Package 'RankProd'

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Title Rank Product method for identifying differentially expressed genes with application in meta-analysis

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Depends R (>= 1.9.0)

Imports graphics

Description Non-parametric method for identifying differentially

expressed (up- or down- regulated) genes based on the estimated percentage of false predictions (pfp). The method can combine data sets from different origins (meta-analysis) to increase the power of the identification.

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biocViews DifferentialExpression

R topics documented:

arab	2
golub	2
lymphoma	3
plotGene	4
plotRP	6
RP	7
RPadvance	9
RSadvance	11
topGene	14

Index

arab

Description

This data if from Affy ATH1 array experiments of genomic response to brassinosteroid in Arabidopsis conducted at two laoratories. The data set contains random selected 500 genes and 10 samples, 6 from lab 1 and 4 from lab 2. Data was pre-processed by RMA

Usage

data(arab)

Value

arab	matrix of gene expression levels of 500 genes from 10 samples, rows correspond to genes and columns to mRNA samples.
arab.cl	numeric vector indicating the treatment class, 5 brassinosteroid-treated cases (code 1) and 5 control cases (code 0)
arab.gnames	character vector containing the AffyID of the 500 genes for the expression ma- trix arab
arab.origin	numeric vector indicating the origin of the samples, 6 samples from lab 1 (code 1) and 4 samples from lab 2 (code 2)

References

Nemhauser JL, Mockler TC, Chory J. Interdependency of brassinosteroid and auxin signaling in Arabidopsis. *PLoS Biol.* 2004 21460.

Microarray data from AtGenExpress (http://arabidopsis.org/info/expression/ATGenExpress.jsp)

golub

sub set of the Gene expression dataset from Golub et al. (1999)

Description

Gene expression data (500 genes and 38 tumor mRNA samples) from the leukemia microarray study of Golub et al. (1999). Original data set contain 3051 genes

Usage

data(golub)

16

lymphoma

Value

golub	matrix of gene expression levels for the 38 tumor mRNA samples, rows correspond to genes (500 genes) and columns to mRNA samples.
golub.cl	numeric vector indicating the tumor class, 27 acute lymphoblastic leukemia (ALL) cases (code 0) and 11 acute myeloid leukemia (AML) cases (code 1).
golub.gnames	a matrix containing the names of the 500 genes for the expression matrix golub. The three columns correspond to the gene index, ID, and Name, respectively.

Source

Golub et al. (1999). Molecular classification of cancer: class discovery and class prediction by gene expression monitoring, *Science*, Vol. 286:531-537. http://www-genome.wi.mit.edu/MPR/.

References

S. Dudoit, J. Fridlyand, and T. P. Speed (2002). Comparison of discrimination methods for the classification of tumors using gene expression data. *Journal of the American Statistical Association*, Vol. 97, No. 457, p. 77–87.

lymphoma

Subset of the Intensity data for 8 cDNA slides with CLL and DLBL samples from the Alizadeh et al. paper in Nature 2000

Description

8 cDNA chips from Alizadeh lymphoma paper

Usage

data(lymphoma)

Format

lymphoma is an exprSet containing the data from 8 chips from the lymphoma data set by Alizadeh et al. (see references). Each chip represents two samples: on color channel 1 (CH1, Cy3, green) the common reference sample, and on color channel 2 (CH2, Cy5, red) the various disease samples. See pData(lymphoma). The 9216x16 matrix exprs(lymphoma) contains the background-subtracted spot intensities (CH1I-CH1B and CH2I-CH2B, respectively).

Details

The chip intensity files were downloaded from the Stanford microarray database. Starting from the link below, this was done by following the links *Published Data -> Alizadeh AA, et al. (2000) Nature 403(6769):503-11 -> Data in SMD -> Display Data*, and selecting the following 8 slides:

```
lc7b019
lc7b047
lc7b048
lc7b056
lc7b057
lc7b058
lc7b069
lc7b070
```

Then, the script makedata.R from the scripts subdirectory of this package was run to generate the R data object.

Source

http://genome-www5.stanford.edu/MicroArray/SMD

References

A. Alizadeh et al., Distinct types of diffuse large B-cell lymphoma identified by gene expression profiling. Nature 403(6769):503-11, Feb 3, 2000.

nl	.otGene	
PT	.ocuene	

Graphical Display of The expression levels

Description

Plot the expression values and print statistical results for each individual gene based on a user query

Usage

Arguments

gene.to.plot	Name of the gene to be plotted
х	the value returned by function RP, RPadvance or RSadvance
gene.names	Names of all genes in the data set. If "NULL", rownames of the data will be used
data	the same as that used in RP or RPadvance
cl	the same as that used in RP or RPadvance

plotGene

origin	a vector containing the origin labels of the sample. e.g. for the data sets gener- ated at multiple laboratories, the label is the same for samples within one lab and different for samples from different labs. The same as that used in RPadvance
logged	if "TRUE", data has bee logged, otherwise set it to "FALSE"
logbase	base used when taking log, used to restore the fold change.The default value is 2, this will be ignored if logged=FALSE

Value

A graphical display of the expression levels of the input gene. The estimated statistics for differential expression will be printed on the plot as well as output in the screen. The statistics include: F.C.:fold-change under each dataset if multiple datasets are used AveFC: average fold-change across all datasets pfp(pval): estimated percentage of false prediction (p-value) for differential expression under each of the two tests: up-regulation in class 2 compared with classs 1 and down-regulation in class 2 compared with class 1

Author(s)

Fangxin Hong <fhong@salk.edu>

See Also

topGene RP RPadvance RSadvance

Examples

```
# Load the data of Golub et al. (1999). data(golub)
#contains a 3051x38 gene expression
# matrix called golub, a vector of length called golub.cl
#that consists of the 38 class labels,
# and a matrix called golub.gnames whose third column contains the gene names.
data(golub)
#use a subset of data as example, apply the rank product method
subset <- c(1:4,28:30)
#Setting rand=123, to make the results reproducible,
#identify genes that are up-regulated in class 2
#(class label =1)
RP.out <- RP(golub[,subset],golub.cl[subset], rand=123)</pre>
```

```
#plot the results
plotRP(RP.out,cutoff=0.05)
```

plotRP

Description

Plot a graphical display of the estimated pfp vs number of identified genes

Usage

```
plotRP(x, cutoff=NULL)
```

Arguments

х	the value returned by function RP, RPadvance or RSadvance
cutoff	threshold in pfp used to select genes

Value

A graphical display of the estimated pfp vs number of identified genes, which is also the gene rank of its original rank product/sum across all comparison. If cutoff is sepcified, a horizontal line will be plotted on the graphic to indicate the positon of the cutoff point, and all genes identified will be marked red.

Two plots will be displayed, one for the identification of up-regulated genes in class 2, one for the identification of down-regulated genes in class 2

Author(s)

Fangxin Hong <fhong@salk.edu>

See Also

topGene RP RPadvance RSadvance

Examples

```
# Load the data of Golub et al. (1999). data(golub)
#contains a 3051x38 gene expression
# matrix called golub, a vector of length called golub.cl
#that consists of the 38 class labels,
# and a matrix called golub.gnames whose third column contains the gene names.
data(golub)
#use a subset of data as example, apply the rank product method
subset <- c(1:4,28:30)
#Setting rand=123, to make the results reproducible,
#identify genes that are up-regulated in class 2
```

```
#(class label =1)
RP.out <- RP(golub[,subset],golub.cl[subset], rand=123)</pre>
```

```
#plot the results
plotRP(RP.out,cutoff=0.05)
```

RP

Rank Product Analysis of Microarray

Description

Perform rank product method to identify differentially expressed genes. It is possible to do either a one-class or two-class analysis.

Usage

RP(data,cl,num.perm=100,logged=TRUE, na.rm=FALSE,gene.names=NULL,plot=FALSE, rand=NULL, huge=FALSE)

Arguments

data	the data set that should be analyzed. Every row of this data set must correspond to a gene.
cl	a vector containing the class labels of the samples. In the two class unpaired case, the label of a sample is either 0 (e.g., control group) or 1 (e.g., case group). For one class data, the label for each sample should be 1.
num.perm	number of permutations used in the calculation of the null density. Default is 'num.perm=100'.
logged	if "TRUE", data has bee logged, otherwise set it to "FALSE"
na.rm	if 'FALSE' (default), the NA value will not be used in computing rank. If 'TRUE', the missing values will be replaced by the gene-wise mean of the non-missing values. Gene with all values missing will be assigned "NA"
gene.names	if "NULL", no gene name will be assigned to the estimated percentage of false positive predictions (pfp).
plot	If "TRUE", plot the estimated pfp verse the rank of each gene.
rand	if specified, the random number generator will be put in a reproducible state using the rand value as seed.
huge	If "TRUE", use an alternative method for computation. Using considerably less memory, this allows the Rank Product to be calculated for larger input data (see details). However, the result will not contain the Orirank value.
	For input with n rows, m=m1+m2 columns for two classes, and k permutations, memory requirements are 2n with huge=TRUE instead of n*k+n*m1*m2.

8

A result of identifying differentially expressed genes between two classes. The identification consists of two parts, the identification of up-regulated and down-regulated genes in class 2 compared to class 1, respectively.

pfp	estimated percentage of false positive predictions (pfp) up to the position of each gene under two identification each
pval	estimated pvalue for each gene being up- and down-regulated
RPs	Original rank-product of each genes for two dentificaiton each
RPrank	rank of the rank product of each genes
Orirank	original rank in each comparison, which is used to construct rank product. Not present if huge=TRUE is used.
AveFC	fold change of average expression under class 1 over that under class 2. log-fold change if data is in log scaled, original fold change if data is unlogged.

Note

Percentage of false prediction (pfp), in theory, is equivalent of false discovery rate (FDR), and it is possible to be large than 1.

The function looks for up- and down- regulated genes in two seperate steps, thus two pfps and pvalues are computed and used to identify gene that belong to each group.

This function is suitable to deal with data from a single origin, e.g. single experiment. If the data has different origin, e.g. generated at different laboratories, please refer RP.advance.

Author(s)

Fangxin Hong <fhong@salk.edu>

References

Breitling, R., Armengaud, P., Amtmann, A., and Herzyk, P.(2004) Rank Products: A simple, yet powerful, new method to detect differentially regulated genes in replicated microarray experiments, *FEBS Letter*, 57383-92

See Also

topGene RPadvance plotRP

Examples

Load the data of Golub et al. (1999). data(golub) # contains a 3051x38 gene expression # matrix called golub, a vector of length called golub.cl # that consists of the 38 class labels, # and a matrix called golub.gnames whose third column # contains the gene names. data(golub)

RPadvance

```
#use a subset of data as example, apply the rank
#product method
subset <- c(1:4,28:30)
#Setting rand=123, to make the results reproducible,
RP.out <- RP(golub[,subset],golub.cl[subset],rand=123)
# class 2: label =1, class 1: label = 0
#pfp for identifying genes that are up-regulated in class 2
#pfp for identifying genes that are down-regulated in class 2
head(RP.out$pfp)</pre>
```

```
RPadvance
```

Advanced Rank Product Analysis of Microarray

Description

Advance rank product method to identify differentially expressed genes. It is possible to combine data from different studies, e.g. data sets generated at different laboratories.

Usage

Arguments

data	the data set that should be analyzed. Every row of this data set must correspond to a gene.
cl	a vector containing the class labels of the samples. In the two class unpaired case, the label of a sample is either 0 (e.g., control group) or 1 (e.g., case group). For one group data, the label for each sample should be 1.
origin	a vector containing the origin labels of the sample. e.g. for the data sets gen- erated at multiple laboratories, the label is the same for samples within one lab and different for samples from different labs.
num.perm	number of permutations used in the calculation of the null density. Default is 'B=100'.
logged	if "TRUE", data has bee logged, otherwise set it to "FALSE"
na.rm	if 'FALSE' (default), the NA value will not be used in computing rank. If 'TRUE', the missing values will be replaced by the genewise mean of the non- missing values. Gene will all value missing will be assigned "NA"

gene.names	if "NULL", no gene name will be attached to the estimated percentage of false prediction (pfp).
plot	If "TRUE", plot the estimated pfp verse the rank of each gene
rand	if specified, the random number generator will be put in a reproducible state.
huge	If "TRUE", use an alternative method for computation. Using considerably less memory, this allows the Rank Product to be calculated for larger input data (see details). However, the result will not contain the Orirank value.
	For input with n rows, m=m1+m2 columns for two classes, and k permutations, memory requirements are 2n with huge=TRUE instead of n*k+n*m1*m2.

Value

A result of identifying differentially expressed genes between two classes. The identification consists of two parts, the identification of up-regulated and down-regulated genes in class 2 compared to class 1, respectively.

pfp	estimated percentage of false positive predictions (pfp) up to the position of each gene under two identification each
pval	estimated pvalue for each gene being up- and down-regulated
RPs	Original rank-product of each genes for two i dentificaiton each
RPrank	rank of the rank products of each gene in ascending order
Orirank	original ranks in each comparison, which is used to compute rank product. Not present if huge=TRUE is used.
AveFC	fold change of average expression under class 1 over that under class 2, if multiple origin, than avraged across all origin. log-fold change if data is in log scaled, original fold change if data is unlogged.
all.FC	fold change of class 1/class 2 under each origin. log-fold change if data is in log scaled

Note

Percentage of false prediction (pfp), in theory, is equivalent of false discovery rate (FDR), and it is possible to be large than 1.

The function looks for up- and down- regulated genes in two seperate steps, thus two pfps are computed and used to identify gene that belong to each group.

The function is able to replace function RP in the same library. it is a more general version, as it is able to handle data from differnt origins.

Author(s)

Fangxin Hong <fhong@salk.edu>

References

Breitling, R., Armengaud, P., Amtmann, A., and Herzyk, P.(2004) Rank Products: A simple, yet powerful, new method to detect differentially regulated genes in replicated microarray experiments, *FEBS Letter*, 57383-92

RSadvance

See Also

topGene RP plotRP RSadvance

Examples

```
# Load the data of Golub et al. (1999). data(golub)
# contains a 3051x38 gene expression
# matrix called golub, a vector of length called golub.cl
# that consists of the 38 class labels,
# and a matrix called golub.gnames whose third column
# contains the gene names.
data(golub)
##For data with single origin
subset <- c(1:4,28:30)</pre>
origin <- rep(1,7)</pre>
#identify genes
RP.out <- RPadvance(golub[,subset],golub.cl[subset],</pre>
                     origin,plot=FALSE,rand=123)
#For data from multiple origins
#Load the data arab in the package, which contains
# the expression of 22,081 genes
# of control and treatment group from the experiments
#indenpently conducted at two
#laboratories.
data(arab)
arab.origin #1 1 1 1 1 1 2 2 2 2
arab.cl #0 0 0 1 1 1 0 0 1 1
RP.adv.out <- RPadvance(arab,arab.cl,arab.origin,</pre>
              num.perm=100,gene.names=arab.gnames,logged=TRUE,rand=123)
attributes(RP.adv.out)
head(RP.adv.out$pfp)
head(RP.adv.out$RPs)
head(RP.adv.out$AveFC)
```

RSadvance

Advanced Rank Sum Analysis of Microarray

Description

Advance rank sum method to identify differentially expressed genes. It is possible to combine data from different studies, e.g. data sets generated at different laboratories.

Usage

Arguments

data	the data set that should be analyzed. Every row of this data set must correspond to a gene.
cl	a vector containing the class labels of the samples. In the two class unpaired case, the label of a sample is either 0 (e.g., control group) or 1 (e.g., case group). For one group data, the label for each sample should be 1.
origin	a vector containing the origin labels of the sample. e.g. for the data sets gen- erated at multiple laboratories, the label is the same for samples within one lab and different for samples from different labs.
num.perm	number of permutations used in the calculation of the null density. Default is 'B=100'.
logged	if "TRUE", data has bee logged, otherwise set it to "FALSE"
na.rm	if 'FALSE' (default), the NA value will not be used in computing rank. If 'TRUE', the missing values will be replaced by the genewise mean of the non- missing values. Gene will all value missing will be assigned "NA"
gene.names	if "NULL", no gene name will be attached to the estimated percentage of false prediction (pfp).
plot	If "TRUE", plot the estimated pfp verse the rank of each gene
rand	if specified, the random number generator will be put in a reproducible state.

Value

A result of identifying differentially expressed genes between two classes. The identification consists of two parts, the identification of up-regulated and down-regulated genes in class 2 compared to class 1, respectively.

pfp	estimated percentage of false positive predictions (pfp) up to the position of each gene under two identificaiton each
pval	estimated pvalue for each gene being up- and down-regulated
RSs	Origina rank-sum (average rank) of each genes
RSrank	rank of the rank sum of each gene in ascending order
Orirank	original ranks in each comparison, which is used to compute rank sum
AveFC	fold change of average expression under class 1 over that under class 2, if multiple origin, than avraged across all origin. log-fold change if data is in log scaled, original fold change if data is unlogged.
all.FC	fold change of class 1/class 2 under each origin. log-fold change if data is in log scaled

RSadvance

Note

Percentage of false prediction (pfp), in theory, is equivalent of false discovery rate (FDR), and it is possible to be large than 1.

The function looks for up- and down- regulated genes in two seperate steps, thus two pfps are computed and used to identify gene that belong to each group.

The function is able to deal with single or multiple-orgin studies. It is similar to function RP.advance expect a rank sum is computed instead of rank product. This method is more sensitive to individual rank values, while rank product is more robust to outliers (refer RankProd vignette for details)

Author(s)

Fangxin Hong <fhong@salk.edu>

See Also

topGene RP plotRP RPadvance

Examples

```
#Suppose we want to check the consistence of the data
      #sets generated in two different
      #labs. For example, we would look for genes that were \
      # measured to be up-regulated in
      #class 2 at lab 1, but down-regulated in class 2 at lab 2.\
      data(arab)
      arab.cl2 <- arab.cl
      arab.cl2[arab.cl==0 &arab.origin==2] <- 1</pre>
      arab.cl2[arab.cl==1 &arab.origin==2] <- 0</pre>
      arab.cl2
 ##[1] 0 0 0 1 1 1 1 1 0 0
      #look for genes differentially expressed
      #between hypothetical class 1 and 2
      arab.sub=arab[1:500,] ##using subset for fast computation
      arab.gnames.sub=arab.gnames[1:500]
      Rsum.adv.out <- RSadvance(arab.sub,arab.cl2,arab.origin,</pre>
                          num.perm=100,
logged=TRUE,
                          gene.names=arab.gnames.sub,rand=123)
```

attributes(Rsum.adv.out)

topGene

Description

Identify differentially expressed genes using rank product method

Usage

topGene(x,cutoff=NULL,method="pfp",num.gene=NULL,logged=TRUE,logbase=2,gene.names=NULL)

Arguments

х	the value returned by the function RP, RP.advance or Rsum.advance
cutoff	threshold in pfp used to select genes
method	If cutoff is provided, the method needs to be selected to identify genes."pfp" uses percentage of false prediction, which is the default setting. "pval" used p-value which is less stringent than pfp
num.gene	number of candidate genes of interests, if cutoff is provided, this will be ignored
logged	if "TRUE", data has bee logged, otherwise set it to "FALSE"
logbase	base used when taking log, used to restore the fold change.The default value is 2, this will be ignored if logged=FALSE
gene.names	if "NULL", no gene name will be attached to the output table

Value

Two tables of identified genes with gene.index: index of gene in the original data set RP/Rsum: Computed rank product/sum for each gene FC:(class1/class2): Expression Fold change of class 1/ class 2. pfp: estimated pfp for each gene if the gene is used as cutoff point P.value: estimated p-value for each gene

Table 1 list genes that are up-regulated under class 2, Table 1 ist genes that are down-regulated under class 2,

Author(s)

Fangxin Hong <fhong@salk.edu>

References

Breitling, R., Armengaud, P., Amtmann, A., and Herzyk, P.(2004) Rank Products: A simple, yet powerful, new method to detect differentially regulated genes in replicated microarray experiments, *FEBS Letter*, 57383-92

topGene

See Also

plotRP RP RPadvance RSadvance

Examples

```
# Load the data of Golub et al. (1999). data(golub)
# contains a 3051x38 gene expression
# matrix called golub, a vector of length called golub.cl
# that consists of the 38 class labels,
# and a matrix called golub.gnames whose third column
# contains the gene names.
data(golub)
#use a subset of data as example, apply the rank
#product method
subset <- c(1:4,28:30)</pre>
#Setting rand=123, to make the results reproducible,
#identify genes
RP.out <- RP(golub[,subset],golub.cl[subset],rand=123)</pre>
#get two lists of differentially expressed genes
#by setting FDR (false discivery rate) =0.05
table=topGene(RP.out,cutoff=0.05,method="pfp",logged=TRUE,logbase=2,
             gene.names=golub.gnames[,3])
table$Table1
table$Table2
#using pvalue<0.05</pre>
topGene(RP.out,cutoff=0.05,method="pval",logged=TRUE,logbase=2,
             gene.names=golub.gnames[,3])
#by selecting top 10 genes
topGene(RP.out,num.gene=10,gene.names=golub.gnames[,3])
```

Index

*Topic datasets arab, 2 golub, 2 lymphoma, 3 *Topic htest plotGene, 4 plotRP, 6 RP, 7 RPadvance, 9 RSadvance, 11 topGene, 14 arab, 2 exprSet, 3 golub, 2 lym.exp(lymphoma), 3 lymphoma, 3 plotGene, 4 plotRP, 6, 8, 11, 13, 15 RP, 5, 6, 7, 11, 13, 15 RPadvance, 5, 6, 8, 9, 13, 15

topGene, 5, 6, 8, 11, 13, 14

RSadvance, 5, 6, 11, 11, 15