# Starr

# October 5, 2010

bpmapToProbeAnno

Creating a probeAnno object

# **Description**

This function allows the user to create a probeAnno environment that holds the mapping between probes on the array and their genomic match position(s). The function takes an Affymetrix bpmap file as input.

## Usage

```
bpmapToProbeAnno(bpmap, verbose=T, uniqueSeq=T)
```

#### **Arguments**

bpmap Either a list, created by the function readBpmap() from the affy package. Or a

path to the bpmap file.

verbose should the progress be printed out?

uniqueSeq If TRUE, probes sequences that occur more than once on the chip (and conse-

quently match several positions on the genome) are set to 1 in the probeAnno object. Unique probes are set to 0. If false, all probes are set to 0. To identify all unique and multiple matching probes, a remapping of the probes to the genome

is recommended.

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

```
##
# dataPath <- system.file("extdata", package="Starr")
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))
# probeAnnoChr1 <- bpmapToProbeAnno(bpmapChr1)</pre>
```

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cmarrt.ma	Compute moving average statistics by incorporating the correlation structure

## **Description**

This function extends the moving average approach by incorporating the correlation structure. It also outputs the p-values of the standardized moving average statistics under the Gaussian approximation.

#### Usage

```
cmarrt.ma(eSet, probeAnno, chr=NULL, M=NULL, frag.length, window.opt='fixed.probe'
```

## Arguments

eSet ExpressionSet containing the normalized ratio

probeAnno probeAnno object with mapping

chr which chromosome should be analysed? If chr==NULL, all chromosome in the probeAnno object are taken.

M rough estimate of the percentage of bound probes. If unknown, leave it NULL.

frag.length average fragment length from sonication.

window.opt option for sliding window, either "fixed.probe" or "fixed.gen.dist". Default is 'fixed.probe'.

## **Details**

Computation using window.opt = "fixed.probe" calculates the moving average statistics within a fixed number of probes and is more efficient. Use this option if the tiling array is regular with approximately constant resolution. window.opt="fixed.gen.dist" computes the moving average statistics over a fixed genomic distance.

## Value

data.sort	datafile sorted by genomic position.
ma	unstandardized moving average(MA) statistics.
z.cmarrt	standardized MA under correlation structure.
z.indep	standardized MA under independence (ignoring correlation structure).
pv.cmarrt	p-values of probes under correlation.
pv.indep	p-values of probes under independence (ignoring correlation structure).

#### Note

The p-values are obtained under the Gaussian approximation. Therefore, it is important to check the normal quantile-quantile plot if the Gaussian approximation is valid. The function also outputs the computation under independence (ignoring the correlation structure) for comparisons.

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#### Author(s)

Pei Fen Kuan, Adam Hinz

#### References

P.F. Kuan, H. Chun, S. Keles (2008). CMARRT: A tool for the analysiz of ChIP-chip data from tiling arrays by incorporating the correlation structure. *Pacific Symposium of Biocomputing* **13**:515-526

#### See Also

```
plotcmarrt,cmarrt.peak
```

#### **Examples**

```
# dataPath <- system.file("extdata", package="Starr")
# bpmapChrl <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chrl.bpmap"))
# cels <- c(file.path(dataPath, "Rpb3_IP_chrl.cel"), file.path(dataPath, "wt_IP_chrl.cel"),
# file.path(dataPath, "Rpb3_IP2_chrl.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")
# type <- c("IP", "CONTROL", "IP")
# rpb3Chrl <- readCelFile(bpmapChrl, cels, names, type, featureData=TRUE, log.it=TRUE)
# ips <- rpb3Chrl$type == "IP"
# controls <- rpb3Chrl$type == "CONTROL"
# rpb3_rankpercentile <- normalize.Probes(rpb3Chrl, method="rankpercentile")
# description <- c("Rpb3vsWT")
# rpb3_rankpercentile_ratio <- getRatio(rpb3_rankpercentile, ips, controls, description,
# probeAnnoChrl <- bpmapToProbeAnno(bpmapChrl)
# peaks <- cmarrt.ma(rpb3_rankpercentile_ratio, probeAnnoChrl, chr=NULL, M=NULL, 250, windown in the control in
```

cmarrt.peak

Obtain bound regions for a given error rate control

#### **Description**

Obtain bound regions under a given error rate control using correction method from p.adjust.

## Usage

```
cmarrt.peak(cmarrt.ma, alpha, method, minrun, asCherList=FALSE)
```

## Arguments

```
cmarrt.ma output object from cmarrt.ma.
alpha error rate control for declaring bound region.
method correction method inherited from p.adjust.
minrun minimum number of probes to be called a bound region.
asCherList If TRUE, result is returned as class cherList. See Ringo, for further description.
```

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#### **Details**

The function returns two objects, cmarrt.bound and indep.bound. Each object is a list of bound regions which can be accessed by \$chr (chromosome), \$peak.start (start coordinate of each bound region), \$peak.stop (stop coordinate of each bound region), \$n.probe (number of probes within each bound region), \$min.pv (minimum p-values of each bound region), \$ave.pv (average p-values of each bound region).

#### Value

```
cmarrt.bound list of bound regions obtained under correlation structure.

indep.bound list of bound regions obtained under independence (ignoring correlation).
```

#### Note

The list of bound regions obtained under independence (ignoring the correlation structure) is for comparison. It is not recommended to use this list for downstream analysis.

#### Author(s)

Pei Fen Kuan, Adam Hinz

#### References

P.F. Kuan, H. Chun, S. Keles (2008). CMARRT: A tool for the analysiz of ChIP-chip data from tiling arrays by incorporating the correlation structure. *Pacific Symposium of Biocomputing* **13**:515-526.

# See Also

```
cmarrt.ma,p.adjust
```

```
# dataPath <- system.file("extdata", package="Starr")
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))
# cels <- c(file.path(dataPath, "Rpb3_IP_chr1.cel"), file.path(dataPath, "wt_IP_chr1.cel"),
# file.path(dataPath, "Rpb3_IP2_chr1.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")
# type <- c("IP", "CONTROL", "IP")
# rpb3Chr1 <- readCelFile(bpmapChr1, cels, names, type, featureData=TRUE, log.it=TRUE)
# ips <- rpb3Chr1$type == "IP"
# controls <- rpb3Chr1$type == "CONTROL"
# rpb3_rankpercentile <- normalize.Probes(rpb3Chr1, method="rankpercentile")
# description <- c("Rpb3vsWT")
# rpb3_rankpercentile_ratio <- getRatio(rpb3_rankpercentile, ips, controls, description,
# probeAnnoChr1 <- bpmapToProbeAnno(bpmapChr1)
# peaks <- cmarrt.ma(rpb3_rankpercentile_ratio, probeAnnoChr1, chr=NULL, M=NULL, 250, window
# peaklist <- cmarrt.peak(peaks)</pre>
```

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correlationPlot

correlation of ChIP signals to other data

## **Description**

correlationPlot The correlationPlot can be used to vizualize e.g. the correlation between the mean binding intensity of specific regions around annotated features and gene expression. The regions around the annotated features, that should be analyzed are defined in a data frame. Each row represents one region. In the upper panel of the plot, the correlation is plotted in a barplot. In the lower panel, the annotated feature and the regions defined in the data frame are shown.

## Usage

```
correlationPlot(regions, labels=c("start", "stop"), ...)
```

## **Arguments**

regions

a data frame, containing four columns. Every row defines one region to be analyzed and is plotted in the lower panel. pos=start, upstream=500 and downstream=500 mean characterize the region 500 bp upstream and downstream around the start of the feature. The pos columns is a character with values out of c("start", "region", "end"). upstream and downstream ar integers, indicating how many bases upstream and downstream from the specified position in the feature are included. level is an integer, that says at which level the rectangle in the lower device should be plotted. The numeration goes from the bottom to the ceiling. cor is the correlation of the region, which is plotted in the upper panel.

labels

a character vector which holds the names of the borders of the annotated region.

(e.g. c("TSS", "TTS") for transcripts)

. . . parameters, that are passed to barplot (for plotting the upper panel)

#### Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### See Also

barplot

correlationPlot(info)

```
## Constructing an example data frame
pos <- c("start", "start", "start", "region", "region", "region", "region", "stop", "stop", "
upstream <- c(500, 0, 250, 0, 0, 500, 500, 0, 250)
downstream <- c(0, 500, 250, 0, 500, 0, 500, 0, 500, 250)
level <- c(1, 1, 2, 3, 4, 5, 6, 1, 1, 2)
cor <- seq(-1,1, length=10)
info <- data.frame(pos=pos, upstream=upstream, downstream=downstream, level=level, cor=corownames(info) <- letters[1:10]
## Generate plot</pre>
```

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densityscatter

Compute density of a scatterplot

# Description

A 2d density is computed by kde2D.

# Usage

```
densityscatter(x,y,pch=19,cex=1,ncol=30,grid=100,palette="heat", add=F,...)
```

# Arguments

X	x coordinate of data
У	y coordinate of data
pch	type of point
cex	A numerical value giving the amount by which plotting text and symbols should be magnified relative to the default
grid	Number of grid points in each direction
ncol	number of colors
palette	color palette to choose
add	should data points be added to an exisiting plot?
	parameters passed to plot or points

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

# See Also

```
kde2dplot
```

```
##
points = 10^4
x <- rnorm(points/2)
x = c(x,x+2.5)
y <- x + rnorm(points,sd=0.8)
x = sign(x)*abs(x)^1.3
densityscatter(x,y)</pre>
```

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expressionByFeature

Getting expression value by feature from an ExpressionSet

# Description

This function gets the expression of a specified feature (e.g. orf, gene) from an ExpressionSet.

# Usage

```
expressionByFeature(eSet, fkt, method="median")
```

# **Arguments**

eSet An ExpressionSet, containing the normalized expression values

fkt Function to convert the featureNames (e.g. affy IDs) of eSet to the required

features (e.g. ORFs)

method If one feature (e.g. ORF) has more than one feature (e.g. affy ID) on the chip,

the mean/median over the intensities is taken

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

# See Also

mget

filterGenes Filter Features/Genes	filterGenes	Filter Features/Genes	
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# Description

This fucntion filters genes and other annotated features with respect to length, ovelaps and distance to other features.

## Usage

```
filterGenes(gffAnno, distance_us=500, distance_ds=500, minLength=-Inf, maxLength
```

# Arguments

gffAnno	a data frame containing the annotation
distance_us	how many basepairs upstream to the feature should not overlap with other features.
distance_ds	how many basepairs downstream to the feature should not overlap with other features.
minLength	minimal length of the feature
maxLength	maximal length of the feature

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#### Value

a character vector with the names of the features, that passed the filter.

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### **Examples**

```
##
# dataPath <- system.file("extdata", package="Starr")
# transcriptAnno <- read.gffAnno(file.path(dataPath, "transcriptAnno.gff"), feature="tran"
# filtered_transcripts <- filterGenes(transcriptAnno, distance_us = 0, distance_ds = 0, m</pre>
```

getMeans

Get mean ChIP-signal over annotated features

# Description

getMeans calculates the mean ChIP-signal over annotated features

# Usage

```
getMeans(eSet, probeAnno, geneAnno, regions)
```

#### **Arguments**

eSet an ExpressionSet

probeAnno object for the given ExpressionSet

geneAnno a data frame containing the annotation of the features of interest

regions a data frame, containing four columns. The pos columns is a character with

values out of c("start", "region", "end"). upstream and downstream ar integers, indicating how many bases upstream and downstream from the specified position in the feature are included. level is an integer, that says at which level the rectangle in the lower device should be plotted. The numeration goes from the bottom to the ceiling. cor is the correlation of the region, which is plotted in the

upper panel.

# Value

a list. Each entry contains the mean signals over the specified regions (in the regions data frame) of all features in geneAnno.

## Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

## See Also

```
getProfiles
```

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#### **Examples**

```
##
# dataPath <- system.file("extdata", package="Starr")</pre>
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))</pre>
# cels <- c(file.path(dataPath, "Rpb3_IP_chr1.cel"), file.path(dataPath, "wt_IP_chr1.cel"),
 file.path(dataPath, "Rpb3_IP2_chr1.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")</pre>
# type <- c("IP", "CONTROL", "IP")
# rpb3Chr1 <- readCelFile(bpmapChr1, cels, names, type, featureData=TRUE, log.it=TRUE)</pre>
# ips <- rpb3Chr1$type == "IP"</pre>
# controls <- rpb3Chr1$type == "CONTROL"
# rpb3_rankpercentile <- normalize.Probes(rpb3Chr1, method="rankpercentile")</pre>
# description <- c("Rpb3vsWT")</pre>
# rpb3_rankpercentile_ratio <- getRatio(rpb3_rankpercentile, ips, controls, description,</pre>
# probeAnnoChr1 <- bpmapToProbeAnno(bpmapChr1)</pre>
# transcriptAnno <- read.gffAnno(file.path(dataPath, "transcriptAnno.gff"), feature="trar
# filtered_orfs <- filterGenes(transcriptAnno, distance_us = 0, distance_ds = 0, minLengt
# pos <- c("start", "start", "start", "region", "region", "region", "region", "stop", "stop"</pre>
# upstream <- c(500, 0, 250, 0, 0, 500, 500, 500, 0, 250)
# downstream <- c(0, 500, 250, 0, 500, 0, 500, 0, 500, 250)
# info <- data.frame(pos=pos, upstream=upstream, downstream=downstream, stringsAsFactors=
# means_rpb3 <- getMeans(rpb3_rankpercentile_ratio, probeAnnoChr1, transcriptAnno[which(t
```

getProfiles

Get profiles of ChIP-signal over annotated features

# Description

This function associates the measured ChIP signals to annotated features and stores the profile of each feature in a list. Each profile is divided in three parts. The first entry is "upstream", which saves the signal upstream of start. Then follows "region", which is from start to end and then "downstream", which stores the signals downstream of end.

#### Usage

```
getProfiles(eSet, probeAnno, gffAnno, upstream, downstream, feature="ORF", borde
```

### Arguments

eSet an ExpressionSet, containing on sample.

probeAnno a probeAnno object for the given ExpressionSet

gffAnno a data frame containing the annotation of the features of interest

upstream how many basepairs upstream of the feature start (feature start on the crick strand is end in gffAnno) should be taken.

downstream how many basepairs downtream of the feature start (feature end on the crick strand is start in gffAnno) should be taken.

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feature name of the features (e.g. ORF, transcript, rRNA, ...)

borderNames names of the borders, flaking the feature (e.g. c("start", "stop"))

method Two methods are available. "middle", just takes the middle position of each

probe and its corresponding value. This method should be used if the whole genome is tiled in an high resolution. "basewise" calculates for each base the

mean of all probes overlapping with this position.

fill if "middle" is chosen the distance of the taken values equals the probe spacing

on the chip. To avoid errors, because of regions lacking of probes, one can fill

up these regions with NAs.

distance if method "middle" and fill==TRUE are chosen, distance is the max distance of

no value occuring before filling in one NA.

spacing probe spacing on the chip. Only used for filling up with NAs in method "mid-

dle".

sameLength if method "middle" is chosen it can occur that the length of the upstream/downstream

region vary a little. If sameLength==TRUE, upstream/downstream regions get

all the same length.

#### Value

a list with the following entries

ID the ID/name of the sample

upstream number of basepairs, taken upstream of the feature downstream number of basepairs, taken upstream of the feature

method used

 $\begin{array}{ll} \mbox{borderNames} & \mbox{names of the borders} \\ \mbox{feature} & \mbox{feature type (e.g. "ORF")} \end{array}$ 

profile a list which contains all profiles of the features in the gffAnno. Each entry

consists of a list with the elements "upstream", "region", "downstream".

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### See Also

fill,fillNA,mapFeatures,getIntensities,getFeature,fill,getProfilesByBase

```
##
# dataPath <- system.file("extdata", package="Starr")
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))

# cels <- c(file.path(dataPath, "Rpb3_IP_chr1.cel"), file.path(dataPath, "wt_IP_chr1.cel"),
# file.path(dataPath, "Rpb3_IP2_chr1.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")
# type <- c("IP", "CONTROL", "IP")
# rpb3Chr1 <- readCelFile(bpmapChr1, cels, names, type, featureData=TRUE, log.it=TRUE)

# ips <- rpb3Chr1$type == "IP"</pre>
```

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```
# controls <- rpb3Chr1$type == "CONTROL"

# rpb3_rankpercentile <- normalize.Probes(rpb3Chr1, method="rankpercentile")

# description <- c("Rpb3vsWT")

# rpb3_rankpercentile_ratio <- getRatio(rpb3_rankpercentile, ips, controls, description,

# probeAnnoChr1 <- bpmapToProbeAnno(bpmapChr1)

# transcriptAnno <- read.gffAnno(file.path(dataPath, "transcriptAnno.gff"), feature="transcriptAnno")

# profile <- getProfiles(rpb3_rankpercentile_ratio, probeAnnoChr1, transcriptAnno, 500, 5</pre>
```

getRatio

Building ratio over experiments

# **Description**

This function calculates the ratio over experiments.

# Usage

```
getRatio(eSet, ip, control, description, fkt=median, featureData=F)
```

## **Arguments**

eSet	An ExpressionSet, containing the logged raw intensities
ip	a boolean or integer vector, that indicate, which columns in the matrix are IP experiments
control	a boolean or integer vector, that indicate, which columns in the matrix are CONTROL or REFERENCE experiments
description	description of the new data (e.g. IPvsCONTROL)
fkt	mean or median to calculate the averaged intensity over replicates
featureData	if TRUE, featureData is added to the new ExpressionSet

## Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

```
##
# dataPath <- system.file("extdata", package="Starr")
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))

# cels <- c(file.path(dataPath, "Rpb3_IP_chr1.cel"), file.path(dataPath, "wt_IP_chr1.cel"),
# file.path(dataPath, "Rpb3_IP2_chr1.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")
# type <- c("IP", "CONTROL", "IP")
# rpb3Chr1 <- readCelFile(bpmapChr1, cels, names, type, featureData=TRUE, log.it=TRUE)

# ips <- rpb3Chr1$type == "IP"
# controls <- rpb3Chr1$type == "CONTROL"</pre>
```

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```
# rpb3_rankpercentile <- normalize.Probes(rpb3Chr1, method="rankpercentile")
# description <- c("Rpb3vsWT")
# rpb3_rankpercentile_ratio <- getRatio(rpb3_rankpercentile, ips, controls, description,</pre>
```

heatmapplot

heatmapplot

# **Description**

Heatmap representation of binding profiles

# Usage

```
heatmapplot(profiles, colpal=c("black", "dark blue", "dark green", "green", "gold",
```

# **Arguments**

profiles a list of profiles returned by getProfiles(). Features must have same lengths.

colpal color palette for intensity coding

abl positions of vertical lines that are added to the panel subset subset of genes in the list that should be plotted

#### Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

list2matrix

Convert profile list to matrix

#### **Description**

This function converts the list of profiles generated by the getProfiles function to a matrix, if all entries have the same length.

## Usage

```
list2matrix(profiles)
```

#### **Arguments**

```
profiles a list, generated by the getProfiles
```

## Value

a list with with a matrix at the entry profile.

#### Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

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makeProbeAnno

Creating a probeAnno object

# **Description**

Creates a probeAnno object (package: Ringo) from a given Affymetrix bpmap file or a Nimblegen POS file. The posToProbeAnno function from the Ringo package is called to build the object.

## Usage

```
makeProbeAnno(posFile=NULL, bpmap=NULL, probeIDAsStrings=F)
```

# **Arguments**

posFile path to the POS file

bpmap Either a list, created by the function readBpmap() from the affy package, or a

path to the bpmap file.

probeIDAsStrings

should the mapping of the probes to the rows in the assayData be integers or

characters.

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### See Also

posToProbeAnno,readBpmap

makeSplines

Fit splines to profiles

# **Description**

This function uses the pspline package to fit spilnes to each entry in a list of profiles.

# Usage

```
makeSplines(profiles, df=1000)
```

## **Arguments**

profiles a list as it is created by the getProfiles package.

df the degree of freedom of the fit

#### Value

a list as it is created by the getProfiles function.

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#### Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### See Also

```
smooth.Pspline, predict.smooth.Pspline
```

### **Examples**

```
##
# dataPath <- system.file("extdata", package="Starr")</pre>
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))</pre>
# cels <- c(file.path(dataPath, "Rpb3_IP_chr1.cel"), file.path(dataPath, "wt_IP_chr1.cel"),</pre>
# file.path(dataPath, "Rpb3_IP2_chr1.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")</pre>
# type <- c("IP", "CONTROL", "IP")
# rpb3Chr1 <- readCelFile(bpmapChr1, cels, names, type, featureData=TRUE, log.it=TRUE)</pre>
# ips <- rpb3Chr1$type == "IP"</pre>
# controls <- rpb3Chr1$type == "CONTROL"</pre>
# rpb3_rankpercentile <- normalize.Probes(rpb3Chr1, method="rankpercentile")</pre>
# description <- c("Rpb3vsWT")</pre>
# rpb3_rankpercentile_ratio <- getRatio(rpb3_rankpercentile, ips, controls, description,
# probeAnnoChr1 <- bpmapToProbeAnno(bpmapChr1)</pre>
# transcriptAnno <- read.gffAnno(file.path(dataPath, "transcriptAnno.gff"), feature="transcriptAnno.gff")</pre>
# profile <- getProfiles(rpb3_rankpercentile_ratio, probeAnnoChr1, transcriptAnno, 500, 5
# profile_splines <- makeSplines(profile)</pre>
```

normalize.Probes Normalization of probes

# **Description**

Normalization of probe intensities with a given method.

#### Usage

```
normalize.Probes(eSet, method=NULL, ratio=FALSE, ip, control, description, fkt=m
```

# **Arguments**

eSet	An ExpressionSet, containing the logged raw intensities
method	character string specifying the normalization method to be used. Choices are "none", "scale", "quantile", "Aquantile", "Gquantile", "Rquantile", "Tquantile", "vsn", "rankpercentile", "loess", "substract".
ratio	if TRUE, the ratios are calcualted.
ip	a boolean vector, indicating which sample are IP experiments
control	a boolean vector, indicating which sample are CONTROL experiments

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description description of the normalized data

fkt function to chose for averaging over replicates

featureData should the featureData of eSet be passed to the new ExpressionSet?

targets vector, factor or matrix of length twice the number of arrays, used to indicate target groups if method="Tquantile"

arrays Subset of experiments (colnames in ExpressionSet) in the ExpressionSet, that are supposed to be normalized seperately.

... arguments, that should be passed to the normalization methods.

#### **Details**

The procedure calls different functions from this package or from affy and limma, depending on the method.

none Calls normalizeWithinArrays with method="none" from package limma.

scale Calls normalizeWithinArrays with method="scale" from package limma.

quantile Calls normalizeBetweenArrays with method="quantile" from package limma.

**Gquantile** Calls normalizeBetweenArrays with method="Gquantile" from package limma.

Rquantile Calls normalizeBetweenArrays with method="Rquantile" from package limma.

**Tquantile** Calls normalizeBetweenArrays with method="Tquantile" from package limma.

Rquantile Calls normalizeBetweenArrays with method="Rquantile" from package limma.

vsn Calls normalizeBetweenArrays with method="vsn" from package limma.

loess Calls normalize.loess from package affy.

rankpercentile Calls rankPercentile.normalize from this package.

substract Calls substract from this package.

#### Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

# See Also

normalizeBetweenArrays, normalize.loess, substract, rankPercentile.normalize

```
##
# dataPath <- system.file("extdata", package="Starr")
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))

# cels <- c(file.path(dataPath, "Rpb3_IP_chr1.cel"), file.path(dataPath, "wt_IP_chr1.cel"),
# file.path(dataPath, "Rpb3_IP2_chr1.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")
# type <- c("IP", "CONTROL", "IP")
# rpb3Chr1 <- readCelFile(bpmapChr1, cels, names, type, featureData=TRUE, log.it=TRUE)
# rpb3_rankpercentile <- normalize.Probes(rpb3Chr1, method="rankpercentile")</pre>
```

16 plotBoxes

plotDensity

density plots of experiments

# **Description**

Generates a plot, showing the densities of the experiments.

## Usage

```
plotDensity(eSet, oneDevice=T, main="")
```

# **Arguments**

eSet an ExprssionSet or a matrix, containing the data oneDevice should all lines be plotted to one device?

main head of the plot

#### Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### See Also

```
plot.default, density
```

#### **Examples**

```
##
mat <- matrix(rnorm(1000000), ncol=2)
colnames(mat) <- c("Sample1", "Sample2")
mat[,1] <- mat[,1]-2
plotDensity(mat)</pre>
```

plotBoxes

boxplots of experiments

# Description

Generates a boxplot of the of the given experiments.

## Usage

```
plotBoxes(eSet, col=NULL)
```

# Arguments

eSet Either an ExpressionSet or a matrix, containing the data.

col color, to fill the boxes

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# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### See Also

```
boxplot
```

# **Examples**

```
##
mat <- matrix(rnorm(1000000), ncol=2)
colnames(mat) <- c("Sample1", "Sample2")
mat[,1] <- mat[,1]-2
plotBoxes(mat)</pre>
```

plotcmarrt

Histogram of p-values and normal QQ plots for standardized MA statistics

# **Description**

Plot the histograms of p-values and normal QQ plots under correlation structure and independence.

# Usage

```
plotcmarrt(cmarrt.ma, freq=FALSE)
```

#### **Arguments**

```
cmarrt.ma output object from cmarrt.ma.

freq see ?hist
```

# **Details**

Diagnostic plots for comparing the distribution of standardized MA statistics under correlation and independence.

# Value

Histogram of p-values and normal QQ plots under correlation structure and independence.

#### Note

If the normal quantile-quantile plot deviates from the reference line for unbound probes, this indicates that Gaussian approximation is not suitable for analyzing this data.

# Author(s)

Pei Fen Kuan, Adam Hinz

18 plotGCbias

#### References

P.F. Kuan, H. Chun, S. Keles (2008). CMARRT: A tool for the analysiz of ChIP-chip data from tiling arrays by incorporating the correlation structure. *Pacific Symposium of Biocomputing* **13**:515-526

#### See Also

```
cmarrt.ma,qqnorm
```

### **Examples**

```
# dataPath <- system.file("extdata", package="Starr")
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))
# cels <- c(file.path(dataPath, "Rpb3_IP_chr1.cel"), file.path(dataPath, "wt_IP_chr1.cel"),
# file.path(dataPath, "Rpb3_IP2_chr1.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")
# type <- c("IP", "CONTROL", "IP")
# rpb3Chr1 <- readCelFile(bpmapChr1, cels, names, type, featureData=TRUE, log.it=TRUE)
# ips <- rpb3Chr1$type == "IP"
# controls <- rpb3Chr1$type == "CONTROL"
# rpb3_rankpercentile <- normalize.Probes(rpb3Chr1, method="rankpercentile")
# description <- c("Rpb3vsWT")
# rpb3_rankpercentile_ratio <- getRatio(rpb3_rankpercentile, ips, controls, description,
# probeAnnoChr1 <- bpmapToProbeAnno(bpmapChr1)
# peaks <- cmarrt.ma(rpb3_rankpercentile_ratio, probeAnnoChr1, chr=NULL, M=NULL,250,windown plotcmarrt(peaks)</pre>
```

plotGCbias

Visualize GC-Bias of Hybridization

# Description

Generates a plot showing the GC-bias of the hybridization.

# Usage

```
plotGCbias(intensity, sequence, main="")
```

## **Arguments**

```
intensity a vector of type numeric, containing the measured intensities.

sequence a vector of type character, containing the sequences.

main head of the plot
```

## Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

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#### See Also

```
boxplot
```

# **Examples**

```
##
sequence <- unlist(lapply(1:50000, function(x) {paste(sample(c("A","T","C","G"),prob=c(0.values <- runif(50000,min=-2,max=2)
plotGCbias(values, sequence)</pre>
```

plotImage

Reconstruct the array image

# **Description**

Function to visualize spatial distribution of raw intensities on Affymetrix Oligoarrays.

# Usage

```
plotImage(cel)
```

# Arguments

cel

a character, specifying the path to the CEL file

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

# See Also

```
readCel,levelplot
```

```
# dataPath <- system.file("extdata", package="Starr")
# plotImage(file.path(dataPath, "Rpb3_IP_chr1.cel"))</pre>
```

20 plotPosBias

plotMA

M versus A plot

# **Description**

A matrix of M vs. A plots of each pair (ip, control) is produced.

# Usage

```
plotMA(eSet, ip=NULL, control=NULL, col=NULL)
```

# **Arguments**

eSet	an ExprssionSet or matrix, containing the data
ip	an integer, or boolean vector, that indicates, which columns in the ExpressionSet are IP experiments
control	an integer, or boolean vector, that indicates, which columns in the ExpressionSet are CONTROL or REFERENCE experiments
col	color, to fill the boxes

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

# See Also

```
ma.plot
```

# **Examples**

```
##
mat <- matrix(rnorm(1000000), ncol=4)
colnames(mat) <- c("Sample1", "Sample2", "Sample3", "Sample4")
mat[,1] <- mat[,1]^2
plotMA(mat, c(TRUE, FALSE, TRUE, FALSE), c(FALSE, TRUE, FALSE, TRUE))</pre>
```

plotPosBias

Bias of hybridzation, depending on base position in sequence.

## **Description**

plotPosBias generates a plot showing the bias of hybridzation, depending on base position in sequence.

# Usage

```
plotPosBias(intensity, sequence, main="", ylim)
```

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# **Arguments**

a vector of type numeric, containing the measured intensities intensity

a vector of type character, containing the sequneces sequence

main head of the plot ylim ylim of plot

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

# **Examples**

```
##
 sequence <- unlist(lapply(1:50000, function(x) \{paste(sample(c("A","T","C","G"),prob=c(0.0000, function(x)),prob=c(0.0000, function(x), function(x
values <- runif(50000,min=-2,max=2)</pre>
plotPosBias(values, sequence)
```

plotRatioScatter Plot ratios of all possible combinations of IP and CONTROL

# **Description**

A matrix of pairwise scatterplots of the ratios is created. The lower panel shows the correlation of the data.

# Usage

```
plotRatioScatter(eSet, ip, control, density=F, sample=NULL, cluster=T, cex=1)
```

# **Arguments**

eSet		an ExprssionSet or matrix, containing the data
ip		an integer, or boolean vector, that indicates, which columns in the ExpressionSet are IP experiments
cont	rol	an integer, or boolean vector, that indicates, which columns in the ExpressionSet are CONTROL or REFERENCE experiments
dens	sity	if TRUE, a density scatter plot is plotted. This plot shows the density of the data.
samp	ole	An integer, indicating the number of subsamples to take for the density scatter- plot. This is only recommended if the data is very large, as the density compu- tation takes some time. #
clust	er	if cluster=T, the experiments are clustered and similiar experiments are plotted together.
cex		see ?par

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

22 plotScatter

#### See Also

```
pairs, density scatter
```

# **Examples**

```
##
points <- 10^4
x <- rnorm(points/2)
x <- c(x,x+2.5)
x <- sign(x)*abs(x)^1.3
y <- x + rnorm(points,sd=0.8)
z <- y*2
mat <- matrix(c(x,y,z), ncol=3)
colnames(mat) <- c("A", "B1", "B2")
plotRatioScatter(mat, c(TRUE, FALSE, FALSE), c(FALSE, TRUE, TRUE), density=TRUE)</pre>
```

plotScatter

High level scatterplot of experiments

# Description

A matrix of pairwise scatterplots is created. The lower panle shows the correlation of the data.

#### Usage

```
plotScatter(eSet, density=F, cluster=T, sample=NULL, cex=1)
```

# **Arguments**

eSet	an ExprssionSet or matrix, containing the data
density	if TRUE, a density scatter plot is plotted. This plot shows the density of the data.
sample	An integer, indicating the number of subsamples to take for the density scatter- plot. This is only recommended if the data is very large, as the density compu- tation takes some time.
cluster	if cluster=T, the experiments are clustered and similiar experiments are plotted together.
cex	see ?par

# Author(s)

```
Benedikt Zacher < zacher@lmb.uni-muenchen.de>
```

# See Also

```
pairs, density scatter
```

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## **Examples**

```
##
points <- 10^4
x <- rnorm(points/2)
x <- c(x,x+2.5)
x <- sign(x) *abs(x)^1.3
y <- x + rnorm(points,sd=0.8)
mat <- matrix(c(x,y), ncol=2)
colnames(mat) <- c("a", "b")
plotScatter(mat, density=TRUE)</pre>
```

plotProfiles

Plotting ChIP profiles of one or more clusters

# Description

plotProfiles plots the ChIP profiles of one or more clusters. Additionally on can display the distribution of e.g. gene expression in the clusters.

# Usage

```
plotProfiles(profiles, mfcol=NULL, mfrow=NULL, ylab="intensity", xlab="position"
```

#### **Arguments**

```
profiles
                  a list constructed by the function getProfiles().
mfcol
                  see ?par
mfrow
                  see ?par
                  see ?par
ylab
xlab
                  see ?par
                  a list of named vectors. Density plots are created for every vector and cluster.
histograms
                  A named integer vector, that maps the features to the cluster.
cluster
                  should a clusterplot be shown?
profileplot
meanprofile
                  should the mean profiles of each cluster be plotted??
                  arguments, passed to plot.default
```

#### Author(s)

```
Benedikt Zacher < zacher@lmb.uni-muenchen.de>
```

#### See Also

```
density, profileplot
```

24 profileplot

#### **Examples**

```
##
sampls = 100
probes = 63
clus = matrix(rnorm(probes*sampls,sd=1),ncol=probes)
clus= rbind( t(t(clus)+sin(1:probes/10))+1:nrow(clus)/sampls , t(t(clus)+sin(pi/2+1:probeclustering = kmeans(clus,3)$cluster
names(clustering) <- 1:length(clustering)

profiles <- apply(clus, 1, function(x) {list(upstream=x[1:20], region=x[21:43], downstreanames(profiles) <- 1:length(clustering)
profiles <- list(profile=profiles, upstream=20, downstream=20, borderNames=c("start", "start")
plotProfiles(profiles, cluster=clustering, ylim=c(-1,2.5), type="1", lwd=2)</pre>
```

profileplot

Vizualize clusters

# **Description**

Visualization of a set of "profiles" (i.e. a consecutive series of measurements like a time series, or the DNA binding levels along different positions on a gene). The profiles are given as the rows of a (samples x positions) matrix that contains the measurements. Instead of plotting a line for each profile (row of the matrix), the q-quantiles for each position (column of the matrix) are calculated, where q runs through a set of representative quantiles. Then for each q, a line of q-quantiles is plotted along the positions. Color coding of the quantile profiles aids the interpretation of the plot: There is a color gradient from the median profile to the 0 (=min) resp. 1(=max) quantile.

#### Usage

```
profileplot(cluster, label=NULL, at=NULL, main = "", xlim=NULL, xlab = "", xaxt =
```

# **Arguments**

cluster	a (samples x columns) matrix with numerical entries. Each sample row is understood as a consecutive series of measurements. Missing values are not allowed so far
label	if multiple clusters should be plotted in one diagram, the cluster labels for each item are given in this vector
at	optional vector of length $ncol(cluster)$ , default = 1: $ncol(cluster)$ . Specifies the x-values at which the positions will be plotted.
main	the title of the plot, standard graphics parameter
xlim	xlimits, standard graphics parameter
xlab	x-axis legend, standard graphics parameter
xaxt	should an x axis be plotted at all? (="n" if not), standard graphics parameter
xlabels	character vector. If specified, this text will be added at the "at"-positions as x-axis labels.
las	direction of the xlabels text. las=1: horizontal text, las=2: vertical text

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	ylim	ylimits, standard graphics parameter
	ylab	y-axis legend, standard graphics parameter
	fromto	determines the smallest and the largest quantile that are plotted in colors, more distant values are plotted as outliers
	colpal	either "red", "green", "blue" (predefined standard color palettes in profileplot), or a vector of colors to be used instead.
	nrcolors	not very important. How many colors will the color palette contain? Usually, the default = 25 is sufficient
	outer.col	color of the outlier lines, default = "light grey". For no outliers, choose outer.col="none"
add.quartiles		S
		should the quartile lines be plotted (grey/black)? default=TRUE
	add	should the profile plot be added to the current plot? Defaults to FALSE
	separate	should each cluster, be plotted in a separate window? Defaults to TRUE

## Author(s)

Achim Tresch, Benedikt Zacher <tresch@lmb.uni-muenchen.de>

# **Examples**

```
sampls = 100
probes = 63
at = (-31:31)*14
clus = matrix(rnorm(probes*sampls,sd=1),ncol=probes)
clus= rbind( t(t(clus)+sin(1:probes/10))+1:nrow(clus)/sampls , t(t(clus)+sin(pi/2+1:probellabs = paste("cluster",kmeans(clus,4)$cluster)

profileplot(clus,main="All data",fromto=c(0,1))
profileplot(clus,label=labs,main="Clustered data",colpal=c("heat","blue","red","topo"),accolpal=c("heat","blue","red","green"),outer.col="none")
```

read.gffAnno

Reading gff annotation

# Description

This functions reads the annotation from a gff file.

# Usage

```
read.gffAnno(gffFile, feature=NULL)
```

# Arguments

gffFile path to file

feature feature to select ("character"). If feature="gene", then only rows, representing

this feature are read.

26 readCelFile

#### Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### **Examples**

```
##
# dataPath <- system.file("extdata", package="Starr")
# transcriptAnno <- read.gffAnno(file.path(dataPath, "transcriptAnno.gff"), feature="trar
```

readCelFile

Read raw intensities from CEL files

#### **Description**

Function to read the raw intensities of the perfect match probes (PM) of Affymetrix CEL files into an ExpressionSet. This function is used to read one-color data. For two-color data use the functions from the Ringo package.

#### Usage

```
readCelFile(bpmap, cel_files, names, type, experimentData=NULL, featureData=T, l
```

# **Arguments**

bpmap Either a list, created by the function readBpmap() from the affy package, or the

path to the bpmap file.

cel\_files a character vector, specifying the path to the CEL files
names a character vector, containing the names of the experiments

type a character vector, containing the type of experiment, e.g. "IP" for an Immuno-

precipitation, or "CONTROL" for a control or reference experiment was done

experimentData

This must be an object of type MIAME, which details information about e.g.,

the investigator or lab where the experiment was done, an overall title, and other

notes

featureData

If TRUE, a featureData object is added to the ExpressionSet, containing information about the chromosome, position in the genome and sequence of the

features

log.it If TRUE, logged intesities are read

phenodata data.frame, containing columns name, type, CEL.

#### Value

Returns raw intensity values in form of an ExpressionSet with additional information:

assayData This object contains the measured probe intensities.

phenoData contains further description of the experiments, such as names or type

featureData containing information about the chromosome, position in the genome and se-

quence of the features

experimentData

details information about e.g., the investigator or lab where the experiment was

remap 27

#### Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

#### See Also

```
readCelIntensities, xy2indices
```

### **Examples**

```
##
# dataPath <- system.file("extdata", package="Starr")
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))

# cels <- c(file.path(dataPath, "Rpb3_IP_chr1.cel"), file.path(dataPath, "wt_IP_chr1.cel"),
# file.path(dataPath, "Rpb3_IP2_chr1.cel"))
# names <- c("rpb3_1", "wt_1", "rpb3_2")
# type <- c("IP", "CONTROL", "IP")
# rpb3Chr1 <- readCelFile(bpmapChr1, cels, names, type, featureData=TRUE, log.it=TRUE)</pre>
```

remap

Remap reporter sequences to the genome and create a new bpmap file

# **Description**

This function remaps the reporter sequences on the chip on the genome and outputs a new bpmap annotation, containing only unique matches to the genome. A remapping is recommended if the bpmap file was built on an outdated genome, or if sequences, that match the genome more than once should be excluded.

# Usage

```
remap(bpmap=NULL, seqs=NULL, nseq=NULL, path="", complementary=FALSE, reverse=FA
```

# Arguments

A list, created by the function readBpmap() from the affy package.

Number of sequences, that are searched in one iteration.

Seqs Sequences to search as a character vector

path path to genomic fasta files

complementary

If TRUE, the sequences are searched in the complementary strand of the text

reverse If TRUE, the sequences are searched in the reverse strand of the text

reverse\_complementary

If TRUE, the sequences are searched in the reverse complementary strand of the text

reverse\_path

If TRUE, the sequences are searched in the reverse complementary strand of the text

return\_bpmap If TRUE, the output is a list in bpmap format

## Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

28 writePosFile

#### **Examples**

```
# dataPath <- system.file("extdata", package="Starr")
# bpmapChr1 <- readBpmap(file.path(dataPath, "Scerevisiae_tlg_chr1.bpmap"))
# newbpmap <- remap(bpmapChr1, nseq=5000000, path=dataPath, reverse_complementary=TRUE, n</pre>
```

writeGFF

write ChIP-chip data to a gff file

# Description

This function writes the all columns of the assayData to a gff file.

# Usage

```
writeGFF(expressionSet, probeAnno, file)
```

#### **Arguments**

expressionSet

an ExpressionSet object

probeAnno a probeAnno object
file path to write to

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

writePosFile

Creating a pos file

# **Description**

Writes a Nimblegen pos file from a given Affymetrix bpmap file.

#### Usage

```
writePosFile(bpmap, file)
```

# **Arguments**

bpmap Either a list, created by the function readBpmap() from the affy package. Or a

path to the bpmap file.

file a character, specifying the path to the file to be written

# Author(s)

Benedikt Zacher < zacher@lmb.uni-muenchen.de>

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