# Package 'meaca'

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<b>Description</b> This package documents the functions used when performing MEACA gene set enrichment analysis.
<b>Depends</b> R (>= $3.2.1$ )
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btw_gene_corr estimate_sigma meaca_multiple meaca_single simulate_expression_data standardize_expression_data transform_count_edgeR transform_count_vst
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btw\_gene\_corr

Calculate sample (Pearson) correlations among gene clusters

#### **Description**

Calculate sample (Pearson) correlations among gene clusters

#### Usage

```
btw_gene_corr(expression_data, trt, go_term, standardize = T)
```

# **Arguments**

expression\_data

the expressoin matrix.

trt treatment indicators, 1 for treatment, 0 for control group go\_term an indicator vector. 1 for genes in the test set, 0 otherwise

standardize whether the data should be standaridzed

#### Value

a \$1 \times 3\$ data frame containing values for \$\rho\_1\$, \$\rho\_3\$ and \$\rho\_2\$ respectively.

testSetCor Average correlation for genes in the test set

interCor Average correlation between genes in the test set and those not in the test set

backSetCor Average correlations for genes not in the test set.

estimate\_sigma

Estimate sample covariance.

#### **Description**

Estimate sample covariance and calculate the gene-level statistics

# Usage

```
estimate_sigma(expression_data, trt)
```

#### Arguments

expression\_data

the expression matrix.

trt sample labels. 0 for control and 1 for treatment

# Value

a list

sigma a covariance matrix

t\_val a vector of gene level test statistics

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meaca\_multiple

meaca-multiple.

#### **Description**

meaca for testing multiple gene sets.

# Usage

```
meaca_multiple(
  expression_data,
  trt,
  geneset,
  standardize = T,
  min_set_size = 5,
  fdr_method = "BH"
)
```

#### **Arguments**

expression\_data

the expressoin matrix.

trt treatment labels.

geneset gene sets to be tested, an object from read\_gene\_set.

standardize whether the data should be standaridzed.

min\_set\_size the minimum number of genes contained for a gene set to be considered.

fdr\_method which method is ued to adjust the p values. see arguments in function p. adjust.

#### Value

a data frame

meaca\_single

meaca-single.

#### **Description**

meaca for single gene set test.

#### Usage

```
meaca_single(expression_data, trt, go_term, standardize = F)
```

# Arguments

```
expression_data
```

the expressoin matrix.

trt treatment indicators, 1 for treatment, 0 for control group.
go\_term an indicator vector. 1 for genes in the test set, 0 otherwise.

standardize whether the data should be standaridzed.

#### Value

```
a list

stat the test statistic

p1 chi-square test p value

status "up" or "down", the direction of differential expression

p2 two-sided test p-value using normal distribution
```

# **Examples**

simulate\_expression\_data

Simulate expression data.

#### **Description**

simulate normally distributed expression data with desired DE probabilities for genes in the test set and for those not in the test set..

# Usage

```
simulate_expression_data(
    size,
    n_gene,
    n_test,
    prop,
    de_mu,
    de_sd,
    rho1,
    rho2,
    rho3,
    data_gen_method = "chol",
    seed = 123
)
```

#### **Arguments**

```
number of samples to be simulated
n_gene total number of genes to be simulated
n_test number of genes in the test set.

prop a vector of length 2, proportion of DE genes within go term and outside go_term, corresponding to $p_t$ and $p_b$.

de_mu, de_sd if the gene is DE, delta ~ N(de_mu, de_sd)
```

rho1 a scalar, correlation between two test genes (i.e.,  $\rho_1$  in the paper)

rho2 a scalar, correlation between two background genes (i.e.,  $\rho_2$  in the paper) rho3 correlation between a test gene and a background gene (i.e.,  $\rho_3$  in the paper)

data\_gen\_method

data generation method; if 'data\_gen\_method = MASS', then mvrnorm is used,

otherwise see function rmvnorm

seed the seed used for simulation (for reproducibility purpose)

#### Value

a list

data a expression matrix of  $m \times n$  where m is the number of genes and n is the

number of samples.

trt sample labels of length n, 1 for treatment and 0 for control.
go\_term gene labels of length m, 1 for go\_term genes and 0 otherwise.

sigma true covariance matrix upon which data is simulated.

#### **Examples**

```
t1 <- simulate_expression_data(size = 50, n_gene = 500, n_test = 100, prop = c(0.1, 0.1), de_mu = 2, de_sd = 1, rho1 = 0.1, rho2 = 0.05, rho3 = -0.05, data_gen_method = "cho1", seed = 123)
```

standardize\_expression\_data

standardize expression data, with method described in the paper.

#### **Description**

Standardize the expression data.

#### Usage

```
standardize_expression_data(expression_data, trt)
```

#### **Arguments**

expression\_data

the expression matrix.

trt sample labels. 0 for control and 1 for treatment

#### Value

a matrix of the same dimension as input data.

#### **Description**

Perform count matrix transformation using edgeR procedure

# Usage

```
transform_count_edgeR(y, group)
```

# Arguments

```
y the expression count matrix, columns being samples, rows being genes group a vector of treatment label (e.g., 0 for control, 1 for treatment).
```

#### Value

a matrix of transformed data

#### See Also

```
camera.DGEList
```

# **Examples**

```
mu <- matrix(10, 100, 4)
group <- factor(c(0,0,1,1))
design <- model.matrix(~group)
set.seed(123)
library(edgeR)
y0 <- matrix(rnbinom(100*4, mu=mu, size=10),100,4)
y <- DGEList(counts=y0, group=group)
y <- estimateDisp(y, design)

iset1 <- 1:10
camera.DGEList(y, iset1, design)

# the Pvalue should be the same
y2 <- transform_count_edgeR(y = y0, group = group)
camera(y2, iset1, design)</pre>
```

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transform\_count\_vst

Perform variance stabilizing transformation for the count matrix data

#### **Description**

Perform variance stabilizing transformation for the count matrix data

#### Usage

```
transform_count_vst(y, group, ...)
```

#### **Arguments**

```
y the expression count matrix, columns being samples, rows being genes group a vector of treatment label (e.g., 0 for control, 1 for treatment).
... other parameters used in varianceStabilizingTransformation
```

#### **Details**

It is absolutely critical that the columns of the count matrix and the rows of the column data (information about samples) are in the same order. For more details, see <a href="https://bioconductor.org/packages/release/bioc/vignettes/DESeq2/inst/doc/DESeq2.html">https://bioconductor.org/packages/release/bioc/vignettes/DESeq2/inst/doc/DESeq2.html</a>

#### Value

a matrix of transformed data

#### **Examples**

```
y \leftarrow matrix(rbinom(6000, 20, 0.4), nrow = 1000)
group <- c(0, 0, 0, 1, 1, 1)
yr \leftarrow transform\_count\_vst(y = y, group = group)
```

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