# Package 'debiased.subgroup'

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Title Sharp Inference on Selected Subgroup in Observational Studies
Version 0.0.0.9000
Author Waverly(Linqing) Wei <li>linqing_wei@berkeley.edu&gt;</li>
<b>Description</b> 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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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  $\beta$  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

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

alpha level of CI

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#### Value

LowerBound lower confidence bound

UpperBound upper confidence bound

betaMax bias-reduced maximum beta estimate

betaEst debiased beta estimate for each subgroup

op optimal tuning

BSSplitLasso

Bootstrap-calibrated R-split method

## Description

This method first obtains the estimate of  $\beta$  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(
   y,
   x,
   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
ВВ	split number
alpha	level ## change other places
splitRatio	split ratio
fold	cross validation fold

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#### Value

LowerBound lower confidence bound UpperBound upper confidence bound

betaMax bias-reduced maximum beta estimate
betaEst debiased beta estimate for each subgroup

modelSize selected model size for R-split

op optimal tuning

cvDesparse 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 subgroupsB bootstrap repetitions

fold number of folds in cross-validation

## Value

op optimal tuning parameter

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)
```

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#### **Arguments**

y response

x design matrix

r candidate tuning parameter

G subgroup indicator
B 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 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

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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

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

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