Package ‘pdmclass’

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Type Package

Title Classification of Microarray Samples using Penalized Discriminant Methods

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Description This package can be used to classify microarray data using one of three penalized regression methods; partial least squares, principal components regression, or ridge regression.

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Depends Biobase (>= 1.4.22), R (>= 1.9.0), fibroEset, mda

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R topics documented:

pdmClass ................................................................. 1
pdmClass.cv ............................................................ 3
pdmGenes ................................................................. 4
predict.pls .............................................................. 6

Index 7

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pdmClass Function to Classify Microarray Data using Penalized Discriminant Methods

Description

This function is used to classify microarray data. Since the underlying model fit is based on penalized discriminant methods, there is no need for a pre-filtering step to reduce the number of genes.
Usage
pdmClass(formula, method = c("pls", "pcr", "ridge"), keep.fitted = TRUE, ...)

Arguments

formula  A symbolic description of the model to be fit. Details given below.
method  One of "pls", "pcr", "ridge", corresponding to partial least squares, principal
components regression and ridge regression.
keep.fitted  Boolean. Should the fitted values be kept? Default is TRUE, as this is necessary
for the plotting and predict functions.
...  Additional parameters to pass to method or fda. See fda for more information.

Details
The formula interface is identical to all other formula calls in R, namely \( Y \sim X \), where \( Y \) is a
numeric vector of class assignments and \( X \) is a matrix or data.frame containing the gene expression
values. Note that unlike most microarray analyses, in this instance the columns of \( X \) are genes and
rows are samples, so most calls will require something similar to \( Y \sim t(X) \).

Value
an object of class "fda". Use predict to extract discriminant variables, posterior probabilities or
predicted class memberships. Other extractor functions are coef, and plot.
The object has the following components:
percent.explained  the percent between-group variance explained by each dimension (relative to the
total explained.)
values  optimal scaling regression sum-of-squares for each dimension (see reference).
The usual discriminant analysis eigenvalues are given by \( \text{values} / (1 - \text{values}) \),
which are used to define percent.explained.
means  class means in the discriminant space. These are also scaled versions of the final
theta's or class scores, and can be used in a subsequent call to fda (this only
makes sense if some columns of theta are omitted—see the references).
theta.mod  (internal) a class scoring matrix which allows predict to work properly.
dimension  dimension of discriminant space.
prior  class proportions for the training data.
fit  fit object returned by method.
call  the call that created this object (allowing it to be update-able)
confusion  A 'confusion' matrix that shows how well the classifier works using the training
data.

Author(s)
James W. MacDonald and Debashis Ghosh, based on fda in the mda package of Trevor Hastie and
Robert Tibshirani, which was ported to R by Kurt Hornik, Brian D. Ripley, and Friedrich Leisch.
pdmClass.cv

References

http://www.sph.umich.edu/~ghoshd/COMPBIO/POPTSCORE
"Flexible Discriminant Analysis by Optimal Scoring" by Hastie, Tibshirani and Buja, 1994, JASA, 1255-1270.

Examples

library(fibroEset)
data(fibroEset)
y <- as.factor(pData(fibroEset)\[\,2\])
x <- t(exprs(fibroEset))
pdmClass(y ~ x)

pdmClass.cv

Leave One Out Crossvalidation

Description

This function performs a leave one out crossvalidation to estimate the accuracy of a classifier built using pdmClass.

Usage

pdmClass.cv(Y, X, method = c("pls", "pcr", "ridge"))

Arguments

Y  
A vector of factors giving the class assignments for the samples to be used in the crossvalidation.

X  
A matrix with samples in rows and observations in columns. Note that this is different than the usual paradigm for microarray data.

method  
One of "pls", "pcr", "ridge", corresponding to partial least squares, principal components regression and ridge regression.

Details

This function performs a leave one out crossvalidation, which can be used to estimate the accuracy of a classifier. Each sample is removed in turn and a classifier is built using the remaining samples. The class of the removed sample is then predicted using the classifier. This is repeated for each sample, resulting in a vector of predicted class assignments for each sample in the original training set.

Although far from perfect, this method can be used to estimate the accuracy of a given classifier without splitting data into a training and testing set.

Value

A vector of factors giving the predicted class assignments for each of the samples in the training set. A confusion matrix can be constructed using confusion.
Author(s)
James W. MacDonald

References
http://www.sph.umich.edu/~ghoshd/COMPBIO/POPTSCORE
"Flexible Discriminant Analysis by Optimal Scoring" by Hastie, Tibshirani and Buja, 1994, JASA, 1255-1270.

Examples
library(fibroEset)
data(fibroEset)
y <- as.factor(pData(fibroEset)[,2])
x <- t(exprs(fibroEset))
tmp <- pdmClass.cv(y, x)
confusion(tmp, y)

pdmGenes
A Function to output the Top Ranked Genes from a Penalized Discriminant Classifier

Description
After fitting a classifier, it is often desirable to output the most "interesting" genes for further validation. This function will output the top 'n' genes that discriminate between each class, along with an estimate of the stability of the observed rankings (see details for more information).

Usage
pdmGenes(formula = formula(data), method = c("pls", "pcr", "ridge"),
data = sys.frame(sys.parent()), weights, theta, dimension = J - 1,
eps = .Machine$double.eps, genelist = NULL, list.length = NULL, B = 100, ...)

Arguments
formula A symbolic description of the model to be fit. Details given below.
method One of "pls", "pcr", "ridge", corresponding to partial least squares, principal components regression and ridge regression.
data An optional data.frame that contains the variables in the model. If not found in data, the variables are taken from environment(formula), typically the environment from which pdmClass is called. Note that unlike most microarray analyses, in this case rows are samples and columns are genes.
weights An optional vector of sample weights.Defaults to 1.
theta An optional matrix of class scores, typically with less than J - 1 columns.
dimension The dimension of the solution, no greater than J - 1, where J is the number of classes. Defaults to J - 1.
pdmGenes


genelist A vector of gene names, one per gene.
lst.length The number of 'top' genes to output.
B The number of bootstrap samples to use for estimating stability. Defaults to 100.
More than this may take an inordinate amount of time.

Details

The formula interface is identical to all other formula calls in R, namely Y ~ X, where Y is a numeric vector of class assignments and X is a matrix or data.frame containing the gene expression values. Note that unlike most microarray analyses, in this instance the columns of X are genes and rows are samples, so most calls will require something similar to Y ~ t(X).

The dimension of the solution is typically J - 1, where J is the number of classes. The model fit uses contr.treatment contrasts, which means that all of the coefficients in the model are comparing the given class to a baseline class. Therefore, the genes listed are those that discriminate between a given class and the baseline. For instance, if there are three classes (characterized by a numeric vector of 1s, 2s, and 3s), then there will be two sets of 'top genes'. The first set will be those genes that discriminate between class 2 and class 1, whereas the second set will be the genes that discriminate between class 3 and class 1. The 'Y' vector will therefore need to be constructed to give the comparisons of interest.

Value

A list containing a data.frame for each comparison. The first column of each data.frame contains the gene names, and the second column contains the frequency that the gene was observed in the bootstrapped samples.

Author(s)

James W. MacDonald and Debashis Ghosh. Partial least squares and principal components regression based on code written by Mike Denham and contributed to StatLib. Model fit based on code from the mda package written by Trevor Hastie and Robert Tibshirani and ported to R by Kurt Hornik, Brian D. Ripley, and Friedrich Leisch.

References

http://www.sph.umich.edu/~ghoshd/COMPBIO/POPTSCORE

Examples

library(fibroEset)
data(fibroEset)
y <- as.factor(pData(fibroEset)[,2])
x <- t(exprs(fibroEset))
genesis <- featureNames(fibroEset)
pdmGenes(y ~ x, genelist = genes, list.length = 25, B = 10)
predict.pls

Classify Observations using Penalized Discriminant Methods

Description

These are functions that can be used to classify new samples (a test set) based on an existing classifier created using a training set.

Usage

```r
## S3 method for class 'pls'
predict(object, x, ...)
## S3 method for class 'svd'
predict(object, x, ...)
```

Arguments

- `object` An object created by a call to `pdmClass`.
- `x` A matrix of new observations in which rows are samples and columns are genes. If not supplied, prediction will be performed on the original training set.
- `...` Other variables passed to `predict`.

Value

A vector of predicted class assignments.

Author(s)

Debashis Ghosh

References

http://www.sph.umich.edu/~ghoshd/COMPBIO/POPTSCORE

Examples

```r
library(fibroEset)
data(fibroEset)
y <- as.numeric(pData(fibroEset)[,2])
x <- t(exprs(fibroEset))
genres <- featureNames(fibroEset)
tmp <- pdmClass(y ~ x)
predict(tmp)
```
Index

*Topic **classif**
  pdmClass, 1
  pdmClass.cv, 3
  pdmGenes, 4
  predict.pls, 6

*Topic **models**
  pdmClass, 1
  pdmClass.cv, 3
  pdmGenes, 4
  predict.pls, 6

*Topic **robust**
  pdmClass, 1
  pdmClass.cv, 3
  pdmGenes, 4
  predict.pls, 6

fda, 2

pdmClass, 1
pdmClass.cv, 3
pdmGenes, 4
predict.pls, 6
predict.svd(predict.pls), 6