Package 'QDRS'

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Type Package

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Author Danqing Xu			
Maintainer Danqing Xu <xud@pstat.ucsb.edu></xud@pstat.ucsb.edu>			
Description Functions for calculating quantitative disease risk scores: (1) Phenotype Risk Score (PheRS), (2) a spectral approach (Eigen), (3) principal component approaches (PCA) that include individual PCs and a linear combination of multiple PCs (LPC), (4), Non-negative Matrix Factorization with Rank 1 (NMF1), and (5) Latent Variable Score based on an unsupervised multivariate mixed model framework (LVS).			
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```

curve.palettes

Curve Palettes

Description

A vector of colors for performance curve

Usage

```
data(curve.palettes)
```

Format

An object of class character of length 10.

Examples

```
data(curve.palettes)
## Not run:
curve.palettes
## End(Not run)
```

dist.plot

Distribution Plots of QDRS

Description

Generate a pdf file with a density plot, a boxplot and a prevalence plot for the resulting scores. These three plots show the distribution of scores by disease status for visualization of risk stratification.

Usage

```
dist.plot(
   disease,
   output.date = NULL,
   score.mat,
   score.name,
   group,
   cutoff = c(seq(0.05, 0.45, by = 0.05), seq(0.5, 1, by = 0.01)),
   unknown.show = FALSE
)
```

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Arguments

disease A string of disease name.

 $\hbox{output.date}\qquad \hbox{ An output date. The default is system date (today)}.$

score.mat A resulting score matrix with multiple types of scores.

score.name The score of interest.
group A grouping vector.

cutoff A vector of cutoff points to set the bins of prevalence plot. The default vector

has 60 bins.

unknown. show A logical value indicates whethe unknowns should be used for plotting.

Value

It produces a pdf file of plots and returns the list of three distribution plots.

Examples

```
## Not run:
group1 = example.scores$group
group1[group1 != "Control"] = "Case"
set.seed(830)
na.ind = sample(1:length(group), size = 1000)
group1[na.ind] = NA
res = dist.plot(disease = "Disease",
    score.mat = example.scores$score.mat,
    score.name = "LPC",
    group = group1,
    unknown.show = T)
## End(Not run)
```

EHR

An example data set

Description

It contains sample.set a set of 95 features for 5,000 subjects, sample.group a vector of group with two levels "Case" and "Control", and training a logical vector that indicates the usage for training.

Usage

data(EHR)

Format

An object of class list of length 3.

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Examples

```
data(EHR)
## Not run:
# check the number of cases and controls
table(EHR$sample.group)
# check the prevalence of features
colMeans(EHR$sample.set)
## End(Not run)
```

eigen.score

Eigen Score

Description

Compute Eigen weights and scores. This method is unsupervised, and assigns weights that are proportional to the balanced accuracies (the average between the sensitivity and the specificity) of the input feature.

Usage

```
eigen.score(X, training, scale = TRUE)
```

Arguments

X The original data set that include training and test sets. It should be a matrix of

numbers.

training A logical or index vector to indicate whether the subject belongs to the training

set.

scale A logical value to indicate whether the input features need to be scaled.

Value

It returns a list of following components:

weights The Eigen weights for input features.

scores The resulting Eigen scores for the whole set.

References

Iuliana Ionita-Laza, Kenneth McCallum, Bin Xu, and Joseph D Buxbaum. A spectral approach integrating functional genomic annotations for coding and noncoding variants. Nature genetics, 48(2):214, 2016.

```
## Not run:
data(EHR)
res1 <- eigen.score(EHR$sample.set, EHR$training)
## End(Not run)</pre>
```

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example.scores	An example resulting score set

Description

It contains score.mat a matrix of five types of QDRSs for $5{,}000$ subjects, score.names a vector of score names, group a vector of group with two levels "Case" and "Control", and group.levels the unique grouping values.

Usage

```
data(example.scores)
```

Format

An object of class list of length 4.

Examples

```
data(example.scores)
## Not run:
# check the number of cases and controls
table(example.scores$.group)
## End(Not run)
```

LPC

Linear Combination of Principal Components

Description

Computes LPC weights and scores based on an almost unsupervised method that combines multiple PCs and only requires weak labels to help select the signs of individual PCs.

Usage

```
LPC(X, group, training, scale = TRUE)
```

Arguments

X	The original data set that include training and test sets. It should be a matrix of numbers.
group	A vector that indicate cases ("Case") and controls ("Control"). NA is allowed, and means that the observation is not used in individual PC sign determination.
training	A logical or index vector to indicate whether the subject belongs to the training set.
scale	A logical value to indicate whether the input features need to be scaled.

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Value

It returns a list of following components:

The number of significant PCs (eigenvalues) suggested by Tracy-Widom test, a vector of the sign of individual PC.

pc.sign A vector of the sign of individual PC.

weights The LPC weights for input features.

scores The resulting LPC scores for the whole set.

The resulting Li C scores for the whole s

Examples

```
## Not run:
data(EHR)
res1 <- LPC(X = EHR$sample.set, group = EHR$sample.group, training = EHR$training)
## End(Not run)</pre>
```

LVS.fit

Fit a Latent Variable Model

Description

LVS. fit computes a generalized lineatr latent variable model for multivariate bernoulli data, which assumes one latent variable.

Usage

```
LVS.fit(X, familiy, starting.choice = "random", p.seed = 3080)
```

Arguments

X The training data set. It should be a matrix of binary numbers.

starting.choice

Starting values can be generated by fitting model without latent variables, and applying factorial analysis to residuals to get starting values for latent variables and their coefficients (starting.choice = "res"). Another options are to use zeros as a starting values (starting.choice = "zero") or initialize starting values for

latent variables with (n x 1) matrix. Defaults to "random".

p. seed A random seed for model fitting.

family distribution function for responses. Options are poisson(link = "log"), "nega-

tive.binomial" (with log link), binomial(link = "probit") , zero inflated poisson ("ZIP"), gaussian(link = "identity"), "gamma" (with log link), and "ordinal".

Value

An object of class "gllvm" includes the following components.

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LVS.score

Latent Variable Score

Description

Computes Latent Variable Scores based on an unsupervised multivariate mixed model framework for multivariate bernoulli data that assumes one latent variable.

Usage

```
LVS.score(
   X,
   Y = NULL,
   family = "binomial",
   starting.choice = "random",
   p.seed = 3080
)
```

Arguments

X A training data set. It should be a matrix of binary numbers.

Y A new data set. It should be a matrix of binary numbers. Default is NULL, then

X will be used in prediction.

family distribution function for responses. Options are poisson(link = "log"), "nega-

tive.binomial" (with log link), binomial(link = "probit"), zero inflated poisson ("ZIP"), gaussian(link = "identity"), "gamma" (with log link), and "ordinal".

starting.choice

Starting values can be generated by fitting model without latent variables, and applying factorial analysis to residuals to get starting values for latent variables and their coefficients (starting.choice = "res"). Another options are to use zeros as a starting values (starting.choice = "zero") or initialize starting values for

latent variables with (n x 1) matrix. Defaults to "random".

p. seed A random seed for model fitting.

Value

lvs The resulting LVSs for the new set Y.

References

Jenni Niku, Wesley Brooks, Riki Herliansyah, Francis KC Hui, Sara Taskinen, and David I Warton. Efficient estimation of generalized linear latent variable models. PloS one, 14(5), 2019.

```
## Not run:
data(EHR)
sample.set = EHR$sample.set
res1 <- LVS.score(X = sample.set[seq(1,50),])
## End(Not run)</pre>
```

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multiple.pr.curve

Multiple PR curves

Description

Generate a pdf file with multiple plots for several group level contrast. Each plot consists of PR curves for different quantitative disease scores, and the corrsponding AUPRs for performance comparison.

Usage

```
multiple.pr.curve(
   disease,
   output.date = NULL,
   score.mat,
   score.names,
   group,
   group.levels,
   legend.pos = "topright",
   pairs.sub = NULL
)
```

Arguments

disease A string of disease name. An output date. The default is system date (today). output.date A resulting score matrix with multiple types of scores. score.mat A vector that indicates the scores of interest for comparison. score.names A grouping vector. group group.levels A vector of grouping levels that reflects the severity from low to high. The typical first element represents control or healthy status. A vector of legend position. The default is "bottomright" for all plots. legend.pos pairs.sub An index vector for subset of group level pairs, e.g. the pairs contrasting control vs. other case stages.

Value

It produces a pdf file of plots.

```
## Not run:
multiple.pr.curve(disease = "Disease",
    score.mat = example.scores$score.mat,
    score.names = example.scores$score.names,
    group = example.scores$group,
    group.levels = example.scores$group.levels,
    legend.pos = "topright", pairs.sub = 1:5)
## End(Not run)
```

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multiple.roc.curve

Multiple ROC curves

Description

Generate a pdf file with multiple plots for several group level contrast. Each plot consists of ROC curves for different quantitative disease scores, and the corrsponding AUROCs for performance comparison.

Usage

```
multiple.roc.curve(
   disease,
   output.date = NULL,
   score.mat,
   score.names,
   group,
   group.levels,
   legend.pos = "bottomright",
   pairs.sub = NULL
)
```

Arguments

disease A string of disease name. An output date. The default is system date (today). output.date A resulting score matrix with multiple types of scores. score.mat A vector that indicates the scores of interest for comparison. score.names A grouping vector. group group.levels A vector of grouping levels that reflects the severity from low to high. The typical first element represents control or healthy status. A vector of legend position. The default is "bottomright" for all plots. legend.pos pairs.sub An index vector for subset of group level pairs, e.g. the pairs contrasting control vs. other case stages.

Value

It produces a pdf file of plots.

```
## Not run:
multiple.roc.curve(disease = "Disease",
    score.mat = example.scores$score.mat,
    score.names = example.scores$score.names,
    group = example.scores$group,
    group.levels = example.scores$group.levels,
    legend.pos = "bottomright", pairs.sub = 1:5)
## End(Not run)
```

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NMF1

Non-negative Matrix Factorization with Rank 1 Score

Description

Computes NMF1 scores with factor loadings constrained to be non-negative.

Usage

```
NMF1(X, p.seed = 123)
```

Arguments

X The original data set. It should be a matrix of numbers.

p. seed A random seed for model fitting.

Value

It returns a list of following components:

weights The NMF1 weights for input features.

scores The resulting NMF1 scores for the whole set.

Examples

```
## Not run:
data(EHR)
sample.set = EHR$sample.set
res1 <- NMF1(X = sample.set[seq(1,50),])
## End(Not run)</pre>
```

pairwise.auc

Pairwise Area Under the ROC Curve

Description

Calculates AUROC for pairwise levels of group contrast .

Usage

```
pairwise.auc(x, g, format = "1")
```

Arguments

x A response vector.

g A grouping vector or factor.

format Format of result. The default is "l", long format. The other option is square

format.

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Value

A data frame with group 1, group 2, and the AUROC with comparison between them.

Examples

```
## Not run:
data(EHR)
x = c(rnorm(2500),runif(2500))
res = pairwise.auc(x, EHR$sample.group)
## End(Not run)
```

pairwise.upr

Pairwise Area Under the PR Curve

Description

Calculates AUPRC for pairwise group levels contrast.

Usage

```
pairwise.upr(x, g, format = "1")
```

Arguments

x A response vector.

g A grouping vector or factor.

format Format of result. The default is "1", long format. The other option is square

format.

Value

A data frame with group 1, group 2, and the AUPRC with comparison between them.

```
## Not run:
data(EHR)
x = c(rnorm(2500),runif(2500))
res = pairwise.upr(x, EHR$sample.group)
## End(Not run)
```

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pairwise.wilcox

Pairwise Wilcoxon Rank Sum Tests

Description

Calculates pairwise comparisons between group levels with corrections for multiple testing.

Usage

```
pairwise.wilcox(x, g, p.adjust.method = "bonferroni")
```

Arguments

```
x A response vector.
```

g A grouping vector or factor.

p.adjust.method

Method for adjusting p values (can be abbreviated): "holm", "hochberg", "hommel", "bonferroni", "BH", "BY", "fdr", "none". The default is "bonferroni".

Value

A data frame with group 1, group 2, and the Wilcoxon Rank Sum test p value of comparison between them.

Examples

```
## Not run:
data(EHR)
x = c(rnorm(2500),runif(2500))
res = pairwise.wilcox(x, EHR$sample.group)
## End(Not run)
```

PC

Individual Principal Components with Selected Signs

Description

Computes individual PC weights and scores. The approach only requires weak labels to help select the signs of individual PCs.

Usage

```
PC(X, group, training, scale = TRUE, pc.num)
```

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Arguments

X The original data set that include training and test sets. It should be a matrix of

numbers.

group A grouping vector or factor that indicate cases ("Case") and controls ("Control").

NA is allowed, and means that the observation is not used in individual PC sign

determination.

training A logical or index vector to indicate whether the subject belongs to the training

set.

scale A logical value to indicate whether the input features need to be scaled.

pc.num The vector of desired individual PCs.

Value

It returns a list of following components:

weights The selected individual PCs' weights for input features. scores The resulting individual PC scores for the whole set.

Examples

```
## Not run:
data(EHR)
res1 <- PC(X = EHR$sample.set,
  group = EHR$sample.group,
  training = EHR$training, pc.num = 1:2)
## End(Not run)</pre>
```

PheRS

Phenotype Risk Score

Description

Computes Phenotype Risk Scores based on a approach that was proposed in the context of rare Mendelian phenotypes.

Usage

```
PheRS(X, feature.prevalence = NULL, group = NULL)
```

Arguments

X The original data set that include training and test sets. It should be a matrix of binary numbers.

feature.prevalence

A vector of feature prevalence.

group A vector that indicate cases ("Case") and controls ("Control"). NA is allowed.

The default is NULL, meaning all observations will be used in prevalence computation if the feature prevalence vector is not provided by the user. Otherwise,

only the controls will be used.

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Value

It returns a list of following components:

weights The PheRS weights for input features.
scores The resulting PheRSs for the whole set.

References

Lisa Bastarache, Jacob J Hughey, Scott Hebbring, Joy Marlo, Wanke Zhao, Wanting T Ho, Sara L Van Driest, Tracy L McGregor, Jonathan D Mosley, Quinn S Wells, et al. Phenotype risk scores identify patients with unrecognized mendelian disease patterns. Science, 359(6381):1233–1239, 2018.

Examples

```
## Not run:
data(EHR)
res1 <- PheRS(X = EHR$sample.set, group = EHR$sample.group)
res2 <- PheRS(X = EHR$sample.set, group = NULL)
## End(Not run)</pre>
```

rankOne.R

R Matrix with Rank One for Eigen Approach

Description

Computes an estimated rank-one matrix R with unit-norm eigenvector v. Up to a sign ambiguity, the entries of v are proportional to the balanced accuracies (the average between the sensitivity and the specificity) of the input features.

Usage

```
rankOne.R(Qmat)
```

Arguments

Qmat

The correlation matrix Q.

Value

Rmat An estimate of the rank-one R matrix

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