# Package 'RBM'

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Author Dongmei Li and Chin-Yuan Liang	
Maintainer Dongmei Li <dongmei_li@urmc.rochester.edu></dongmei_li@urmc.rochester.edu>	
<b>Depends</b> R (>= 3.2.0), limma, marray	
<b>Description</b> Use A Resampling-Based Empirical Bayes Approach to Assess Differential Expression in Two-Color Microarrays and RNA-Seq data sets.	
License GPL (>= 2)	
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RBM-package

RBM:a package for microarray and RNA-Seq data analysis

# **Description**

Use A Resampling-Based Empirical Bayes Approach to Assesse Differential Expression or Identifying differntially methylated loci in Two-Color Microarrays and RNA-Seq data sets. Significant features selected through RBM\_T or RBM\_F functions could be further used as input for pathway analysis or experimental vilidations.

#### **Details**

Package: RBM
Type: Package
Version: 0.99.0
Date: 2014-10-05

Depends: R (>= 3.0.0), limma, marray

License: GPL (>= 2)

#### Author(s)

Dongmei Li and Chin-Yuan Liang Maintainer: Dongmei Li <dongmeiliur@gmail.com> and Chin-Yuan Liang liang.tony@gmail.com>

#### References

Li D, Le Pape MA, Parikh NI, Chen WX, Dye TD (2013) Assessing Differential Expression in Two-Color Microarrays: A Resampling-Based Empirical Bayes Approach. PLoS ONE 8(11): e80099. doi: 10.1371/journal.pone.0080099

#### See Also

The RBM\_T and RBM\_F functions defined in this package. The limma and marray packages.

# **Examples**

```
normal_data <- matrix(rnorm(200*6), 200, 6)
mydesign <- c(0,0,0,1,1,1)
norm_result <- RBM_T(normal_data,mydesign,50,0.05)
unif_data <- matrix(runif(200*7, 0.10, 0.95), 200, 7)
mydesign2 <- c(0,0,0, 1,1,1,1)
unif_result <- RBM_T(unif_data,mydesign2,100,0.05)</pre>
```

```
normdata_F <- matrix(rnorm(200*9, 0, 2), 200, 9)
mydesign_F <- c(0, 0, 0, 1, 1, 1, 2, 2, 2)
aContrast <- c("X1-X0", "X2-X1", "X2-X0")
normresult_F <- RBM_F(normdata_F, mydesign_F, aContrast, 100, 0.05)
unifdata_F <- matrix(runif(200*18, 0.15, 0.98), 200, 18)
mydesign2_F <- c(rep(0, 6), rep(1, 6), rep(2, 6))
aContrast <- c("X1-X0", "X2-X1", "X2-X0")
unifresult_F <- RBM_F(unifdata_F, mydesign2_F, aContrast, 100, 0.05)</pre>
```

ovarian\_cancer\_methylation

ovarian cancer methylation example from United Kingdom Ovarian Cancer Population Study (UKOPS)

## **Description**

This data set contains DNA methylation level from 1000 DNA methylation loci in 8 randomly selected women with 4 ovarian cancer cases (pre-treatment) and 4 age-matched healthy controls.

# Usage

```
ovarian_cancer_methylation
```

# Format

A matrix containing 1000 rows and 8 columns with each row denoting a methyaltion locus and each column denoting a subject.

# Value

The ovarian cancer methylation example data set contains the following information:

IlmnID Name of DNA methylation loci

case Ovarian cancer patients

control Healthy controls

# **Source**

NCBI GEO website with access number GSE19711

#### References

Teschendorff AE, Menon U, Gentry-Maharaj A, Ramus SJ et al. Age-dependent DNA methylation of genes that are suppressed in stem cells is a hallmark of cancer. Genome Res 2010 Apr;20(4):440-6. PMID: 20219944

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RBM_F	RBM_F: a R function for microarray and RNA-Seq data analysis for designs with more than two groups

# Description

Use A Resampling-Based Empirical Bayes Approach to Assess Differential Expression in Two-Color Microarrays and RNA-Seq data sets for designs with more than two groups.

# Usage

```
RBM_F(aData, vec_trt, aContrast, repetition, alpha)
```

# Arguments

aData	The input data set with rows and columns denoting features and samples, respectively
vec_trt	A vector for group notation such as 1s denote treatment group and 0s denote control group
aContrast	A vector for contrast. For example: if we want to compare group 1 with group 0, group 2 with group 1, and group 2 with group 0, then the contrast vector will be ("X1-X0", "X2"-"X1", "X2-X0")
repetition	The number of resamplings used in the analysis. You could use 1000 or higher number
alpha	The signifiance level

# **Details**

Combine resampling with empirical Bayes approach for Microarrays and RNA-Seq data analysis.

# Value

RBM\_F produces a named list with the following components:

ordfit_t	orignal t statistics
ordfit_pvalue	original p-values from lmFit and eBayes
ordfit_beta0	estimated mean for the control group
ordfit_beta1	estimated mean difference between treatment and control group
permutation_p	calculated p-values from permutation method based on resampled test statistics
bootstrap_p	calculated p-values from bootstrap method based on resampled test statistics

# Author(s)

Dongmei Li and Chin-Yuan Liang

RBM\_T

## References

Li D, Le Pape MA, Parikh NI, Chen WX, Dye TD (2013) Assessing Differential Expression in Two-Color Microarrays: A Resampling-Based Empirical Bayes Approach. PLoS ONE 8(11): e80099. doi: 10.1371/journal.pone.0080099

#### See Also

The RBM\_T function defined in this package. The limma and marray packages.

# Examples

```
normdata_F <- matrix(rnorm(200*9, 0, 2), 200, 9)
mydesign_new <- c(0, 0, 0, 1, 1, 1, 2, 2, 2)
aContrast <- c("X1-X0", "X2-X1", "X2-X0")
normresult_F <- RBM_F(normdata_F, mydesign_new, aContrast, 100, 0.05)
unifdata_F <- matrix(runif(200*18, 0.15, 0.98), 200, 18)
mydesign2_new <- c(rep(0, 6), rep(1, 6), rep(2, 6))
aContrast <- c("X1-X0", "X2-X1", "X2-X0")
unifresult_F <- RBM_F(unifdata_F, mydesign2_new, aContrast, 100, 0.05)</pre>
```

RBM\_T

RBM\_T: a R function for microarray and RNA-Seq data analysis for two-group comparisons

# **Description**

Use A Resampling-Based Empirical Bayes Approach to Assess Differential Expression or Identify differntially methylated loci in Two-Color Microarrays and RNA-Seq data sets.

# Usage

```
RBM_T(aData, vec_trt, repetition, alpha)
```

## **Arguments**

aData	The input data set with rows and columns denoting features and samples, respectively
vec_trt	A vector for group notation such as 1s denote treatment group and 0s denote control group
repetition	The number of resamplings used in the analysis. You could use 1000 or higher number
alpha	The signifiance level

# **Details**

Combine resampling with empirical Bayes approach for Microarrays and RNA-Seq data analysis.

RBM\_T

## Value

RBM\_T produces a named list with the following components:

ordfit\_t original t statistics

ordfit\_pvalue original p-values from lmFit and eBayes

ordfit\_beta0 estimated mean for the control group

ordfit\_beta1 estimated mean difference between treatment and control group

permutation\_p calculated p-values from permutation method based on resampled test statistics

bootstrap\_p calculated p-values from bootstrap method based on resampled test statistics

## Author(s)

Dongmei Li and Chin-Yuan Liang

#### References

Li D, Le Pape MA, Parikh NI, Chen WX, Dye TD (2013) Assessing Differential Expression in Two-Color Microarrays: A Resampling-Based Empirical Bayes Approach. PLoS ONE 8(11): e80099. doi: 10.1371/journal.pone.0080099

#### See Also

The RBM\_F function defined in this package. The limma and marray packages.

#### **Examples**

```
normal_data <- matrix(rnorm(200*6), 200, 6)
mydesign <- c(0,0,0,1,1,1)
norm_result <- RBM_T(normal_data,mydesign,50,0.05)
unif_data <- matrix(runif(200*7, 0.10, 0.95), 200, 7)
mydesign2 <- c(0,0,0, 1,1,1,1)
unif_result <- RBM_T(unif_data,mydesign2,100,0.05)</pre>
```

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