indices of subgroups

Value

Z nodewise Lasso matrix

Index

```
BSciCoverfun, 2
BSDesparseLasso, 2
BSSplitLasso, 3
cvDesparse, 4
cvSplit, 4
IFvarestbiascorr, 5
sigmaMatNew, 5
Zmatrix, 6
```