# Package 'DLPCA'

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Type Package

Title The Distributed Local PCA Algorithm

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**Description** Algorithm to handle with optimal subset selection for distributed local principal component analysis. The philosophy of the package is described in Guo G. (2020) <doi:10.1080/02331888.2020.1823979>.

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NeedsCompilation no

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Suggests testthat (>= 3.0.0)

**Config/testthat/edition** 3

**Repository** CRAN

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## DLPCA

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Application Application

## Description

Application data set

## Usage

data("Application")

# Format

The format is: int [1:48, 1:15] 6 9 7 5 6 7 9 9 9 4 ... - attr(\*, "dimnames")=List of 2 ...\$ : NULL ...\$ : chr [1:15] "FL" "APP" "AA" "LA" ...

## Details

It is the scoring of 15 indicators on 48 interviewees

## Examples

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

DLPCA

Distributed local PCA

# Description

Calculate the estimator on the DLPCA method

# Usage

DLPCA(X = X, n = n, p = p, m = m, K = K, L = L)

## Arguments

Х	is the original data matrix
n	is the sample size
р	is the number of variables
m	is the number of eigenvalues
К	is the number of nodes
L	is the number of subgroups

# gt2011

# Value

time	is the time cost
V	is the right singular matrix
Vm	is the m-right singular matrix
Smean	is the mean covariance matrix
MMSER	is the mean MSE values of the robust covariance matrix sub-estimators
MMSES	is the mean MSE values of the covariance matrix sub-estimators
MMSEX	is the mean MSE values of the sub-estimators of the matrix X
MSER	is the min MSE values of the robust covariance matrix sub-estimators
MSES	is the min MSE values of the covariance matrix sub-estimators
MSEX	is the min MSE values of the sub-estimators of the matrix X
wMSER	is the location of the min MSE values of the robust covariance matrix sub- estimators
wMSES	is the location of the min MSE values of the covariance matrix sub-estimators
wMSEX	is the location of the min MSE values of the sub-estimators of the matrix X
sigm	is the estimator of the covariance matrix of the matrix X

# Examples

```
data(Application)
X=Application
n=nrow(Application);p=ncol(Application)
m=5;L=4;K=4
DLPCA_result=DLPCA(X=X,n=n,p=p,m=m,K=K,L=L)
```

gt2011

Gas-Turbine CO and NOx Emission Data

# Description

Gas-Turbine CO and NOx Emission Data in 2011

# Usage

data("gt2011")

## Format

A data frame with 7411 observations on the following 11 variables.

- AT a numeric vector
- AP a numeric vector
- AH a numeric vector
- AFDP a numeric vector
- GTEP a numeric vector
- TIT a numeric vector
- TAT a numeric vector
- TEY a numeric vector
- CDP a numeric vector
- CO a numeric vector
- NOX a numeric vector

# Details

The dataset contains 36733 instances of 11 sensor measures aggregated over one hour, from a gas turbine located in Turkey for the purpose of studying flue gas emissions, namely CO and NOx.

## Source

Heysem Kaya, Department of Information and Computing Sciences, Utrecht University, 3584 CC, Utrecht, The Netherlands

## Examples

data(gt2011)

gt2012

Gas-Turbine CO and NOx Emission Data

## Description

Gas-Turbine CO and NOx Emission Data in 2012

## Usage

data("gt2012")

## gt2013

# Format

A data frame with 7628 observations on the following 11 variables.

- AT a numeric vector
- AP a numeric vector
- AH a numeric vector
- AFDP a numeric vector
- GTEP a numeric vector
- TIT a numeric vector
- TAT a numeric vector
- TEY a numeric vector
- CDP a numeric vector
- CO a numeric vector
- NOX a numeric vector

## Details

The dataset contains 36733 instances of 11 sensor measures aggregated over one hour, from a gas turbine located in Turkey for the purpose of studying flue gas emissions, namely CO and NOx.

## Source

Heysem Kaya, Department of Information and Computing Sciences, Utrecht University, 3584 CC, Utrecht, The Netherlands

## Examples

data(gt2012)

gt2013

Gas-Turbine CO and NOx Emission Data

# Description

Gas-Turbine CO and NOx Emission Data in 2013

## Usage

data("gt2013")

## Format

A data frame with 7152 observations on the following 11 variables.

- AT a numeric vector
- AP a numeric vector
- AH a numeric vector
- AFDP a numeric vector
- GTEP a numeric vector
- TIT a numeric vector
- TAT a numeric vector
- TEY a numeric vector
- CDP a numeric vector
- CO a numeric vector
- NOX a numeric vector

# Details

The dataset contains 36733 instances of 11 sensor measures aggregated over one hour, from a gas turbine located in Turkey for the purpose of studying flue gas emissions, namely CO and NOx.

#### Source

Heysem Kaya, Department of Information and Computing Sciences, Utrecht University, 3584 CC, Utrecht, The Netherlands

## Examples

data(gt2013)

gt2014

Gas-Turbine CO and NOx Emission Data

## Description

Gas-Turbine CO and NOx Emission Data in 2014

## Usage

data("gt2014")

## gt2015

# Format

A data frame with 7158 observations on the following 11 variables.

- AT a numeric vector
- AP a numeric vector
- AH a numeric vector
- AFDP a numeric vector
- GTEP a numeric vector
- TIT a numeric vector
- TAT a numeric vector
- TEY a numeric vector
- CDP a numeric vector
- CO a numeric vector
- NOX a numeric vector

## Details

The dataset contains 36733 instances of 11 sensor measures aggregated over one hour, from a gas turbine located in Turkey for the purpose of studying flue gas emissions, namely CO and NOx.

## Source

Heysem Kaya, Department of Information and Computing Sciences, Utrecht University, 3584 CC, Utrecht, The Netherlands

## Examples

data(gt2014)

gt2015

Gas-Turbine CO and NOx Emission Data

## Description

Gas-Turbine CO and NOx Emission Data in 2015

## Usage

data("gt2015")

# Format

A data frame with 7384 observations on the following 11 variables.

- AT a numeric vector
- AP a numeric vector
- AH a numeric vector
- AFDP a numeric vector
- GTEP a numeric vector
- TIT a numeric vector
- TAT a numeric vector
- TEY a numeric vector
- CDP a numeric vector
- CO a numeric vector
- NOX a numeric vector

## Details

The dataset contains 36733 instances of 11 sensor measures aggregated over one hour, from a gas turbine located in Turkey for the purpose of studying flue gas emissions, namely CO and NOx.

## Source

Heysem Kaya, Department of Information and Computing Sciences, Utrecht University, 3584 CC, Utrecht, The Netherlands

## Examples

data(gt2015)

Iris

Iris

## Description

Iris data set

## Usage

data("Iris")

## **MSEpca**

# Format

A data frame with 150 observations on the following 5 variables.

Sepal.length a numeric vector Sepal.width a numeric vector Petal.length a numeric vector Petal.width a numeric vector

Species a character vector

# Details

It contains 150 samples with 5 variables

# Source

Gaspar peninsula in Canada

# Examples

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

MSEpca

MSE on PCA

# Description

Caculate the MSE value on PCA

## Usage

MSEpca(V = V, X = X, n = n, p = p, m = m, K = K, L = L)

## Arguments

V	is the right singular matrix
Х	is the orignal data set
n	is the sample size
р	is the number of variables
m	is the number of eigenvalues
К	is the number of nodes
L	is the number of subgroups

## Value

MSEpca the MSE value on PCA

MSEpca

# Examples

```
data(Application)
X=Application
n=nrow(Application);p=ncol(Application)
m=5;L=4;K=4
DLPCA_result=DLPCA(X=X,n=n,p=p,m=m,K=K,L=L)
V=DLPCA_result$V
MSEpca_result=MSEpca(V=V,X=X,n=n,p=p,m=m,K=K,L=L)
MSE_PCA=MSEpca_result$MSEpca
```

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