

Package ‘COR’

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Title The COR for Optimal Subset Selection in Distributed Estimation

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Description An algorithm of optimal subset selection, related to Covariance matrices, Observation matrices and Response vectors (COR) to select the optimal subsets in distributed estimation. The philosophy of the package is described in Guo G. (2020) <[doi:10.1080/02331888.2020.1823979](https://doi.org/10.1080/02331888.2020.1823979)>.

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Imports stats

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Contents

beta_AD	2
beta_cor	3
beta_LW	3
communities	4
COR	9
ethylene_CO	10
LICbeta	11
LICnew	12
MSEbeta	12
MSEcom	13
MSEver	14

beta_AD	<i>Calculate the estimators of beta on the A-opt and D-opt</i>
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Description

Calculate the estimators of beta on the A-opt and D-opt

Usage

```
beta_AD(K = K, nk = nk, alpha = alpha, X = X, y = y)
```

Arguments

K	is the number of subsets
nk	is the length of subsets
alpha	is the significance level
X	is the observation matrix
y	is the response vector

Value

A list containing:

betaA	The estimator of beta on the A-opt.
betaD	The estimator of beta on the D-opt.

Examples

```
p=6;n=1000;K=2;nk=200;alpha=0.05;sigma=1
e=rnorm(n,0,sigma); beta=c(sort(c(runif(p,0,1))));
data=c(rnorm(n*p,5,10));X=matrix(data, ncol=p);
y=X%%beta+e;
beta_AD(K=K,nk=nk,alpha=alpha,X=X,y=y)
```

 beta_cor

Calculate the estimator of beta on the COR

Description

Calculate the estimator of beta on the COR

Usage

```
beta_cor(K = K, nk = nk, alpha = alpha, X = X, y = y)
```

Arguments

K	is the number of subsets
nk	is the length of subsets
alpha	is the significance level
X	is the observation matrix
y	is the response vector

Value

A list containing:

betaC	The estimator of beta on the COR.
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Examples

```
p=6;n=1000;K=2;nk=200;alpha=0.05;sigma=1
e=rnorm(n,0,sigma); beta=c(sort(c(runif(p,0,1))));
data=c(rnorm(n*p,5,10));X=matrix(data, ncol=p);
y=X%%beta+e;
beta_cor(K=K,nk=nk,alpha=alpha,X=X,y=y)
```

 beta_LW

Calculate the estimators of beta on the LEV-opt#'

Description

Calculate the estimators of beta on the LEV-opt#'

Usage

```
beta_LW(X, Y, K, nk)
```

Arguments

X	is the observation matrix
Y	is the response vector
K	is the number of subsets
nk	is the length of subsets

Value

A list containing:

betalev	The estimator of beta on the LEV-opt subset.
betam	The mean of the beta estimators across all K subsets.
AMSE	The Average Mean Squared Error (AMSE) for the estimator.
WMSE	The Weighted Mean Squared Error (WMSE) for the estimator.
MSElevb	The Mean Squared Error (MSE) of the LEV-opt estimator compared to the true beta.
MSEb	The Mean Squared Error (MSE) of the mean estimator (betam) compared to the true beta.
MSEyleva	The Mean Squared Error (MSE) of the LEV-opt estimator on the subset with the maximum hat value (Xleva).
MSEyleviy	The Mean Squared Error (MSE) of the LEV-opt estimator on the subset with the minimum hat value (Xlevi).
MSEW	The Mean Squared Error (MSE) of the weighted estimator (Wbeta) compared to the true beta.
MSEw	The Mean Squared Error (MSE) of the weighted estimator (wbeta) compared to the true beta.

communities

The communities and crime data set

Description

A data set about the communities and crime

Usage

```
data("communities")
```

Format

A data frame with 1994 observations on the following 128 variables.

- V1 a numeric vector
- V2 a numeric vector
- V3 a numeric vector
- V4 a character vector
- V5 a numeric vector
- V6 a numeric vector
- V7 a numeric vector
- V8 a numeric vector
- V9 a numeric vector
- V10 a numeric vector
- V11 a numeric vector
- V12 a numeric vector
- V13 a numeric vector
- V14 a numeric vector
- V15 a numeric vector
- V16 a numeric vector
- V17 a numeric vector
- V18 a numeric vector
- V19 a numeric vector
- V20 a numeric vector
- V21 a numeric vector
- V22 a numeric vector
- V23 a numeric vector
- V24 a numeric vector
- V25 a numeric vector
- V26 a numeric vector
- V27 a numeric vector
- V28 a numeric vector
- V29 a numeric vector
- V30 a numeric vector
- V31 a numeric vector
- V32 a numeric vector
- V33 a numeric vector
- V34 a numeric vector
- V35 a numeric vector

V36 a numeric vector
V37 a numeric vector
V38 a numeric vector
V39 a numeric vector
V40 a numeric vector
V41 a numeric vector
V42 a numeric vector
V43 a numeric vector
V44 a numeric vector
V45 a numeric vector
V46 a numeric vector
V47 a numeric vector
V48 a numeric vector
V49 a numeric vector
V50 a numeric vector
V51 a numeric vector
V52 a numeric vector
V53 a numeric vector
V54 a numeric vector
V55 a numeric vector
V56 a numeric vector
V57 a numeric vector
V58 a numeric vector
V59 a numeric vector
V60 a numeric vector
V61 a numeric vector
V62 a numeric vector
V63 a numeric vector
V64 a numeric vector
V65 a numeric vector
V66 a numeric vector
V67 a numeric vector
V68 a numeric vector
V69 a numeric vector
V70 a numeric vector
V71 a numeric vector
V72 a numeric vector

- V73 a numeric vector
- V74 a numeric vector
- V75 a numeric vector
- V76 a numeric vector
- V77 a numeric vector
- V78 a numeric vector
- V79 a numeric vector
- V80 a numeric vector
- V81 a numeric vector
- V82 a numeric vector
- V83 a numeric vector
- V84 a numeric vector
- V85 a numeric vector
- V86 a numeric vector
- V87 a numeric vector
- V88 a numeric vector
- V89 a numeric vector
- V90 a numeric vector
- V91 a numeric vector
- V92 a numeric vector
- V93 a numeric vector
- V94 a numeric vector
- V95 a numeric vector
- V96 a numeric vector
- V97 a numeric vector
- V98 a numeric vector
- V99 a numeric vector
- V100 a numeric vector
- V101 a numeric vector
- V102 a numeric vector
- V103 a numeric vector
- V104 a numeric vector
- V105 a numeric vector
- V106 a numeric vector
- V107 a numeric vector
- V108 a numeric vector
- V109 a numeric vector

V110 a numeric vector
V111 a numeric vector
V112 a numeric vector
V113 a numeric vector
V114 a numeric vector
V115 a numeric vector
V116 a numeric vector
V117 a numeric vector
V118 a numeric vector
V119 a numeric vector
V120 a numeric vector
V121 a numeric vector
V122 a numeric vector
V123 a numeric vector
V124 a numeric vector
V125 a numeric vector
V126 a numeric vector
V127 a numeric vector
V128 a numeric vector

Source

UCI repository

References

Redmond, M. A. and A. Baveja: A Data-Driven Software Tool for Enabling Cooperative Information Sharing Among Police Departments. *European Journal of Operational Research* 141 (2002) 660-678.

Examples

```
data(communities)
## maybe str(communities) ; plot(communities) ...
```


COR

*Calculate the optimal subset lengths on the COR***Description**

Calculate the optimal subset lengths on the COR

Usage

```
COR(K = K, nk = nk, alpha = alpha, X = X, y = y)
```

Arguments

K	is the number of subsets
nk	is the length of subsets
alpha	is the significance level
X	is the observation matrix
y	is the response vector

Value

A list containing:

seqL	The index of the subset with the minimum L value.
seqN	The index of the subset with the minimum N value.
lWMN	The optimal subset lengths on the COR.

Examples

```
p=6;n=1000;K=2;nk=200;alpha=0.05;sigma=1
e=rnorm(n,0,sigma); beta=c(sort(c(runif(p,0,1)))));
data=c(rnorm(n*p,5,10));X=matrix(data, ncol=p);
y=X**beta+e;
COR(K=K,nk=nk,alpha=alpha,X=X,y=y)
```

ethylene_CO

The chemical sensor data set

Description

A data set about chemical sensor

Usage

```
data("ethylene_CO")
```

Format

A data frame with 4001 observations on the following 19 variables.

V1 a character vector

V2 a character vector

V3 a character vector

V4 a character vector

V5 a character vector

V6 a character vector

V7 a character vector

V8 a character vector

V9 a character vector

V10 a character vector

V11 a character vector

V12 a character vector

V13 a character vector

V14 a character vector

V15 a character vector

V16 a character vector

V17 a character vector

V18 a character vector

V19 a character vector

Details

We selected the first 4001 rows on the original data set about 1048576 observations on 19 variables.

Source

UCI Repository

References

Wang, H. Y., Zhu, R., and Ma, P. (2018). Optimal subsampling for large sample logistic regression. *Journal of the American Statistical Association*, 113(522), 829-844.

Examples

```
data(ethylene_CO)
## maybe str(ethylene_CO) ; plot(ethylene_CO) ...
```

LICbeta

Calculate the LIC estimator for linear regression

Description

This function estimates the coefficients of a linear regression model using a design matrix ‘X’ and a response vector ‘Y’. It implements an A-optimal and D-optimal design criteria to choose optimal subsets of observations.

Usage

```
LICbeta(X, Y, alpha, K, nk)
```

Arguments

X	The observation matrix (n x p)
Y	The response vector (n x 1)
alpha	The significance level for computing confidence intervals
K	The number of subsets
nk	The number of observations per subset

Value

A list containing:

E5	The LIC estimator for linear regression.
----	------------------------------------------

LICnew	<i>Calculate the LIC estimator based on A-optimal and D-optimal criterion</i>
--------	-------------------------------------------------------------------------------

Description

Calculate the LIC estimator based on A-optimal and D-optimal criterion

Usage

LICnew(X, Y, alpha, K, nk)

Arguments

X	A matrix of observations (design matrix) with size n x p
Y	A vector of responses with length n
alpha	The significance level for confidence intervals
K	The number of subsets to consider
nk	The size of each subset

Value

A list containing:

E5	The LIC estimator based on A-optimal and D-optimal criterion.
----	---------------------------------------------------------------

Examples

```
p = 6; n = 1000; K = 2; nk = 200; alpha = 0.05; sigma = 1
e = rnorm(n, 0, sigma); beta = c(sort(c(runif(p, 0, 1)))));
data = c(rnorm(n * p, 5, 10)); X = matrix(data, ncol = p);
Y = X %*% beta + e;
LICnew(X = X, Y = Y, alpha = alpha, K = K, nk = nk)
```

MSEbeta	<i>Calculate MSE values for different beta estimation methods</i>
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Description

Calculate MSE values for different beta estimation methods

Usage

MSEbeta(X, Y, alpha, K, nk)

Arguments

X	The design matrix (observations).
Y	The response vector.
alpha	The significance level.
K	The number of subsets.
nk	The length of subsets (number of observations in each subset).

Value

A list containing:

MSECOR	The MSE of the COR beta estimator.
MSEAopt	The MSE of the A-optimal beta estimator.
MSEDopt	The MSE of the D-optimal beta estimator.
MSElic	The MSE of the LIC beta estimator.

MSEcom

Calculate the MSE values of the COR criterion in simulation

Description

Calculate the MSE values of the COR criterion in simulation

Usage

MSEcom(K = K, nk = nk, alpha = alpha, X = X, y = y)

Arguments

K	is the number of subsets
nk	is the length of subsets
alpha	is the significance level
X	is the observation matrix
y	is the response vector

Value

A list containing:

MSEx	The Mean Squared Error between the true beta and the estimate betax based on the COR.
MSEA	The Mean Squared Error between the true beta and the estimate betaA based on the least squares estimate for subset A.

MSEc	The Mean Squared Error between the true beta and the estimate betac based on the COR-selected subset.
MSEm	The Mean Squared Error between the true beta and the median estimator betamm across all subsets.
MSEa	The Mean Squared Error between the true beta and the mean estimator betaa across all subsets.

Examples

```
p=6;n=1000;K=2;nk=500;alpha=0.05;sigma=1
e=rnorm(n,0,sigma); beta=c(sort(c(runif(p,0,1)))));
data=c(rnorm(n*p,5,10));X=matrix(data, ncol=p);
y=X%%beta+e;
MSEcom(K=K,nk=nk,alpha=alpha,X=X,y=y)
```

MSEver	<i>Calculate the MSE values of the COR criterion for redundant data in simulation</i>
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Description

Calculate the MSE values of the COR criterion for redundant data in simulation

Usage

```
MSEver(K = K, nk = nk, alpha = alpha, X = X, y = y)
```

Arguments

K	is the number of subsets
nk	is the length of subsets
alpha	is the significance level
X	is the observation matrix
y	is the response vector

Value

A list containing:

minE	The minimum value of the error variance estimator.
Mcor	The MSE of the COR estimator.
Mx	The MSE of the estimator based on the subset with the maximum M.
MA	The MSE of the estimator based on the subset with the minimum W.

Examples

```
p=6;n=1000;K=2;nk=200;alpha=0.05;sigma=1
e=rnorm(n,0,sigma); beta=c(sort(c(runif(p,0,1)))));
data=c(rnorm(n*p,5,10));X=matrix(data, ncol=p);
y=X%%beta+e;
MSEver(K=K,nk=nk,alpha=alpha,X=X,y=y)
```

Index

* datasets

communities, [4](#)
ethylene_CO, [10](#)

beta_AD, [2](#)
beta_cor, [3](#)
beta_LW, [3](#)

communities, [4](#)
COR, [9](#)

ethylene_CO, [10](#)

LICbeta, [11](#)
LICnew, [12](#)

MSEbeta, [12](#)
MSEcom, [13](#)
MSEver, [14](#)