# Package 'NormalyzerDE'

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<b>Title</b> Evaluation of normalization methods and calculation of differential expression analysis statistics
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Maintainer Jakob Willforss < jakob.willforss@immun.lth.se>
Author Jakob Willforss
Description NormalyzerDE provides screening of normalization methods for LC-MS based expression data. It calculates a range of normalized matrices using both existing approaches and a novel time-segmented approach, calculates performance measures and generates an evaluation report. Furthermore, it provides an easy utility for Limma- or ANOVA- based differential expression analysis.
Imports vsn, preprocessCore, limma, MASS, ape, car, ggplot2, methods, Biobase, RcmdrMisc, raster, utils, stats, SummarizedExperiment, matrixStats
Suggests knitr, testthat, rmarkdown, roxygen2, hexbin
VignetteBuilder knitr
<b>biocViews</b> Normalization, MultipleComparison, Visualization, Bayesian, Proteomics, Metabolomics, DifferentialExpression
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analyzeNormalizations Calculate measures for normalization results

### Description

This function prepares an NormalyzerEvaluationResults object containing the evaluation measures CV (coefficient of variance), MAD (median absolute deviation), average variance, significance measures (ANOVA between condition groups) and correlation between replicates.

#### Usage

```
analyzeNormalizations(nr, categoricalAnova = FALSE)
```

### **Arguments**

 $\begin{tabular}{ll} nr & Normalyzer\ results\ object\ with\ calculated\ results. \\ categorical Anova \end{tabular}$ 

Whether categorical or numerical (ordered) ANOVA should be calculated.

### Value

Normalyzer results with attached evaluation results object.

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#### **Examples**

```
data(example_summarized_experiment)
normObj <- getVerifiedNormalyzerObject("job_name", example_summarized_experiment)
normResults <- normMethods(normObj)
normResultsWithEval <- analyzeNormalizations(normResults)</pre>
```

calculateContrasts

Performs statistical comparisons between the supplied conditions. It uses the design matrix and data matrix in the supplied Normalyzer-Statistics object. A column is supplied specifying which of the columns in the design matrix that is used for deciding the sample groups. The comparisons vector specifies which pairwise comparisons between condition levels that are to be calculated.

### Description

Optionally, a batch column can be specified allowing compensation for covariate variation in the statistical model. This is only compatible with a Limma-based statistical analysis.

#### Usage

```
calculateContrasts(nst, comparisons, condCol, batchCol = NULL,
   splitter = "-", type = "limma")

## S4 method for signature 'NormalyzerStatistics'
calculateContrasts(nst, comparisons,
   condCol, batchCol = NULL, splitter = "-", type = "limma")
```

#### **Arguments**

nst	Results evaluation object.
comparisons	String with comparisons for contrasts.
condCol	Column name in design matrix containing condition information.
batchCol	Column name in design matrix containing batch information.
splitter	Character dividing contrast conditions.
type	Type of statistical test (Limma or welch).

### Value

nst Statistics object with statistical measures calculated

```
data(example_stat_summarized_experiment)
nst <- NormalyzerStatistics(example_stat_summarized_experiment)
results <- calculateContrasts(nst, c("1-2", "2-3"), "group")
resultsBatch <- calculateContrasts(nst, c("1-2", "2-3"), "group", "batch")</pre>
```

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```
generateAnnotatedMatrix
```

Generate an annotated data frame from statistics object

#### **Description**

Extracts key values (p-value, adjusted p-value, log2-fold change and average expression values) from an NormalyzerStatistics instance and appends these to the annotation- and data-matrices

### Usage

```
generateAnnotatedMatrix(nst)
```

#### **Arguments**

nst

NormalyzerDE statistics object.

#### Value

outDf Annotated statistics matrix

### Examples

```
data(example_stat_summarized_experiment)
statObj <- NormalyzerStatistics(example_stat_summarized_experiment)
statObj <- calculateContrasts(statObj, comparisons=c("1-2", "2-3"), condCol="group", type="limma")
annotDf <- generateAnnotatedMatrix(statObj)</pre>
```

generatePlots

Generates a number of visualizations for the performance measures calculated for the normalized matrices. These contain both general measures and direct comparisons for different normalization approaches.

### Description

They include:

#### Usage

```
generatePlots(nr, jobdir, plotRows = 3, plotCols = 4)
```

#### **Arguments**

nr Normalyzer results object.
jobdir Path to output directory for run.

plotRows Number of plot rows.
plotCols Number of plot columns.

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#### **Details**

"Total intensity" Barplot showing the summed intensity in each sample for thelog2-transformed data

"Total missing" Barplot showing the number of missing values found in each sample for the log2-tranformed data

Log2-MDS plot: MDS plot where data is reduced to two dimensions allowing inspection of the main global changes in the data

PCV - Intragroup: Mean of intragroup CV of all replicate groups

PMAD - Intragroup: Mean of intragroup median absolute deviation across replicate groups

PEV - Intragroup: Mean of intragroup pooled estimate of variance across the replicate groups

Relative PCV, PMAD and PEV compared to log2: The results from PCV, PMAD and PEV from all normalized data compared to the log2 data

Stable variables plot: 5 analysis of log2 transformed data. Thereafter, global CV of these variables is estimated from different normalized datasets. A plot of global CV of the stable variables from all datasets on the y-axis and PCV-compared to log2 on the x-axis is generated.

CV vs Raw Intensity plots: For the first replicate group in each of the normalized dataset, a plot of PCV of each variable compared to the average intensity of the variable in the replicate group is plotted.

MA plots: Plotted using the plotMA function of the limma package. The first sample in each dataset is plotted against the average of the replicate group that sample belong to.

Scatterplots: The first two samples from each dataset are plotted.

Q-Q plots: QQ-plots are plotted for the first sample in each normalized dataset.

Boxplots: Boxplots for all samples are plotted and colored according to the replicate grouping.

Relative Log Expression (RLE) plots: Relative log expression value plots. Ratio between the expression of the variable and the median expression of this variable across all samples. The samples should be aligned around zero. Any deviation would indicate discrepancies in the data.

Density plots: Density distributions for each sample using the density function. Can capture outliers (if single densities lies far from the others) and see if there is batch effects in the dataset (if for instance there is two clear collections of lines in the data).

MDS plots Multidimensional scaling plot using the cmdscale() function from the stats package. Is often able to show whether replicates group together, and whether there are any clear outliers in the data.

MeanSDplots Displays the standard deviation values against values ordered according to mean. If no dependency on mean is present (as is desired) a flat red line is shown.

Pearson and Spearman correlation Mean of intragroup Pearson and Spearman correlation values for each method.

Dendograms Generated using the hclust function. Data is centered and scaled prior to analysis. Coloring of replicates is done using as.phylo from the ape package.

P-value histograms Histogram plots of p-values after calculating an ANOVA between different condition groups. If no effect is present in the data a flat distribution is expected. If an effect is present a flat distribution is still expected, but with a sharp peak close to zero. If other effects are present it might indicate that the data doesn't support the assumptions of ANOVA, for instance if there are batch effects present in the data.

#### Value

None

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#### **Examples**

```
data(example_summarized_experiment)
normObj <- getVerifiedNormalyzerObject("job_name", example_summarized_experiment)
normResults <- normMethods(normObj)
normResultsWithEval <- analyzeNormalizations(normResults)
outputDir <- tempdir()
generatePlots(normResultsWithEval, outputDir)</pre>
```

generateStatsReport

Generate full output report plot document. Plots p-value histograms for each contrast in the NormalyzerStatistics instance and writes these to a PDF report.

#### **Description**

Generate full output report plot document. Plots p-value histograms for each contrast in the NormalyzerStatistics instance and writes these to a PDF report.

#### Usage

```
generateStatsReport(nst, jobName, jobDir, plotRows = 3, plotCols = 4)
```

#### **Arguments**

nst NormalyzerDE statistics object.

jobName Name of processing run.

jobDir Path to output directory.

plotRows Number of plot rows.

plotCols Number of plot columns.

#### Value

None

```
data(example_stat_summarized_experiment)
statObj <- NormalyzerStatistics(example_stat_summarized_experiment)
statObj <- calculateContrasts(statObj, comparisons=c("1-2", "2-3"),
    condCol="group", type="limma")
outputDir <- tempdir()
generateStatsReport(statObj, "jobName", outputDir)</pre>
```

getRTNormalizedMatrix

getRTNormalizedMatrix Perform RT-segmented normalization by performing the supplied normalization over retention-time sliced data

#### **Description**

The function orders the retention times and steps through them using the supplied step size (in minutes). If smaller than a fixed lower boundary the window is expanded to ensure a minimum amount of data in each normalization step. An offset can be specified which can be used to perform multiple RT-segmentations with partial overlapping windows.

### Usage

```
getRTNormalizedMatrix(rawMatrix, retentionTimes, normMethod,
  stepSizeMinutes = 1, windowMinCount = 100, offset = 0)
```

#### **Arguments**

rawMatrix Target matrix to be normalized

retentionTimes Vector of retention times corresponding to rawMatrix

normMethod The normalization method to apply to the time windows

stepSizeMinutes

Size of windows to be normalized

windowMinCount Minimum number of values for window to not be expanded.

offset Whether time window should shifted half step size

### Value

Normalized matrix

```
data(example_data_small)
data(example_design_small)
data(example_data_only_values)
dataMat <- example_data_only_values
retentionTimes <- as.numeric(example_data[, "Average.RT"])
performCyclicLoessNormalization <- function(rawMatrix) {
    log2Matrix <- log2(rawMatrix)
    normMatrix <- limma::normalizeCyclicLoess(log2Matrix, method="fast")
    colnames(normMatrix) <- colnames(rawMatrix)
    normMatrix
}
rtNormMat <- getRTNormalizedMatrix(dataMat, retentionTimes,
performCyclicLoessNormalization, stepSizeMinutes=1, windowMinCount=100)</pre>
```

```
getSmoothedRTNormalizedMatrix
```

Generate multiple RT time-window normalized matrices where one is shifted. Merge them using a specified method (mean or median) and return the result.

#### **Description**

Uses the function getRTNormalizedMatrix to generate multiple normalized matrices which are shifted respective to each other and finally merged into a single matrix. This could potentially reduce effect of fluctuations within individual windows.

#### Usage

```
getSmoothedRTNormalizedMatrix(rawMatrix, retentionTimes, normMethod,
   stepSizeMinutes, windowShifts = 2, windowMinCount = 100,
   mergeMethod = "mean")
```

### **Arguments**

```
rawMatrix Target matrix to be normalized
retentionTimes Vector of retention times corresponding to rawMatrix
normMethod The normalization method to apply to the time windows
stepSizeMinutes
Size of windows to be normalized
windowShifts Number of frame shifts.
windowMinCount Minimum number of features within window.
mergeMethod Layer merging approach. Mean or median.
```

#### Value

Normalized matrix

```
data(example_data_small)
data(example_data_only_values)
data(example_design_small)
retentionTimes <- as.numeric(example_data[, "Average.RT"])
dataMat <- example_data_only_values
performCyclicLoessNormalization <- function(rawMatrix) {
    log2Matrix <- log2(rawMatrix)
    normMatrix <- limma::normalizeCyclicLoess(log2Matrix, method="fast")
    colnames(normMatrix) <- colnames(rawMatrix)
    normMatrix
}
rtNormMat <- getSmoothedRTNormalizedMatrix(dataMat, retentionTimes,
    performCyclicLoessNormalization, stepSizeMinutes=1, windowMinCount=100,
    windowShifts=2, mergeMethod="median")</pre>
```

```
getVerifiedNormalyzerObject
```

Verify that input data is in correct format, and if so, return a generated NormalyzerDE data object from that input data

#### **Description**

This function performs a number of checks on the input data and provides informative error messages if the data isn't fulfilling the required format. Checks include verifying that the design matrix matches to the data matrix, that the data matrix contains valid numbers and that samples have enough values for analysis

### Usage

```
getVerifiedNormalyzerObject(jobName, summarizedExp, threshold = 15,
  omitSamples = FALSE, requireReplicates = TRUE, quiet = FALSE)
```

#### **Arguments**

jobName Name of ongoing run.

summarizedExp Summarized experiment input object

threshold Minimum number of features.

omitSamples Automatically omit invalid samples from analysis.

requireReplicates

Require there to be at least to samples per condition

quiet Don't print output messages during processing

#### Value

Normalyzer data object representing verified input data.

### **Examples**

```
data(example_summarized_experiment)
normObj <- getVerifiedNormalyzerObject("job_name", example_summarized_experiment)</pre>
```

```
globalIntensityNormalization
```

The normalization divides the intensity of each variable in a sample with the sum of intensities of all variables in the sample and multiplies with the median of sum of intensities of all variables in all samples. The normalized data is then log2-transformed.

### **Description**

The normalization divides the intensity of each variable in a sample with the sum of intensities of all variables in the sample and multiplies with the median of sum of intensities of all variables in all samples. The normalized data is then log2-transformed.

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#### Usage

```
globalIntensityNormalization(rawMatrix)
```

#### **Arguments**

rawMatrix Target matrix to be normalized

#### Value

Normalized and log-transformed matrix

### **Examples**

```
data(example_data_only_values_small)
normMatrix <- globalIntensityNormalization(example_data_only_values)</pre>
```

loadData

Load raw data into dataframe

### Description

General function which allows specifying different types of input data including "proteios", "maxquant-pep" (peptide output from MaxQuant) and "maxquantprot" (protein output from MaxQuant) formats.

### Usage

```
loadData(dataPath, inputFormat = "default", zeroToNA = FALSE)
```

#### **Arguments**

dataPath File path to design matrix.

inputFormat If input is given in standard NormalyzerDE format, Proteios format or in MaxQuant

protein or peptide format

zeroToNA Automatically convert zeroes to NA values

#### Value

rawData Raw data loaded into data frame

```
## Not run:
df <- loadData("data.tsv")
## End(Not run)</pre>
```

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loadDesign	Load raw design into dataframe

#### **Description**

Takes a design path, loads the matrix and ensures that the sample column is in character format and that the group column is in factor format.

#### Usage

```
loadDesign(designPath, sampleCol = "sample", groupCol = "group")
```

#### **Arguments**

designPath File path to design matrix.

sampleCol Column name for column containing sample names.
groupCol Column name for column containing condition levels.

### Value

designMatrix Design data loaded into data frame

#### **Examples**

```
## Not run:
df <- loadDesign("design.tsv")
## End(Not run)</pre>
```

meanNormalization

Intensity of each variable in a given sample is divided by the mean of sum of intensities of all variables in the sample and then multiplied by the mean of sum of intensities of all variables in all samples. The normalized data is then transformed to log2.

#### **Description**

Intensity of each variable in a given sample is divided by the mean of sum of intensities of all variables in the sample and then multiplied by the mean of sum of intensities of all variables in all samples. The normalized data is then transformed to log2.

### Usage

```
meanNormalization(rawMatrix)
```

#### **Arguments**

rawMatrix Target matrix to be normalized

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#### Value

Normalized and log-transformed matrix

### **Examples**

```
data(example_data_only_values_small)
normMatrix <- meanNormalization(example_data_only_values)</pre>
```

medianNormalization

Intensity of each variable in a given sample is divided by the median of intensities of all variables in the sample and then multiplied by the mean of median of sum of intensities of all variables in all samples. The normalized data is then log2-transformed.

#### **Description**

Intensity of each variable in a given sample is divided by the median of intensities of all variables in the sample and then multiplied by the mean of median of sum of intensities of all variables in all samples. The normalized data is then log2-transformed.

### Usage

```
medianNormalization(rawMatrix)
```

### **Arguments**

rawMatrix Target matrix to be normalized

### Value

Normalized and log-transformed matrix

#### **Examples**

```
data(example_data_only_values_small)
normMatrix <- medianNormalization(example_data_only_values)</pre>
```

normalyzer

NormalyzerDE pipeline entry point

### Description

This function is the main execution point for the normalization part of the NormalyzerDE analysis pipeline. When executed it performs the following steps:

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#### **Usage**

```
normalyzer(jobName, designPath = NULL, dataPath = NULL,
   experimentObj = NULL, outputDir = ".", forceAllMethods = FALSE,
   omitLowAbundSamples = FALSE, sampleAbundThres = 15,
   requireReplicates = TRUE, normalizeRetentionTime = TRUE,
   plotRows = 3, plotCols = 4, zeroToNA = FALSE,
   sampleColName = "sample", groupColName = "group",
   inputFormat = "default", skipAnalysis = FALSE, quiet = FALSE,
   rtStepSizeMinutes = 1, rtWindowMinCount = 100, rtWindowShifts = 1,
   rtWindowMergeMethod = "mean")
```

#### **Arguments**

jobName Give the current run a name.

designPath Path to file containing design matrix.

dataPath Specify an output directory for generated files. Defaults to current working

directory.

experimentObj SummarizedExperiment object, can be provided as input as alternative to 'de-

signPath' and 'dataPath'

outputDir Directory where results folder is created.

forceAllMethods

Debugging function. Run all normalizations even if they aren't in the recom-

mended range of number of values

omitLowAbundSamples

Automatically remove samples with fewer non-NA values compared to threshold given by sampleAbundThres. Will otherwise stop with error message if such sample is encountered.

sampleAbundThres

Threshold for omitting low-abundant samples. Is by default set to 15.

requireReplicates

Require multiple samples per condition to pass input validation.

normalizeRetentionTime

Perform normalizations over retention time.

plotRows Number of plot-rows in output documentation.

plotCols Number of plot-columns in output documentation.

zeroToNA Convert zero values to NA.

sampleColName Column name in design matrix containing sample IDs. groupColName Column name in design matrix containing condition IDs.

inputFormat Type of input format.

skipAnalysis Only perform normalization steps.

quiet Omit status messages printed during run

rtStepSizeMinutes

Retention time normalization window size.

rtWindowMinCount

Minimum number of datapoints in each retention-time segment.

rtWindowShifts Number of layered retention time normalized windows.

rtWindowMergeMethod

Merge approach for layered retention time windows.

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#### **Details**

1: Loads the data matrix containing expression values and optional annotations, as well as the design matrix containing the experimental setup 2: Performs input data verification to validate that the data is in correct format. This step captures many common formatting errors. It returns an instance of the NormalyzerDataset class representing the unprocessed data. 3: Calculate a range of normalizations for the dataset. The result is provided as a NormalyzerResults object containing the resulting data matrices from each normalization. 4: Analyze the normalizations and generate performance measures for each of the normalized datasets. This result is provided as a NormalyzerEvaluationResults object. 5: Output the matrices containing the normalized datasets to files. 6: Generate visualizations overviewing the performance measures and write them to a PDF report.

#### Value

None

#### **Examples**

```
## Not run:
data_path <- system.file(package="NormalyzerDE", "extdata", "tiny_data.tsv")</pre>
design_path <- system.file(package="NormalyzerDE", "extdata", "tiny_design.tsv")</pre>
out_dir <- tempdir()</pre>
normalyzer(
    jobName="my_jobname",
    designPath=design_path,
    dataPath=data_path,
    outputDir=out_dir)
normalyzer(
    "my_jobname",
    designMatrix="design.tsv",
    "data.tsv",
    outputDir="path/to/output",
    normalizeRetentionTime=TRUE,
    retentionTimeWindow=2)
normalyzer(
    "my_jobname",
    designMatrix="design.tsv",
    "data.tsv",
    outputDir="path/to/output",
    inputFormat="maxquantprot")
## End(Not run)
```

normalyzerDE

NormalyzerDE differential expression

### Description

Performs differential expression analysis on a normalization matrix. This command executes a pipeline processing the data and generates an annotated normalization matrix and a report containing p-value histograms for each of the performed comparisons.

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#### Usage

```
normalyzerDE(jobName, comparisons, designPath = NULL, dataPath = NULL,
   experimentObj = NULL, outputDir = ".", logTrans = FALSE,
   type = "limma", sampleCol = "sample", condCol = "group",
   batchCol = NULL, techRepCol = NULL, leastRepCount = 1,
   quiet = FALSE)
```

#### **Arguments**

jobName Name of job

comparisons Character vector containing target contrasts. If comparing condA with condB,

then the vector would be c("condA-condB")

designPath File path to design matrix
dataPath File path to normalized matrix

experimentObj SummarizedExperiment object, can be provided as input as alternative to 'de-

signPath' and 'dataPath'

outputDir Path to output directory

logTrans Log transform the input (needed if providing non-logged input)

type Type of statistical comparison, "limma" or "welch"

sampleCol Design matrix column header for column containing sample IDs

condCol Design matrix column header for column containing sample conditions

batchCol Provide an optional column for inclusion of possible batch variance in the model

techRepCol Design matrix column header for column containing technical replicates

leastRepCount Minimum required replicate count

quiet Omit status messages printed during run

### **Details**

When executed, it performs the following steps:

1: Read the data and the design matrices into dataframes. 2: Generate an instance of the NormalyzerStatistics class representing the data and their statistical comparisons. 3: Optionally reduce technical replicates in both the data matrix and the design matrix 4: Calculate statistical contrats between supplied groups 5: Generate an annotated version of the original dataframe where columns containing statistical key measures have been added 6: Write the table to file 7: Generate a PDF report displaying p-value histograms for each calculated contrast

### Value

None

```
data_path <- system.file(package="NormalyzerDE", "extdata", "tiny_data.tsv")
design_path <- system.file(package="NormalyzerDE", "extdata", "tiny_design.tsv")
out_dir <- tempdir()
normalyzerDE(
   jobName="my_jobname",
   comparisons=c("4-5"),
   designPath=design_path,</pre>
```

```
dataPath=data_path,
outputDir=out_dir,
condCol="group")
```

NormalyzerEvaluationResults

Representation of evaluation results by calculating performance measures for an an NormalyzerResults instance

#### **Description**

Contains the resulting information from the processing which subsequently can be used to generate the quality assessment report.

### Usage

```
NormalyzerEvaluationResults(nr)
NormalyzerEvaluationResults(nr)
```

#### **Arguments**

nr

NormalyzerResults object

#### Value

nds Generated NormalyzerEvaluationResults instance

#### **Slots**

avgcvmem Average coefficient of variance per method

avgcvmempdiff Percentage difference of mean coefficient of variance compared to log2-transformed data

featureCVPerMethod CV calculated per feature and normalization method.

avgmadmem Average median absolute deviation

avgmadmempdiff Percentage difference of median absolute deviation compared to log2-transformed data

avgvarmem Average variance per method

avgvarmempdiff Percentage difference of mean variance compared to log2-transformed data

lowVarFeaturesCVs List of 5 for log2-transformed data

lowVarFeaturesCVsPercDiff Coefficient of variance for least variable entries

anovaP ANOVA calculated p-values

repCorPear Within group Pearson correlations

repCorSpear Within group Spearman correlations

```
data(example_summarized_experiment)
normObj <- getVerifiedNormalyzerObject("job_name", example_summarized_experiment)
normResults <- normMethods(normObj)
normEval <- NormalyzerEvaluationResults(normResults)</pre>
```

NormalyzerResults-class

Representation of the results from performing normalization over a dataset

### **Description**

It is linked to a NormalyzerDataset instance representing the raw data which has been processed. After performing evaluation it also links to an instance of NormalyzerEvaluationResults representing the results from the evaluation.

#### **Slots**

normalizations SummarizedExperiment object containing calculated normalization results nds Normalyzer dataset representing run data

ner Normalyzer evaluation results for running extended normalizations

NormalyzerStatistics Class representing a dataset for statistical processing in NormalyzerDE

### Description

Is initialized with an annotation matrix, a data matrix and a design data frame. This object can subsequently be processed to generate statistical values and in turn used to write a full matrix with additional statistical information as well as a graphical report of the comparisons.

#### Usage

```
NormalyzerStatistics(experimentObj, conditionCol = "group",
    logTrans = FALSE, leastRepCount = 2)

NormalyzerStatistics(experimentObj, conditionCol = "group",
    logTrans = FALSE, leastRepCount = 2)
```

### **Arguments**

 ${\tt experimentObj} \quad Instance \ of \ Summarized Experiment \ containing \ matrix \ and \ design \ information$ 

as column data

conditionCol Column in column data containing the condition information for which contrasts

will be performed

logTrans Whether the input data should be log transformed

leastRepCount Least replicates in each group to be retained for contrast calculations

### Value

nds Generated NormalyzerStatistics instance

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#### **Slots**

```
annotMat Matrix containing annotation information
dataMat Matrix containing (normalized) expression data
filteredDataMat Filtered matrix with low-count rows removed
designDf Data frame containing design conditions
filteringContrast Vector showing which entries are filtered (due to low count)
pairwiseCompsP List with P-values for pairwise comparisons
pairwiseCompsFdr List with FDR-values for pairwise comparisons
pairwiseCompsAve List with average expression values
pairwiseCompsFold List with log2 fold-change values for pairwise comparisons
```

### **Examples**

```
data(example_stat_summarized_experiment)
nst <- NormalyzerStatistics(example_stat_summarized_experiment)</pre>
```

normMethods

Perform normalizations on Normalyzer dataset

#### **Description**

Perform normalizations on Normalyzer dataset

#### Usage

```
normMethods(nds, forceAll = FALSE, normalizeRetentionTime = TRUE,
  quiet = FALSE, rtStepSizeMinutes = 1, rtWindowMinCount = 100,
  rtWindowShifts = 1, rtWindowMergeMethod = "mean")
```

#### **Arguments**

nds Normalyzer dataset object.

forceAll Force all methods to run despite not qualifying for thresholds.

 ${\tt normalize} Retention {\tt Time}$ 

Perform retention time based normalization methods.

quiet Prevent diagnostic output

rtStepSizeMinutes

Retention time normalization window size.

rtWindowMinCount

Minimum number of datapoints in each retention-time segment.

rtWindowShifts Number of layered retention time normalized windows.

rtWindowMergeMethod

Merge approach for layered retention time windows.

#### Value

Returns Normalyzer results object with performed analyzes assigned as attributes

#### **Examples**

```
data(example_summarized_experiment)
normObj <- getVerifiedNormalyzerObject("job_name", example_summarized_experiment)
normResults <- normMethods(normObj)</pre>
```

 ${\tt performCyclicLoessNormalization}$ 

Cyclic Loess normalization

#### **Description**

Log2 transformed data is normalized by Loess method using the function "normalizeCyclicLoess". Further information is available for the function "normalizeCyclicLoess" in the Limma package.

#### Usage

```
performCyclicLoessNormalization(rawMatrix)
```

#### **Arguments**

rawMatrix

Target matrix to be normalized

#### Value

Normalized matrix

#### **Examples**

```
data(example_data_only_values_small)
normMatrix <- performCyclicLoessNormalization(example_data_only_values)</pre>
```

 ${\tt performGlobalRLRNormalization}$ 

Global linear regression normalization

### Description

Log2 transformed data is normalized by robust linear regression using the function "rlm" from the MASS package.

### Usage

```
{\tt performGlobalRLRNormalization} ({\tt rawMatrix})
```

#### **Arguments**

rawMatrix

Target matrix to be normalized

#### Value

Normalized matrix

#### **Examples**

```
data(example_data_only_values_small)
normMatrix <- performGlobalRLRNormalization(example_data_only_values)</pre>
```

performQuantileNormalization

Quantile normalization is performed by the function "normalize.quantiles" from the package preprocessCore.

### **Description**

It makes the assumption that the data in different samples should originate from an identical distribution. It does this by generating a reference distribution and then scaling the other samples accordingly.

### Usage

```
performQuantileNormalization(rawMatrix)
```

#### **Arguments**

rawMatrix

Target matrix to be normalized

#### Value

Normalized matrix

### **Examples**

```
data(example_data_only_values_small)
normMatrix <- performQuantileNormalization(example_data_only_values)</pre>
```

performSMADNormalization

Median absolute deviation normalization Normalization subtracts the median and divides the data by the median absolute deviation (MAD).

### Description

Median absolute deviation normalization Normalization subtracts the median and divides the data by the median absolute deviation (MAD).

#### Usage

```
performSMADNormalization(rawMatrix)
```

#### **Arguments**

rawMatrix Target matrix to be normalized

#### Value

Normalized matrix

#### **Examples**

```
data(example_data_only_values_small)
normMatrix <- performSMADNormalization(example_data_only_values)</pre>
```

performVSNNormalization

Log2 transformed data is normalized using the function "justvsn" from the VSN package.

#### **Description**

The VSN (Variance Stabilizing Normalization) attempts to transform the data in such a way that the variance remains nearly constant over the intensity spectrum

### Usage

```
performVSNNormalization(rawMatrix)
```

#### **Arguments**

rawMatrix

Target matrix to be normalized

### Value

Normalized matrix

### **Examples**

```
data(example_data_only_values_small)
normMatrix <- performVSNNormalization(example_data_only_values)</pre>
```

reduceDesignTechRep

Remove technical replicates from design matrix.

### Description

Technical replicates are specified as duplicate strings. The first sample name corresponding for each technical replicate group is retained.

### Usage

```
reduceDesignTechRep(designMat, techRepGroups)
```

### **Arguments**

designMat NormalyzerDE design matrix techRepGroups Technical replicates vector

#### Value

collDesignDf Reduced design matrix

#### **Examples**

```
designDf <- data.frame(
   sample=c("a1", "a2", "b1", "b2", "c1", "c2", "d1", "d2"),
   group=c(rep("A", 4), rep("B", 4)),
   techrep=c("a", "a", "b", "b", "c", "c", "d", "d")
)
statObj <- reduceDesignTechRep(designDf, designDf$techrep)</pre>
```

reduceTechnicalReplicates

Remove technical replicates from data matrix

### **Description**

Collapses sample values into their average. If only one value is present due to NA-values in other technical replicates, then that value is used.

#### Usage

```
reduceTechnicalReplicates(dataMat, techRepGroups)
```

#### **Arguments**

```
dataMat NormalyzerDE data matrix techRepGroups Technical replicates vector
```

### Value

collDataMat Reduced data matrix

```
techRep <- c("a", "a", "b", "b", "c", "c", "d", "d")
testData <- data.frame(
    c(1,1,1),
    c(1,2,1),
    c(3,3,3),
    c(5,3,3),
    c(5,5,4),
    c(5,5,5),
    c(7,7,7),
    c(7,9,7))
colnames(testData) <- c("a1", "a2", "b1", "b2", "c1", "c2", "d1", "d2")
statObj <- reduceTechnicalReplicates(testData, techRep)</pre>
```

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setupJobDir

Create empty directory for run

#### **Description**

Creates a directory at provided path named to the jobname.

#### Usage

```
setupJobDir(jobName, outputDir)
```

### Arguments

jobName

Name of the run.

outputDir

Path to directory where to create the output directory.

#### Value

Path to newly created directory.

#### **Examples**

```
setupJobDir("job_name", "path/to/outdir")
```

setupRawContrastObject

Prepare SummarizedExperiment object for statistics data

### **Description**

Prepare SummarizedExperiment object for statistics data

#### Usage

```
setupRawContrastObject(dataPath, designPath, sampleColName)
```

#### **Arguments**

dataPath Path to raw data matrix designPath Path to design matrix

sampleColName Name for column in design matrix containing sample names

#### Value

experimentObj Prepared instance of SummarizedExperiment

```
data_path <- system.file(package="NormalyzerDE", "extdata", "tiny_data.tsv")
design_path <- system.file(package="NormalyzerDE", "extdata", "tiny_design.tsv")
sumExpObj <- setupRawContrastObject(data_path, design_path, "sample")</pre>
```

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setupRawDataObject Prepare SummarizedExperiment object for raw data to be normalized containing data, design and annotation information	setupRawDataObject	
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#### **Description**

Prepare SummarizedExperiment object for raw data to be normalized containing data, design and annotation information

#### Usage

```
setupRawDataObject(dataPath, designPath, inputFormat = "default",
  zeroToNA = FALSE, sampleColName = "sample", groupColName = "group")
```

### **Arguments**

dataPath File path to data matrix.

designPath File path to design matrix.

inputFormat Type of matrix for data, can be either 'default', 'proteios', 'maxquantprot' or

'maxquantpep'

zeroToNA If TRUE zeroes in the data is automatically converted to NA values

sampleColName Column name for column containing sample names groupColName Column name for column containing condition levels

#### Value

experimentObj SummarizedExperiment object loaded with the data

#### **Examples**

```
data_path <- system.file(package="NormalyzerDE", "extdata", "tiny_data.tsv")
design_path <- system.file(package="NormalyzerDE", "extdata", "tiny_design.tsv")
df <- setupRawDataObject(data_path, design_path)</pre>
```

write Normalized Datasets

Write normalization matrices to file

#### **Description**

Outputs each of the normalized datasets to the specified directory.

#### Usage

```
writeNormalizedDatasets(nr, jobdir, includePairwiseComparisons = FALSE,
  includeCvCol = FALSE, includeAnovaP = FALSE,
  normSuffix = "-normalized.txt",
  rawdataName = "submitted_rawdata.txt")
```

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### **Arguments**

nr Results object.

jobdir Path to output directory.

include Pairwise Comparisons

Include p-values for pairwise comparisons.

includeCvCol Include CV column in output.
includeAnovaP Include ANOVA p-value in output.

normSuffix String used to name output together with normalization names.

rawdataName Name of output raw data file.

### Value

None

```
data(example_summarized_experiment)
normObj <- getVerifiedNormalyzerObject("job_name", example_summarized_experiment)
normResults <- normMethods(normObj)
normResultsWithEval <- analyzeNormalizations(normResults)
outputDir <- tempdir()
writeNormalizedDatasets(normResultsWithEval, outputDir)</pre>
```

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