ci.spls

---

**ci.spls**  
*Calculate bootstrapped confidence intervals of SPLS coefficients*

**Description**

Calculate bootstrapped confidence intervals of coefficients of the selected predictors and generate confidence interval plots.

**Usage**

```r
ci.spls( object, coverage=0.95, B=1000,  
       plot.it=FALSE, plot.fix="y",  
       plot.var=NA, K=object$K, fit=object$fit )
```

**Arguments**

- **object**: A fitted SPLS object.
- **coverage**: Coverage of confidence intervals. coverage should have a number between 0 and 1. Default is 0.95 (95% confidence interval).
- **B**: Number of bootstrap iterations. Default is 1000.
- **plot.it**: Plot confidence intervals of coefficients?
- **plot.fix**: If plot.fix="y", then plot confidence intervals of the predictors for a given response. If plot.fix="x", then plot confidence intervals of a given predictor across all the responses. Relevant only when plot.it=TRUE.
- **plot.var**: Index vector of responses (if plot.fix="y") or predictors (if plot.fix="x") to be fixed in plot.fix. The indices of predictors are defined among the set of the selected predictors. Relevant only when plot.it=TRUE.
- **K**: Number of hidden components. Default is to use the same K as in the original SPLS fit.
- **fit**: PLS algorithm for model fitting. Alternatives are "kernelpls", "widekernelpls", "simpls", or "oscorespls". Default is to use the same PLS algorithm as in the original SPLS fit.
Value

Invisibly returns a list with components:

- **cibeta**: A list with as many matrix elements as the number of responses. Each matrix element is p by 2, where i-th row of the matrix lists the upper and lower bounds of the bootstrapped confidence interval of the i-th predictor.
- **betahat**: Matrix of original coefficients of the SPLS fit.
- **lbmat**: Matrix of lower bounds of confidence intervals (for internal use).
- **ubmat**: Matrix of upper bounds of confidence intervals (for internal use).

Author(s)


References


See Also

correct.spls and spls.

Examples

data(mice)
  # SPLS with eta=0.6 & 1 hidden components
  f <- spls( mice$x, mice$y, K=1, eta=0.6 )
  # Calculate confidence intervals of coefficients
  ci.f <- ci.spls( f, plot.it=TRUE, plot.fix="x", plot.var=20 )
  # Bootstrapped confidence intervals
  cis <- ci.f$cibeta
  cis[[20]]  # equivalent, 'cis$1422478_a_at'

Description

Plot estimated coefficients of the selected predictors in the SPLS object.

Usage

coefplot.spls( object, nwin=c(2,2),
               xvar=c(1:length(object$A)), ylim=NA )
Arguments

- **object**: A fitted SPLS object.
- **nwin**: Vector of the number of rows and columns in a plotting area. Default is two rows and two columns, i.e., four plots.
- **xvar**: Index of variables to be plotted among the set of the selected predictors. Default is to plot the coefficients of all the selected predictors.
- **ylimit**: Range of the y axis (the coefficients) in the plot. If `ylimit` is not specified, the y axis of the plot has the range between the minimum and the maximum of all coefficient estimates.

Details

This plot is useful for visualizing coefficient estimates of a variable for different responses. Hence, the function is applicable only with multivariate response SPLS.

Value

NULL.

Author(s)


References


See Also

- `ci.spls`
- `correct.spls`
- `plot.spls`

Examples

```r
data(yeast)
# SPLS with eta=0.7 & 8 hidden components
f <- spls(yeast$x, yeast$y, K=8, eta=0.7)
# Draw estimated coefficient plot of the first four variables
# among the selected predictors
coefplot.spls( f, xvar=c(1:4), nwin=c(2,2) )
```
Correct the initial SPLS coefficient estimates based on bootstrapped confidence intervals

Description

Correct initial SPLS coefficient estimates of the selected predictors based on bootstrapped confidence intervals and draw heatmap of original and corrected coefficient estimates.

Usage

```r
correct.spls( object, plot.it=TRUE )
```

Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>object</td>
<td>An object obtained from the function <code>ci.spls</code>.</td>
</tr>
<tr>
<td>plot.it</td>
<td>Draw the heatmap of original coefficient estimates and corrected coefficient estimates?</td>
</tr>
</tbody>
</table>

Details

The set of the selected variables is updated by setting the coefficients with zero-containing confidence intervals to zero.

Value

Invisibly returns a matrix of corrected coefficient estimates.

Author(s)


References


See Also

`ci.spls`. 
Examples

```r
data(mice)
# SPLS with eta=0.6 & 1 latent components
f <- spls( mice$x, mice$y, K=1, eta=0.6 )
# Calculate confidence intervals of coefficients
ci.f <- ci.spls(f)
# Corrected coefficient estimates
cf <- correct.spls( ci.f )
cf[20,1:5]
```

---

**cv.sgpls**

*Compute and plot the cross-validated error for SGPLS classification*

---

**Description**

Draw heatmap of v-fold cross-validated misclassification rates and return optimal eta (thresholding parameter) and K (number of hidden components).

---

**Usage**

```r
cv.sgpls( x, y, fold=10, K, eta, scale.x=TRUE, plot.it=TRUE, br=TRUE, ftype='iden', n.core=8 )
```

---

**Arguments**

- **x**: Matrix of predictors.
- **y**: Vector of class indices.
- **fold**: Number of cross-validation folds. Default is 10-folds.
- **K**: Number of hidden components.
- **eta**: Thresholding parameter. eta should be between 0 and 1.
- **scale.x**: Scale predictors by dividing each predictor variable by its sample standard deviation?
- **plot.it**: Draw the heatmap of cross-validated misclassification rates?
- **br**: Apply Firth’s bias reduction procedure?
- **ftype**: Type of Firth’s bias reduction procedure. Alternatives are “iden” (the approximated version) or “hat” (the original version). Default is “iden”.
- **n.core**: Number of CPUs to be used when parallel computing is utilized.

---

**Details**

Parallel computing can be utilized for faster computation. Users can change the number of CPUs to be used by changing the argument `n.core`.
cv.spls

Value

Invisibly returns a list with components:

err.mat Matrix of cross-validated misclassification rates. Rows correspond to eta and columns correspond to number of components (K).
eta.opt Optimal eta.
K.opt Optimal K.

Author(s)

Dongjun Chung and Sunduz Keles.

References


See Also

print.sgpls, predict.sgpls, and coef.sgpls.

Examples

data(prostate)
set.seed(1)

# misclassification rate plot. eta is searched between 0.1 and 0.9 and
# number of hidden components is searched between 1 and 5
## Not run:
cv <- cv.sgpls(prostate$x, prostate$y, K = c(1:5), eta = seq(0.1,0.9,0.1),
               scale.x=FALSE, fold=5)
## End(Not run)

(sgpls(prostate$x, prostate$y, eta=cv$eta.opt, K=cv$K.opt, scale.x=FALSE))

---

cv.spls

Compute and plot cross-validated mean squared prediction error for SPLS regression

Description

Draw heatmap of v-fold cross-validated mean squared prediction error and return optimal eta (thresholding parameter) and K (number of hidden components).
Usage

cv.spls( x, y, fold=10, K, eta, kappa=0.5,
       select="pls2", fit="simpls",
       scale.x=TRUE, scale.y=FALSE, plot.it=TRUE )

Arguments

x  Matrix of predictors.
y  Vector or matrix of responses.
fold Number of cross-validation folds. Default is 10-folds.
K  Number of hidden components.
eta  Thresholding parameter. eta should be between 0 and 1.
kappa Parameter to control the effect of the concavity of the objective function and the closeness of original and surrogate direction vectors. kappa is relevant only when responses are multivariate. kappa should be between 0 and 0.5. Default is 0.5.
select PLS algorithm for variable selection. Alternatives are "pls2" or "simpls". Default is "pls2".
fit PLS algorithm for model fitting. Alternatives are "kernelpls", "widekernelpls", "simpls", or "oscorespls". Default is "simpls".
scale.x Scale predictors by dividing each predictor variable by its sample standard deviation?
scale.y Scale responses by dividing each response variable by its sample standard deviation?
plot.it Draw heatmap of cross-validated mean squared prediction error?

Value

Invisibly returns a list with components:

mspemat Matrix of cross-validated mean squared prediction error. Rows correspond to eta and columns correspond to the number of components (K).
eta.opt Optimal eta.
K.opt Optimal K.

Author(s)


References


See Also

print.spls, plot.spls, predict.spls, and coef.spls.
Examples

data(yeast)
set.seed(1)

# MSPE plot. eta is searched between 0.1 and 0.9 and 
# number of hidden components is searched between 1 and 10

## Not run:
  cv <- cv.spls(yeast$x, yeast$y, K = c(1:10), eta = seq(0.1,0.9,0.1))

# Optimal eta and K
  cv$eta.opt
cv$K.opt
  (spls(yeast$x, yeast$y, eta=cv$eta.opt, K=cv$K.opt))

## End(Not run)

---

cv.splsda

*Compute and plot cross-validated error for SPLSDA classification*

Description

Draw heatmap of v-fold cross-validated misclassification rates and return optimal eta (thresholding parameter) and K (number of hidden components).

Usage

```r
cv.splsda( x, y, fold=10, K, eta, kappa=0.5, 
classifier=c("lda","logistic"), scale.x=TRUE, plot.it=TRUE, n.core=8 )
```

Arguments

- **x**: Matrix of predictors.
- **y**: Vector of class indices.
- **fold**: Number of cross-validation folds. Default is 10-folds.
- **K**: Number of hidden components.
- **eta**: Thresholding parameter. eta should be between 0 and 1.
- **kappa**: Parameter to control the effect of the concavity of the objective function and the closeness of original and surrogate direction vectors. kappa should be between 0 and 0.5. Default is 0.5.
- **classifier**: Classifier used in the second step of SPLSDA. Alternatives are "logistic" or "lda". Default is "lda".
- **scale.x**: Scale predictors by dividing each predictor variable by its sample standard deviation?
- **plot.it**: Draw the heatmap of the cross-validated misclassification rates?
- **n.core**: Number of CPUs to be used when parallel computing is utilized.
Details

Parallel computing can be utilized for faster computation. Users can change the number of CPUs to be used by changing the argument \texttt{n.core}.

Value

Invisibly returns a list with components:

- \texttt{err.mat}: Matrix of cross-validated misclassification rates. Rows correspond to \texttt{eta} and columns correspond to number of components (\texttt{K}).
- \texttt{eta.opt}: Optimal \texttt{eta}.
- \texttt{K.opt}: Optimal \texttt{K}.

Author(s)

Dongjun Chung and Sunduz Keles.

References


See Also

- \texttt{print.splsda}, \texttt{predict.splsda}, and \texttt{coef.splsda}.

Examples

```r
data(prostate)
set.seed(1)
# misclassification rate plot. eta is searched between 0.1 and 0.9 and
# number of hidden components is searched between 1 and 5
## Not run: cv <- cv.splsda( prostate$x, prostate$y, K = c(1:5), eta = seq(0.1,0.9,0.1),
# scale.x=FALSE, fold=5 )
## End(Not run)

(splsda( prostate$x, prostate$y, eta=cv$eta.opt, K=cv$K.opt, scale.x=FALSE ))
```

lymphoma

\textit{Lymphoma Gene Expression Dataset}

Description

This is the Lymphoma Gene Expression dataset used in Chung and Keles (2010).

Usage

\texttt{data(lymphoma)}
**Format**

A list with two components:

- **x** Gene expression data. A matrix with 62 rows and 4026 columns.
- **y** Class index. A vector with 62 elements.

**Details**

The lymphoma dataset consists of 42 samples of diffuse large B-cell lymphoma (DLBCL), 9 samples of follicular lymphoma (FL), and 11 samples of chronic lymphocytic leukemia (CLL). DBLCL, FL, and CLL classes are coded in 0, 1, and 2, respectively, in y vector. Matrix x is gene expression data and arrays were normalized, imputed, log transformed, and standardized to zero mean and unit variance across genes as described in Dettling (2004) and Dettling and Beuhlmann (2002). See Chung and Keles (2010) for more details.

**Source**


**References**


**Examples**

```r
data(lymphoma)
lymphoma$x[1:5,1:5]
lymphoma$y
```

```
   mice   Mice Dataset
```

**Description**

This is the Mice dataset used in Chun and Keles (2010).

**Usage**

```r
data(mice)
```
Format

A list with two components:

- **x**  Marker map data. A matrix with 60 rows and 145 columns.
- **y**  Gene expression data. A matrix with 60 rows and 83 columns.

Details

The Mice dataset was published by Lan et al. (2006). Matrix **x** is the marker map consisting of 145 microsatellite markers from 19 non-sex mouse chromosomes. Matrix **y** is gene expression measurements of the 83 transcripts from liver tissues of 60 mice. This group of the 83 transcripts is one of the clusters analyzed by Chun and Keles (2010). See Chun and Keles (2010) for more details.

Source


References


Examples

```r
data(mice)
mice$x[1:5,1:5]
mice$y[1:5,1:5]
```

---

**plot.spls**

*Plot the coefficient path of SPLS regression*

Description

Provide the coefficient path plot of SPLS regression as a function of the number of hidden components (K) when eta is fixed.

Usage

```r
## S3 method for class 'spls'
plot( x, yvar=c(1:ncol(x$y)), ... )
```

Arguments

- **x**  A fitted SPLS object.
- **yvar**  Index vector of responses to be plotted.
- **...**  Other parameters to be passed through to generic plot.
Details

`plot.spls` provides the coefficient path plot of SPLS fits. The plot shows how estimated coefficients change as a function of the number of hidden components ($K$), when $\eta$ is fixed at the value used by the original SPLS fit.

Value

NULL.

Author(s)


References


See Also

`print.spls`, `predict.spls`, and `coef.spls`.

Examples

```r
data(yeast)
# SPLS with $\eta=0.7$ & 8 hidden components
f <- spls( yeast$x, yeast$y, K=8, eta=0.7 )
# Draw coefficient path plots for the first two responses
plot( f, yvar=c(1:2) )
```

Usage

```r
## S3 method for class 'sgpls'
predict( object, newx, type = c("fit","coefficient"),
         fit.type = c("class","response"), ... )
## S3 method for class 'sgpls'
coef( object, ... )
```
predict.sgpls

Arguments

object A fitted SGPLS object.
newx If type="fit", then newx should be the predictor matrix of test dataset. If newx is omitted, then prediction of training dataset is returned. If type="coefficient", then newx can be omitted.
type If type="fit", fitted values are returned. If type="coefficient", coefficient estimates of SGPLS fits are returned.
fit.type If fit.type="class", fitted classes are returned. If fit.type="response", fitted probabilities are returned. Relevant only when type="fit".
... Any arguments for predict.sgpls should work for coef.sgpls.

Details

Users can input either only selected variables or all variables for newx.

Value

Matrix of coefficient estimates if type="coefficient". Matrix of predicted responses if type="fit" (responses will be predicted classes if fit.type="class" or predicted probabilities if fit.type="response").

Author(s)

Dongjun Chung and Sunduz Keles.

References


See Also

print.sgpls.

Examples

data(prostate)
# SGPLS with eta=0.55 & 3 hidden components
f <- sgpls( prostate$x, prostate$y, K=3, eta=0.55, scale.x=FALSE )
# Print out coefficients
coef.f <- coef(f)
coef.f[ coef.f!=0, ]
# Prediction on the training dataset
(pred.f <- predict( f, type="fit" ))
predict.spls

Make predictions or extract coefficients from a fitted SPLS model

Description

Make predictions or extract coefficients from a fitted SPLS object.

Usage

## S3 method for class 'spls'
predict( object, newx, type = c("fit","coefficient"), ... )
## S3 method for class 'spls'
coef( object, ... )

Arguments

object A fitted SPLS object.
newx If type="fit", then newx should be the predictor matrix of test dataset. If newx is omitted, then prediction of training dataset is returned. If type="coefficient", then newx can be omitted.
type If type="fit", fitted values are returned. If type="coefficient", coefficient estimates of SPLS fits are returned.
... Any arguments for predict.spls should work for coef.spls.

Details

Users can input either only selected variables or all variables for newx.

Value

Matrix of coefficient estimates if type="coefficient". Matrix of predicted responses if type="fit".

Author(s)


References


See Also

plot.spls and print.spls.
Examples

data(yeast)
# SPLS with eta=0.7 & 8 latent components
f <- spls( yeast$x, yeast$y, K=8, eta=0.7 )
# Coefficient estimates of the SPLS fit
coef.f <- coef(f)
coef.f[1:5,]
# Prediction on the training dataset
pred.f <- predict( f, type="fit" )
pred.f[1:5,]

predict.splsda  Make predictions or extract coefficients from a fitted SPLSDA model

Description

Make predictions or extract coefficients from a fitted SPLSDA object.

Usage

## S3 method for class 'splsda'
predict( object, newx, type = c("fit","coefficient"),
         fit.type = c("class","response"), ... )
## S3 method for class 'splsda'
coef( object, ... )

Arguments

object  A fitted SPLSDA object.
newx    If type="fit", then newx should be the predictor matrix of test dataset. If newx is omitted, then prediction of training dataset is returned. If type="coefficient", then newx can be omitted.
type    If type="fit", fitted values are returned. If type="coefficient", coefficient estimates of SPLSDA fits are returned.
fit.type If fit.type="class", fitted classes are returned. If fit.type="response", fitted probabilities are returned. Relevant only when type="fit".
...     Any arguments for predict.splsda should work for coef.splsda.

Details

Users can input either only selected variables or all variables for newx.

Value

Matrix of coefficient estimates if type="coefficient". Matrix of predicted responses if type="fit" (responses will be predicted classes if fit.type="class" or predicted probabilities if fit.type="response").
Author(s)

Dongjun Chung and Sunduz Keles.

References


See Also

print.splsda.

Examples

```r
data(prostate)
# SPLSDA with eta=0.8 & 3 hidden components
f <- splsda(prostate$x, prostate$y, K=3, eta=0.8, scale.x=FALSE)
# Print out coefficients
coef.f <- coef(f)
coef.f[ coef.f!=0, ]
# Prediction on the training dataset
(pred.f <- predict( f, type="fit" ))
```

print.sgpls

Print function for a SGPLS object

Description

Print out SGPLS fit, the number and the list of selected predictors.

Usage

```r
## S3 method for class 'sgpls'
print( x, ... )
```

Arguments

- `x` A fitted SGPLS object.
- `...` Additional arguments for generic print.

Value

NULL.

Author(s)

Dongjun Chung and Sunduz Keles.
References


See Also

`predict.sgpls` and `coef.sgpls`.

Examples

```r
data(prostate)
# SGPLS with eta=0.55 & 3 hidden components
f <- sgpls(prostate$x, prostate$y, K=3, eta=0.55, scale.x=FALSE)
print(f)
```

Description

Print function for a SPLS object

Usage

```r
## S3 method for class 'spls'
print( x, ... )
```

Arguments

- `x` A fitted SPLS object.
- `...` Additional arguments for generic print.

Value

NULL.

Author(s)


References


See Also

`plot.spls`, `predict.spls`, and `coef.spls`. 

print.spls

Print function for a SPLS object
print.splsda

Examples

data(yeast)
  # SPLS with eta=0.7 & 8 hidden components
  f <- spls( yeast$x, yeast$y, K=8, eta=0.7 )
  print(f)


print.splsda                      Print function for a SPLSDA object

Description

Print out SPLSDA fits, the number and the list of selected predictors.

Usage

## S3 method for class 'splsda'
print( x, ... )

Arguments

x            A fitted SPLSDA object.
...           Additional arguments for generic print.

Value

NULL.

Author(s)

Dongjun Chung and Sunduz Keles.

References


See Also

predict.splsda and coef.splsda.

Examples

data(prostate)
  # SPLSDA with eta=0.8 & 3 hidden components
  f <- splsda( prostate$x, prostate$y, K=3, eta=0.8, scale.x=FALSE )
  print(f)
Prostate Tumor Gene Expression Dataset

Description

This is the Prostate Tumor Gene Expression dataset used in Chung and Keles (2010).

Usage

data(prostate)

Format

A list with two components:

x  Gene expression data. A matrix with 102 rows and 6033 columns.
y  Class index. A vector with 102 elements.

Details

The prostate dataset consists of 52 prostate tumor and 50 normal samples. Normal and tumor classes are coded in 0 and 1, respectively, in y vector. Matrix x is gene expression data and arrays were normalized, log transformed, and standardized to zero mean and unit variance across genes as described in Dettling (2004) and Dettling and Beuhlmann (2002). See Chung and Keles (2010) for more details.

Source


References


Examples

data(prostate)
prostate$x[1:5,1:5]$
prostate$y$
sgpls

**sgpls**

*Fit SGPLS classification models*

---

**Description**

Fit a SGPLS classification model.

**Usage**

```r
sgpls(x, y, K, eta, scale.x=TRUE,
      eps=1e-5, denom.eps=1e-20, zero.eps=1e-5, maxstep=100,
      br=TRUE, ftype='iden')
```

**Arguments**

- `x` Matrix of predictors.
- `y` Vector of class indices.
- `K` Number of hidden components.
- `eta` Thresholding parameter. `eta` should be between 0 and 1.
- `scale.x` Scale predictors by dividing each predictor variable by its sample standard deviation?
- `eps` An effective zero for change in estimates. Default is 1e-5.
- `denom.eps` An effective zero for denominators. Default is 1e-20.
- `zero.eps` An effective zero for success probabilities. Default is 1e-5.
- `maxstep` Maximum number of Newton-Raphson iterations. Default is 100.
- `br` Apply Firth’s bias reduction procedure?
- `ftype` Type of Firth’s bias reduction procedure. Alternatives are "iden" (the approximated version) or "hat" (the original version). Default is "iden".

**Details**

The SGPLS method is described in detail in Chung and Keles (2010). SGPLS provides PLS-based classification with variable selection, by incorporating sparse partial least squares (SPLS) proposed in Chun and Keles (2010) into a generalized linear model (GLM) framework. `y` is assumed to have numerical values, 0, 1, ..., G, where G is the number of classes subtracted by one.

**Value**

A `sgpls` object is returned. `print`, `predict`, `coef` methods use this object.

**Author(s)**

Dongjun Chung and Sunduz Keles.
References


See Also

print.sgpls, predict.sgpls, and coef.sgpls.

Examples

data(prostate)

# SGPLS with eta=0.6 & 3 hidden components
(f <- sgpls(prostate$x, prostate$y, K=3, eta=0.6, scale.x=FALSE))

# Print out coefficients
coef.f <- coef(f)
coef.f[coef.f!=0,]

---

spls

*Fit SPLS regression models*

Description

Fit a SPLS regression model.

Usage

spls( x, y, K, eta, kappa=0.5, select="pls2", fit="simpls",
 scale.x=TRUE, scale.y=FALSE, eps=1e-4, maxstep=100, trace=FALSE)

Arguments

- `x` Matrix of predictors.
- `y` Vector or matrix of responses.
- `K` Number of hidden components.
- `eta` Thresholding parameter. `eta` should be between 0 and 1.
- `kappa` Parameter to control the effect of the concavity of the objective function and the closeness of original and surrogate direction vectors. `kappa` is relevant only when responses are multivariate. `kappa` should be between 0 and 0.5. Default is 0.5.
- `select` PLS algorithm for variable selection. Alternatives are "pls2" or "simpls". Default is "pls2".
fit

PLS algorithm for model fitting. Alternatives are "kernelpls", "widekernelpls", "simpls", or "oscorespls". Default is "simpls".

scale.x

Scale predictors by dividing each predictor variable by its sample standard deviation?

scale.y

Scale responses by dividing each response variable by its sample standard deviation?

eps

An effective zero. Default is 1e-4.

maxstep

Maximum number of iterations when fitting direction vectors. Default is 100.

trace

Print out the progress of variable selection?

Details

The SPLS method is described in detail in Chun and Keles (2010). SPLS directly imposes sparsity on the dimension reduction step of PLS in order to achieve accurate prediction and variable selection simultaneously. The option select refers to the PLS algorithm for variable selection. The option fit refers to the PLS algorithm for model fitting and spls utilizes algorithms offered by the pls package for this purpose. See help files of the function plsr in the pls package for more details. The user should install the pls package before using spls functions. The choices for select and fit are independent.

Value

A spls object is returned. print, plot, predict, coef, ci.spls, coefplot.spls methods use this object.

Author(s)


References


See Also

print.spls, plot.spls, predict.spls, coef.spls, ci.spls, and coefplot.spls.

Examples

data(yeast)
# SPLS with eta=0.7 & 8 hidden components
(f <- spls(yeast$x, yeast$y, K=8, eta=0.7))

# Print out coefficients
coef.f <- coef(f)
coef.f[,1]

# Coefficient path plot
plot(f, yvar=1)
# Coefficient plot of selected variables
coeffplot.spls(f, xvar=c(1:4))

---

**splsda**  
*Fit SPLSDA classification models*

**Description**

Fit a SPLSDA classification model.

**Usage**

```r
splsda( x, y, K, eta, kappa=0.5,  
          classifier=c("lda","logistic"), scale.x=TRUE, ... )
```

**Arguments**

- `x`  
  Matrix of predictors.

- `y`  
  Vector of class indices.

- `K`  
  Number of hidden components.

- `eta`  
  Thresholding parameter. `eta` should be between 0 and 1.

- `kappa`  
  Parameter to control the effect of the concavity of the objective function and the closeness of original and surrogate direction vectors. `kappa` is relevant only for multicategory classification. `kappa` should be between 0 and 0.5. Default is 0.5.

- `classifier`  
  Classifier used in the second step of SPLSDA. Alternatives are "logistic" or "lda". Default is "lda".

- `scale.x`  
  Scale predictors by dividing each predictor variable by its sample standard deviation?

- `...`  
  Other parameters to be passed through to `spls`.

**Details**

The SPLSDA method is described in detail in Chung and Keles (2010). SPLSDA provides a two-stage approach for PLS-based classification with variable selection, by directly imposing sparsity on the dimension reduction step of PLS using sparse partial least squares (SPLS) proposed in Chun and Keles (2010). `y` is assumed to have numerical values, 0, 1, ..., G, where G is the number of classes subtracted by one. The option `classifier` refers to the classifier used in the second step of SPLSDA and `splsda` utilizes algorithms offered by `MASS` and `nnet` packages for this purpose. If `classifier="logistic"`, then either logistic regression or multinomial regression is used. Linear discriminant analysis (LDA) is used if `classifier="lda"`. `splsda` also utilizes algorithms offered by the `pls` package for fitting `spls`. The user should install `pls`, `MASS` and `nnet` packages before using `splsda` functions.
Value
A `splsda` object is returned. `print`, `predict`, `coef` methods use this object.

Author(s)
Dongjun Chung and Sunduz Keles.

References

See Also
`print.splsda`, `predict.splsda`, and `coef.splsda`.

Examples
```r
data(prostate)
# SPLSDA with eta=0.8 & 3 hidden components
f <- splsda( prostate$x, prostate$y, K=3, eta=0.8, scale.x=FALSE )
print(f)
# Print out coefficients
coef.f <- coef(f)
coef.f[ coef.f!=0, ]
```

---

**yeast**

*Yeast Cell Cycle Dataset*

Description
This is the Yeast Cell Cycle dataset used in Chun and Keles (2010).

Usage
```r
data(yeast)
```

Format
A list with two components:
- `x` ChIP-chip data. A matrix with 542 rows and 106 columns.
- `y` Cell cycle gene expression data. A matrix with 542 rows and 18 columns.
Details

Matrix \( y \) is cell cycle gene expression data (Spellman et al., 1998) of 542 genes from an \( \alpha \) factor based experiment. Each column corresponds to mRNA levels measured at every 7 minutes during 119 minutes (a total of 18 measurements). Matrix \( x \) is the chromatin immunoprecipitation on chip (ChIP-chip) data of Lee et al. (2002) and it contains the binding information for 106 transcription factors. See Chun and Keles (2010) for more details.

Source


References


Examples

```r
data(yeast)
yeast$x[1:5,1:5]
yeast$y[1:5,1:5]
```
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