# Package 'wateRmelon'

December 21, 2024

Type Package

Title Illumina DNA methylation array normalization and metrics

Version 2.13.0

**Description** 15 flavours of betas and three performance metrics, with methods for objects produced by methylumi and minfi packages.

License GPL-3

**Depends** R (>= 3.5.0), Biobase, limma, methods, matrixStats, methylumi, lumi, ROC, IlluminaHumanMethylation450kanno.ilmn12.hg19, illuminaio

Imports Biobase

Enhances minfi

Suggests RPMM, IlluminaHumanMethylationEPICanno.ilm10b2.hg19, BiocStyle, knitr, rmarkdown, IlluminaHumanMethylationEPICmanifest, irlba, FlowSorted.Blood.EPIC, FlowSorted.Blood.450k, preprocessCore

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**biocViews** DNAMethylation, Microarray, TwoChannel, Preprocessing, QualityControl

Collate as.methylumi.R bscon.R bscon\_methy.R bscon\_minfi.R getAnn.R oxyscale.R adaptRefQuantiles.R beta1.R Beta2M.R betaqn.R bgeq.R bgeqqt.R corateAnnotation.R concatenateMatrices.R coRankedMatrices.R correctI.R correctII.R dataDetectPval2NA.R db1.R detectionPval.filter.R dfs2.R dfsfit.R dmrse.R dmrse\_col.R dmrse\_row.R dyebuy1.R dyebuy2.R dyebuy3.R dyebuy4.R estimateCellCounts.R estimateSex.R filterXY.R findAnnotationProbes.R gcoms.R gcose.R genki.R genkme.R genkus.R genotype.R getMethylumiBeta.R getQuantiles.R getSamples.R getsnp.R horv.R loadMethylumi2.R lumiMethyR2.R M2Beta.R melon.R nbBeadsFilter.R normalize.quantiles2.R normalizeIlluminaMethylation.R ot.R outlyx.R pasteque.R peak.correction.R pfilter.R pipelineIlluminaMethylation.batch.R pwod.R detection.R readEPIC.R preprocessIlluminaMethylation.R

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wateF	Rmelon-package Illumina 450K arrays: normalization and performance metrics	

## Description

Functions for calculating the index of DNA methylation proportion beta in 15 different ways, and three different ways of estimating data quality or normalization performance.

.getManifestString

## **Details**

Package: wateRmelon Type: Package Version: 1.0 Date: 2012-10-10

License: GPL3

## Author(s)

Leonard C Schalkwyk, Ruth Pidsley and Chloe Wong Maintainer: Who to complain to <leonard.schalkwyk@kcl.ac.uk>

#### References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

.createAnnotation

Internal function to guess correct pData and retrieve using minfi

## **Description**

Internal function to guess correct pData and retrieve using minfi

## Usage

.createAnnotation(object)

.getManifestString

Internal functions for Illumina i450 normalization functions

## **Description**

got and fot find the annotation column differentiating type I and type II assays in MethylSet (got) or MethyLumiSet (fot) objects. pop extracts columns from IlluminaHumanMethylation450k.db

## Usage

.getManifestString(annotation)

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## **Arguments**

annotation A string naming the array type

x a MethyLumiSet

obj a MethylSet

fd a character vector of the desired annotation columns

rn a character vector of the desired features

#### **Details**

got returns a character vector of 'I' and 'II', fot returns the index of the relevant column. pop returns a data frame

#### Author(s)

lschal@essex.ac.uk

#### References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

adaptRefQuantiles Functions from 450-pipeline (Touleimat & Tost)

## **Description**

These functions are part of the 450K pipeline (Touleimat and Tost, Epigenomics 2012 4:325). For freestanding use of the normalization function, a wrapper is provided, see tost

## Author(s)

Nizar Touleimat, wrapper by Leonard.Schalkwyl@kcl.ac.uk

#### References

Touleimat N, Tost J: Complete pipeline for Infinium R Human Methylation 450K BeadChip data processing using subset quantile normalization for accurate DNA methylation estimation. Epigenomics 2012, 4:325-341

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

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adjustedDasen

adjusted Dasen

## Description

adjustedDasen utilizes dasen normliasation to normalise autosomal CpGs, and infers the sex chromosome linked CpGs by linear interpolation on corrected autosomal CpGs.

## Usage

```
adjustedDasen(
  mns,
  uns,
  onetwo,
  chr,
  offset_fit = TRUE,
  cores = 1,
  ret2 = FALSE,
  fudge = 100,
  ...
)
```

## Arguments

mns	matrix of methylated signal intensities, samples in column and probes in row.
uns	matrix of unmethylated signal intensities, samples in column and probes in row.
onetwo	character vector or factor of length nrow(mns) indicating assay type 'I' or 'II'.
chr	character vector stores the mapped chromosomes for all probes, e.g. chr <- c('1', 'X', '21',, 'Y').
offset_fit	logical (default is TRUE). To use dasen, set it TRUE; to use nasen, set it FALSE.
cores	an integer(e.g. 8) defines the number of cores to parallel processing. Default value is 1, set to -1 to use all available cores.
ret2	logical (default is FALSE), if TRUE, returns a list of intensities and betas instead of a naked matrix of betas.
fudge	default 100, a value added to total intensity to prevent denominators close to zero when calculating betas, e.g. betas <- mns / (mns + uns + fudge).
	additional argument roco for dfsfit giving Sentrix rows and columns. This allows a background gradient model to be fit. This is split from data column names by default. roco=NULL disables model fitting (and speeds up processing), otherwise roco can be supplied as a character vector of strings like 'R01C01' (only 3rd and 6th characters used).

## Value

a matrix of normalised beta values.

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## References

A data-driven approach to preprocessing Illumina 450K methylation array data, Pidsley et al, BMC Genomics.

interpolatedXY: a two-step strategy to normalise DNA methylation microarray data avoiding sex bias, Wang et al., 2021.

## **Examples**

```
data(melon)
normalised_betas <- adjustedDasen(mns = methylated(melon), uns = unmethylated(melon), onetwo = fData(melon)[,fot(
## if input is an object of methylumiset or methylset
normalised_betas <- adjustedDasen(melon)</pre>
```

adjustedFunnorm

adjustedFunnorm

## **Description**

adjustedFunnorm utilizes functional normliasation to normalise autosomal CpGs, and infers the sex chromosome linked CpGs by linear interpolation on corrected autosomal CpGs.

## Usage

```
adjustedFunnorm(
  rgSet,
  nPCs = 2,
  sex = NULL,
  bgCorr = TRUE,
  dyeCorr = TRUE,
  keepCN = TRUE,
  ratioConvert = TRUE,
  verbose = TRUE
```

rgSet	An object of class "RGChannelSet".
nPCs	Number of principal components from the control probes PCA.
sex	An optional numeric vector containing the sex of the samples.
bgCorr	Should the NOOB background correction be done, prior to functional normalization (see "preprocessNoob")
dyeCorr	Should dye normalization be done as part of the NOOB background correction (see "preprocessNoob")?
keepCN	Should copy number estimates be kept around? Setting to 'FALSE' will decrease the size of the output object significantly.

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ratioConvert Should we run "ratioConvert", ie. should the output be a "GenomicRatioSet" or

should it be kept as a "GenomicMethylSet"; the latter is for experts.

verbose Should the function be verbose?

#### Value

an object of class "GenomicRatioSet", unless "ratioConvert=FALSE" in which case an object of class "GenomicMethylSet".

## References

Functional normalization of 450k methylation array data improves replication in large cancer studies, Fortin et al., 2014, Genome biology.

interpolatedXY: a two-step strategy to normalise DNA methylation microarray data avoiding sex bias, Wang et al., 2021.

## **Examples**

```
## Not run:
GRset <- adjustedFunnorm(RGSet)
## End(Not run)</pre>
```

agep

Age Prediction from methylomic expression data

## Description

Predict age of samples using Horvaths Coefficients

#### **Usage**

```
agep(betas, coeff = NULL, method = c('horvath', 'hannum', 'phenoage', 'skinblood', 'lin', 'all'), n_mis
```

betas	Matrix of betas or MethyLumiSet or MethylSet object.
coeff	If NULL, will default to whatever method is specified in method. If not NULL, the expected input should be a vector of coefficients and intercept
method	Currently: "horvath", "hannum", "phenoage", "skinblood", "lin" and "all", if "all" agep will seek to calculate ages using all methods else will use the method specified. Default is "horvath".
n_missing	Logical, additionally output the number of missing CpGs for each sample using the specified method or coeff list.
missing_probes	Logical, additionally output the names of missing CpGs for each sample using the specified method or coeff list.
	To pass to arguments to downstream functions to specify adult.age

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## Value

Returns matrix of predicted ages per sample. With additional columns created whether n\_missing or missing\_probes are specified. If method is "all" then all ages will be provided in the same matrix output

## Author(s)

Original Functions: Steve Horvath

wateRmelon Implementation: Tyler Gorrie-Stone, Leo Schalkwyk, Louis El Khoury

#### References

Horvath S: DNA methylation age of human tissues and cell types. Genome Biology 2013, 14:R115

#### **Examples**

```
data(melon)
agep(melon,coeff=NULL, method="all", n_missing=FALSE)
agep(melon,coeff=NULL, method="horvath", n_missing=TRUE)
```

as.methylumi-methods

Methods for Function as.methylumi

## **Description**

Returns a MethyLumiSet object populaed with the data provided. There are MethyLumiSet and MethylSet methods. In the default method, the data is all optional. Please note that for the results to be sane, mn, un, bn, and pv have to be in the same sample and feature order and the same size. The function does not currently do any checks!

## Usage

```
# default method as.methylumi (mn = NULL, un = NULL, bn = NULL, pv = NULL, qc = NULL, da = NULL,...)
```

mn	matrix of methylated signal intensities, each column representing a sample (generic) or a MethyLumiSet, RGSet, or MethylSet object. Column names are used to get Sentrix row and column by default, see ''.
un	matrix of unmethylated signal intensities, each column representing a sample (default method) or NULL when mn is an object containing methylated and unmethylated values
bn	matrix of precalculated betas, each column representing a sample
pv	matrix of detection p-values, each column representing a sample

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da	annotation data frame, such as $x@featureData@data \#methylumi package$ . If NULL (the default), the IlluminaHumanMethylation450kmanifest package is used. See the fd argument
qc	control probe intensities: list of 2 matrices, Cy3 and Cy5, with rownames, such as produced by intensitiesByChannel(QCdata(x)) (methylumi package)
	Other arguments such as a featureData object or optional assayData

#### Methods

signature(mn = "MethylSet") Coerces a MethylSet to a MethyLumiSet, and provides it with a
set of featureData, which by default is just the chromosome and DESIGN (ie typeI or type
II assay). Other data can be included using the fd argument, available data is listed by the
function getColumns()

signature(mn = "MethyLumiSet") This is mainly useful for adding featureData as described under MethylSet above. MethyLumiSet objects produced by methylumiR have the full annotation, those from methylumIDAT do not, and functions such as swan require it

signature(mn = "ANY") as.methylumi (mn = NULL, un = NULL, bn = NULL, pv = NULL, qc = NULL, da = NULL, fd = c("CHR", "DESIGN"))

beadc	Calculates the number of samples with bead count <3 for each probe
	in matrix of bead count values

## **Description**

Calculates the number of samples with bead count <3 for each probe in matrix of bead count values.

## Usage

beadc(x)

#### **Arguments**

x matrix of bead count values returned by the beadcount function

## Value

Vector of number of samples with bead count <3 for each probe

## Note

The beadc function is internal to the pfilter function

## Author(s)

ruth.pidsley@kcl.ac.uk

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## References

[1] Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

beadcount

Creates matrix of beacounts from minfi data.

## **Description**

Creates matrix of beacounts from data read in using the minfi package. NAs represent probes with beadcount <3. An Extended RG Channel Set is required for this function to work.

## Usage

beadcount(x)

## **Arguments**

Х

450K methylation data read in using minfi to create an Extended RG Channel Set

## Value

A matrix of bead counts with bead counts <3 represented by NA for use in the pfilter function for quality control

## Note

The beadcount function is internal to the pfilter function

## Author(s)

Ruth.Pidsley@kcl.ac.uk

#### References

[1] Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

Beta2M

Internal functions for peak.correction (fuks)

## **Description**

Internal functions for peak.correction

## Usage

Beta2M(B)

## **Arguments**

В

a vector or matrix of beta values for conversion

## Value

a vector or matrix of the same shape as the input

## Author(s)

Matthieu Defrance <defrance@bigre.ulb.ac.be>

## References

Dedeurwaerder S, Defrance M, Calonne E, Sotiriou C, Fuks F: Evaluation of the Infinium Methylation 450K technology . Epigenetics 2011, 3(6):771-784.

betaqn-exprmethy450-methods

Calculate normalized betas from exprmethy450 of Illumina 450K methylation arrays

## Description

Quantile normalize betas from exprmethy 450 objects

## **Arguments**

bn An exprmethy450 object.

fudge value added to total intensity to prevent denominators close to zero when calcu-

lating betas

## **Details**

betaqn quantile normalizes betas

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#### Value

exprmethy450 object of the same shape and order as bn.

#### Author(s)

Leonard.Schalkwyk@kcl.ac.uk

#### References

- [1] Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)
- [2] Dedeurwaerder S, Defrance M, Calonne E, Sotiriou C, Fuks F: Evaluation of the Infinium Methylation 450K technology. Epigenetics 2011, 3(6):771-784.
- [3] Touleimat N, Tost J: Complete pipeline for Infinium R Human Methylation 450K BeadChip data processing using subset quantile normalization for accurate DNA methylation estimation. Epigenomics 2012, 4:325-341
- [4] Maksimovic J, Gordon L, Oshlack A: SWAN: Subset quantile Within-Array Normalization for Illumina Infinium HumanMethylation450 BeadChips. Genome biology 2012, 13(6):R44

BMIQ Beta-Mixture Quantile (BMIQ) Normalisation method for Illumina 450k arrays

## Description

BMIQ is an intra-sample normalisation procedure, correcting the bias of type-2 probe values. BMIQ uses a 3-step procedure: (i) fitting of a 3-state beta mixture model, (ii) transformation of state-membership probabilities of type2 probes into quantiles of the type1 distribution, and (iii) a conformal transformation for the hemi-methylated probes. Exact details can be found in the reference below.

#### Usage

```
BMIQ(beta.v, design.v, nL = 3, doH = TRUE, nfit = 50000, th1.v = c(0.2, 0.75), th2.v = NULL, niter = 5, to ## S4 method for signature 'MethyLumiSet' BMIQ(beta.v, nL=3, doH=TRUE, nfit=5000, th1.v=c(0.2, 0.75), th2.v=NULL, niter=5, tol=0.001, plots=FALSE CheckBMIQ(beta.v, design.v, pnbeta.v)
```

## Arguments

beta.v vector consisting of beta-values for a given sample, or a MethyLumiSet or MethylSet containing multiple samples. For the MethyLumiSet and MethylSet methods, this is the only required argument, and the function will be run on each sample.

design.v corresponding vector specifying probe design type (1=type1,2=type2). This must be of the same length as beta.v and in the same order.

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number of states in beta mixture model. 3 by default. At present BMIQ only works for nL=3.
perform normalisation for hemimethylated type2 probes. These are normalised using an empirical conformal transformation and also includes the left-tailed type2 methylated probes since these are not well described by a beta distribution. By default TRUE.
number of probes of a given design type to use for the fitting. Default is 50000. Smaller values (~10000) will make BMIQ run faster at the expense of a small loss in accuracy. For most applications, 5000 or 10000 is ok.
thresholds used for the initialisation of the EM-algorithm, they should represent buest guesses for calling type1 probes hemi-methylated and methylated, and will be refined by the EM algorithm. Default values work well in most cases.
thresholds used for the initialisation of the EM-algorithm, they should represent buest guesses for calling type2 probes hemi-methylated and methylated, and will be refined by the EM algorithm. By default this is null, and the thresholds are estimated based on th1.v and a modified PBC correction method.
maximum number of EM iterations to do. This number should be large enough to yield good fits to the type1 distribution. By default 5.
tolerance convergence threshold for EM algorithm. By default 0.001.
logical specifying whether to plot the fits and normalised profiles out. By default TRUE.
the ID of the sample being normalised.
logical: print verbose progress information?
BMIQ normalised profile.

## **Details**

Full details can be found in the reference below. Note: these functions require the RPMM package, not currently a dependency of the wateRmelon package.

## Value

Default method: A list with following entries:

nbeta	the normalised beta-profile for the sample
class1	the assigned methylation state of type1 probes
class2	the assigned methylation state of type2 probes
av1	the mean beta-values for the nL states for type1 probes
av2	the mean beta-values for the nL states for type2 probes
hf	the estimated "Hubble" dilation factor used in the normalisation of hemi-methylated probes $$
th1	estimated thresholds for calling unmethylated and methylated type1 probes
th2	estimated thresholds for calling unmethylated and methylated type2 probes

MethyLumiSet method: A methyLumiSet object

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#### Author(s)

Andrew Teschendorff, MethyLumiSet method by Leo Schalkwyk Leonard.Schalkwyk@kcl.ac.uk

#### References

Teschendorff AE, Marabita F, Lechner M, Bartlett T, Tegner J, Gomez-Cabrero D, Beck S. A Beta-Mixture Quantile Normalisation method for correcting probe design bias in Illumina Infinium 450k DNA methylation data. Bioinformatics. 2012 Nov 21.

## **Examples**

- # library(RPMM)
- # data(melon)
- # BMIQ(melon,nfit=100)

bscon

Calculate bisulphite conversion

## Description

Uses control data from Infinium HumanMethylation450 BeadChip to calculate bisulfite conversion for each array

#### **Usage**

```
bscon(x, ...) # S4 methods exist for RGChannelSet and MethyLumiset objects
```

#### **Arguments**

x IDAT or report files containing 450k data
... current methods have no optional arguments

## **Details**

This function uses the green and red channels reading of the type I and type II bisilfite conversion data to return the median bisulfite conversion percentage value for each array.

For the type I chemistry the beta values are calculated by dividing the first three probes of the green channel (C1, C2, C3) and the second three probes of the red channel (C4, C5, C6) by the sum of these probes and the unconverted probes of the green (U1, U2, U3) and the red (U4, U5, U6) channel.

The beta values from type II chemistry are calculated by dividing the methylated (red) channels by the sum of methylated (red) and unmethylated (green) channels.

## Value

A vector of percentage values referring to the bisulfite conversion levels of each array.

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## Note

Updates to HumanMethylationEPIC manifest has seen the removal of control probes C6 and U6. This does not appear to grossly affect how function performs however we are considering alternative approaches to account for this.

## Author(s)

Louis El Khoury (louis.el-khoury@essex.ac.uk), Eilis Hannon, Leonard Schalkwyk (lschal@essex.ac.uk)

## **Examples**

```
library(wateRmelon)
data(melon)
bs <- bscon(melon)
bs</pre>
```

canno

canno - process csv manifest into annotation object for illumina methylation preprocessing

## **Description**

canno - process csv manifest into annotation object for illumina methylation preprocessing

## Usage

```
canno(man = "EPIC-8v2-0_A1.csv", name = NULL)
```

## **Arguments**

man

name

## **Details**

This is based on the scripts shipped with minfi annotation packages. It is based on existing Illumina Human Methylation csv format manifests, but because reverse-engineered, may require updates to work on future products.

## Value

IlluminaMethylationManifest

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colnames-methods

Methods for Function colnames in Package wateRmelon

## **Description**

Methods for function colnames in package wateRmelon.

#### **Methods:**

```
signature(x = "MethyLumiSet") returns the sample names
```

combo

Combine MethyLumiSet objects

## **Description**

This is a wrapper for combining different MethyLumiSet objects.

## Usage

```
combo(...)
```

## **Arguments**

Eventually, any number of MethyLumiSet objects. Currently only guaranteed for 2 objects.

## **Details**

This is a wrapper for methylumi::combine, which works around a name clash with a different combine function from the gdata package, and also a bug in methylumi::combine.

## Value

a MethyLumiSet. The assayData, QCdata, experimentData, protocolData and phenoData are joined on sampleName . featureData and annotation are taken from the object given in the first argument

## Note

the function uses sampleNames and gets rid of duplicates. Numeric sampleNames cause problems (and are a Bad Idea anyway). They should be turned into names with make.names() first.

## Author(s)

Leo Schalkwyk <leonard.schalkwyk@kcl.ac.uk>

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#### References

[1] Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

## See Also

```
as.methylumi
```

## **Examples**

```
library(wateRmelon)
data(melon)
## pretend we have two different data sets
melon
pelon <- melon
sampleNames(pelon) <- gsub('^6', 7, sampleNames(pelon))
combo(melon, pelon)</pre>
```

dasen

Calculate normalized betas from Illumina 450K methylation arrays

## **Description**

Multiple ways of calculating the index of methylation (beta) from methylated and unmethylated probe intensities used in Pidsley et al 2012. S4 methods exist where possible for MethyLumiSet, MethylSet, RGSet and exprmethy450 objects.

## Usage

```
dasen ( mns, uns, onetwo, fudge = 100, ret2=FALSE, ... )
nasen ( mns, uns, onetwo, ret2=FALSE, fudge = 100, ... )
betaqn( bn )
naten ( mn, un, fudge = 100, ret2=FALSE, ... )
naten ( mn, un, fudge = 100, ret2=FALSE, ... )
            un, fudge = 100, ret2=FALSE, ...)
nanet ( mn,
nanes ( mns, uns, onetwo, fudge = 100, ret2=FALSE, ... )
danes ( mn,
            un, onetwo, fudge = 100, ret2=FALSE, ...)
                 onetwo, fudge = 100, ret2=FALSE, ...)
danet ( mn,
            un,
danen ( mns, uns, onetwo, fudge = 100, ret2=FALSE, ... )
            un, onetwo, fudge = 100, ret2=FALSE, ...)
daten1( mn,
daten2( mn,
            un,
                 onetwo, fudge = 100, ret2=FALSE, ...)
                 da, pn )
tost ( mn,
            un,
fuks (data, anno)
swan ( mn, un, qc, da=NULL, return.MethylSet=FALSE )
```

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## **Arguments**

mn, mns matrix of methylated signal intensities, each column representing a sample (generic)

or a MethyLumiSet, RGSet, or MethylSet object. Column names are used to get

Sentrix row and column by default, see '...'.

un, uns matrix of unmethylated signal intensities, each column representing a sample

(default method) or NULL when mn is an object containing methylated and

unmethylated values

bn, data matrix of precalculated betas, each column representing a sample

onetwo character vector or factor of length nrow(mn) indicating assay type 'I' or 'II'

pn matrix of detection p-values, each column representing a sample

da, anno annotation data frame, such as x@featureData@data #methylumi package. If

NULL, the swan method requires the IlluminaHumanMethylation450kmanifest

package.

qc control probe intensities: list of 2 matrices, Cy3 and Cy5, with rownames, such

as produced by intensitiesByChannel(QCdata(x)) #methylumi package

fudge value added to total intensity to prevent denominators close to zero when calcu-

lating betas

return.MethylSet

if TRUE, returns a MethylSet object instead of a naked matrix of betas.

ret2 if TRUE, returns a list of intensities and betas instead of a naked matrix of betas.

additional argument roco for dfsfit giving Sentrix rows and columns. This allows

a background gradient model to be fit. This is split from data column names by default. roco=NULL disables model fitting (and speeds up processing), otherwise roco can be supplied as a character vector of strings like 'R01C01' (only

3rd and 6th characters used).

#### **Details**

dasen same as nasen but type I and type II backgrounds are equalized first. This is our recommended method

**betaqn** quantile normalizes betas

**naten** quantile normalizes methylated and unmethylated intensities separately, then calculates betas

**nanet** quantile normalizes methylated and unmethylated intensities together, then calculates betas. This should equalize dye bias

**nanes** quantile normalizes methylated and unmethylated intensities separately, except for type II probes where methylated and unmethylated are normalized together. This should equalize dye bias without affecting type I probes which are not susceptible

danes same as nanes, except type I and type II background are equalized first

danet same as nanet, except type I and type II background are equalized first

danen background equalization only, no normalization

**daten1** same as naten, except type I and type II background are equalized first (smoothed only for methylated)

20 dasen

**daten2** same as naten, except type I and type II background are equalized first (smoothed for methylated an unmethylated)

nasen same as naten but type I and typeII intensities quantile normalized separately

tost method from Touleimat and Tost 2011

**fuks** method from Dedeurwaerder et al 2011. Peak correction only, no normalization **swan** method from Maksimovic et al 2012

#### Value

a matrix (default method) or object of the same shape and order as the first argument containing betas.

## Author(s)

Leonard.Schalkwyk@kcl.ac.uk

#### References

- [1] Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)
- [2] Dedeurwaerder S, Defrance M, Calonne E, Sotiriou C, Fuks F: Evaluation of the Infinium Methylation 450K technology . Epigenetics 2011, 3(6):771-784.
- [3] Touleimat N, Tost J: Complete pipeline for Infinium R Human Methylation 450K BeadChip data processing using subset quantile normalization for accurate DNA methylation estimation. Epigenomics 2012, 4:325-341.
- [4] Maksimovic J, Gordon L, Oshlack A: SWAN: Subset quantile Within-Array Normalization for Illumina Infinium HumanMethylation450 BeadChips. Genome biology 2012, 13(6):R44

## See Also

```
pfilter, as.methylumi
```

## **Examples**

```
#MethyLumiSet method
data(melon)
melon.dasen <- dasen(melon)</pre>
```

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dasen-methods Calculate normalized betas from MethyLumiSets of Illumina 450 methylation arrays
--

## **Description**

Multiple ways of calculating the index of methylation (beta) from methylated and unmethylated probe intensities used in Pidsley et al 2012.

## **Arguments**

mn, mns, data, bn A MethyLumiSet object. Sample names names are used to get Sentrix row and

column by default, see '...'.

fudge value added to total intensity to prevent denominators close to zero when calcu-

lating betas

additional argument roco for dfsfit giving Sentrix rows and columns. This allows

a background gradient model to be fit. This is split from data column names by default. roco=NULL disables model fitting (and speeds up processing), otherwise roco can be supplied as a character vector of strings like 'R01C01' (only

3rd and 6th characters used).

#### **Details**

dasen same as nasen but type I and type II backgrounds are normalized first. This is our recommended method

**betagn** quantile normalizes betas

**naten** quantile normalizes methylated and unmethylated intensities separately, then calculates betas **nanet** quantile normalizes methylated and unmethylated intensities together, then calculates betas. This should equalize dye bias.

**nanes** quantile normalizes methylated and unmethylated intensities separately, except for type II probes where methylated and unmethylated are normalized together. This should equalize dye bias without affecting type I probes which are not susceptible.

danes same as nanes, except typeI and type II background are equalised first.

danet same as nanet, except typeI and type II background are equalised first.

danen background equalisation only, no normalization

**daten1** same as naten, except typeI and type II background are equalised first (smoothed only for methylated)

**daten2** same as naten, except typeI and type II background are equalised first (smoothed for methylated an unmethylated)

nasen same as naten but typeI and typeII intensities quantile normalized separately

tost method from Touleimat and Tost 2011

**fuks** method from Dedeurwaerder et al 2011. Peak correction only, no normalization

swan method from Maksimovic et al 2012

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#### Value

a matrix (default method) or object of the same shape and order as the first argument containing betas.

#### methods

```
dasen ( mns, fudge = 100, ... ) nasen ( mns, fudge = 100 ) betaqn( bn ) naten ( mn, fudge = 100 ) naten ( mn, fudge = 100 ) nanes ( mns, fudge = 100 ) nanes ( mns, fudge = 100 ) danes ( mn, fudge = 100, ... ) danet ( mn, fudge = 100, ... ) danen ( mns, fudge = 100, ... ) daten1( mn, fudge = 100, ... ) daten2( mn, fudge = 100, ... ) tost ( mn ) fuks ( data) swan ( mn )
```

## Author(s)

Leonard.Schalkwyk@kcl.ac.uk

#### References

- [1] Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)
- [2] Dedeurwaerder S, Defrance M, Calonne E, Sotiriou C, Fuks F: Evaluation of the Infinium Methylation 450K technology . Epigenetics 2011, 3(6):771-784.
- [3] Touleimat N, Tost J: Complete pipeline for Infinium R Human Methylation 450K BeadChip data processing using subset quantile normalization for accurate DNA methylation estimation. Epigenomics 2012, 4:325-341
- [4] Maksimovic J, Gordon L, Oshlack A: SWAN: Subset quantile Within-Array Normalization for Illumina Infinium HumanMethylation450 BeadChips. Genome biology 2012, 13(6):R44

dasen-minfi-methods

Calculate normalized betas from Illumina 450K methylation arrays

## Description

Multiple ways of calculating the index of methylation (beta) from methylated and unmethylated probe intensities used in Pidsley et al 2012.

mn, mns	matrix of methylated signal intensities, each column representing a sample (default method), or an object for which a method is available. Column names are used to get Sentrix row and column by default, see ''.
un, uns	matrix of unmethylated signal intensities, each column representing a sample (default method) or NULL when mn is an object containing methylated and unmethylated values
bn, data	matrix of precalculated betas, each column representing a sample
onetwo	character vector or factor of length nrow(mn) indicating assay type 'I' or 'II'

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da, anno annotation data frame, such as x@featureData@data #methylumi package

control probe intensities: list of 2 matrices, Cy3 and Cy5, with rownames, such as produced by intensitiesByChannel(QCdata(x)) #methylumi package

fudge value added to total intensity to prevent denominators close to zero when calculating betas

... additional argument roco for dfsfit giving Sentrix rows and columns. This allows a background gradient model to be fit. This is split from data column names by default. roco=NULL disables model fitting (and speeds up processing), otherwise roco can be supplied as a character vector of strings like 'R01C01' (only

#### **Details**

dasen same as nasen but type I and type II backgrounds are normalized first. This is our recommended method

betaqn quantile normalizes betas

naten quantile normalizes methylated and unmethylated intensities separately, then calculates betasnanet quantile normalizes methylated and unmethylated intensities together, then calculates betas.This should equalize dye bias.

**nanes** quantile normalizes methylated and unmethylated intensities separately, except for type II probes where methylated and unmethylated are normalized together. This should equalize dye bias without affecting type I probes which are not susceptible.

danes same as nanes, except typeI and type II background are equalised first.

3rd and 6th characters used).

danet same as nanet, except typeI and type II background are equalised first.

danen background equalisation only, no normalization

**daten1** same as naten, except typeI and type II background are equalised first (smoothed only for methylated)

**daten2** same as naten, except typeI and type II background are equalised first (smoothed for methylated an unmethylated)

nasen same as naten but typeI and typeII intensities quantile normalized separately

tost method from Touleimat and Tost 2011

**fuks** method from Dedeurwaerder et al 2011. Peak correction only, no normalization **swan** method from Maksimovic et al 2012

## Value

a matrix of betas is returned by the MethySet and RGChannelSet methods because they do not have a defined slot for betas.

## methods

dasen ( mns, uns, onetwo, fudge = 100, ... ) nasen ( mns, uns, onetwo, fudge = 100 ) betaqn( bn ) naten ( mn, un, fudge = 100 ) naten ( mn, un, fudge = 100 ) nanet ( mn, un, fudge = 100 ) danes ( mns, uns, onetwo, fudge = 100 ) danes ( mn, un, onetwo, fudge = 100, ... ) danet (

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mn, un, onetwo, fudge = 100, ... ) danen ( mns, uns, onetwo, fudge = 100, ... ) daten1( mn, un, onetwo, fudge = 100, ... ) tost ( mn, un, da, pn ) fuks ( data, anno) swan ( mn, un, qc )

#### Author(s)

Leonard.Schalkwyk@kcl.ac.uk

#### References

- [1] Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)
- [2] Dedeurwaerder S, Defrance M, Calonne E, Sotiriou C, Fuks F: Evaluation of the Infinium Methylation 450K technology . Epigenetics 2011, 3(6):771-784.
- [3] Touleimat N, Tost J: Complete pipeline for Infinium R Human Methylation 450K BeadChip data processing using subset quantile normalization for accurate DNA methylation estimation. Epigenomics 2012, 4:325-341)
- [4] Maksimovic J, Gordon L, Oshlack A: SWAN: Subset quantile Within-Array Normalization for Illumina Infinium HumanMethylation450 BeadChips. Genome biology 2012, 13(6):R44

db1

Internal wateRmelon functions for calculating betas

## Description

db1 is used for quantile normalizing methylated together with unmethylated (dye bias methods nanet, nanes, danes and danet. dfs\* functions are used for smoothing the background equalization in methods whose names start with d (daten etc).

## Usage

```
db1(mn, un)
dfsfit(mn, onetwo, roco=substring(colnames(mn), regexpr("R0[1-9]C0[1-9]", colnames(mn))), ...)
dfs2(x, onetwo)
```

mn, x	matrix of methylated signal intensities, each column representing a sample (default method), or an object for which a method is available. For dfsfit and dfs2 this can also be a matrix of unmethylated intensities.
un	matrix of unmethylated signal intensities, each column representing a sample (default method) or NULL when mn is an object containing methylated and unmethylated values.
onetwo	character vector or factor of length nrow(mn) indicating assay type 'I' or 'II'

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roco roco for dfsfit giving Sentrix rows and columns. This allows a background gradi-

ent model to be fit. This is split from data column names by default. roco=NULL disables model fitting (and speeds up processing), otherwise roco can be supplied as a character vector of strings like 'R01C01' (3rd and 6th characters used).

... no additional arguments currently used

#### **Details**

db1 - quantile normalizes methylated against unmethylated (basic function for dyebuy\* dye bias methods). dfsfit - corrects the difference in backgrounds between type I and type II assays and fits a linear model to Sentrix rows and columns if these are available to improve precision where there is a background gradient. dfs2 - finds the difference between type I and type II assay backgrounds for one or more samples.

#### Value

db1 - a list of 2 matrices of intensities, methylated and unmethylated dfsfit - a matrix of adjusted intensities dfs2 - a background offset value

## Author(s)

Leonard.Schalkwyk@kcl.ac.uk

#### References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

dmrse

Standard error of iDMR 450k array DNA methylation features

## Description

Imprinting differentially methylated regions (iDMRs) are expected to be approximately half methylated, as is observed at the 227 probes in known iDMRs. These functions calculate measures of dispersion for the beta values at these CpG sites, of which the most useful is dmrse\_row, which is the between-sample standard error.

## Usage

```
dmrse(betas, idmr = iDMR())
dmrse_col(betas, idmr = iDMR())
dmrse_row(betas, idmr = iDMR())
```

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## Arguments

betas a matrix of betas (default method), a MethyLumiSet object (methylumi pack-

age), a MethylSet or RGChannelSet object (minfi package) or a exprmethy450

object (IMA package).

idmr a character vector of iDMR probe names such as returned by iDMR()

#### Value

return a standard error of the mean of betas for all samples and iDMR probes (dmrse) or the standard error of the mean for just the between sample component(dmrse\_row) or between probe(dmrse\_col) component.

## Author(s)

Leonard.Schalkwyk@kcl.ac.uk

#### References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

#### See Also

seabi, a sex-difference metric, and genki, based on SNPs.

## **Examples**

```
#MethyLumiSet method
  data(melon)
  dmrse(melon)

#MethyLumiSet method after normalization
  melon.dasen <- dasen(melon)
  dmrse(melon.dasen)</pre>
```

dmrse-methods

 $Methods\ for\ Function\ dmrse\ in\ Package\ wateRmelon$ 

## **Description**

Methods for function dmrse, dmrse\_row and dmrse\_col in package **wateRmelon**. Please see dmrse for details of the calculation of this standard-error performance metric.

#### **Methods:**

signature(betas = "exprmethy450") all of the methods simply extract betas from the data object (which can be a exprmethy450, MethylSet, MethylLumiSe, or RGChannelSet) and calculate the metric.

epicv2clean.default 27

epicv2clean.default

Strip and subset EPICv2 data to work with legacy data and methods

## **Description**

Returns an object with rownames stripped of the EPICv2 suffixes, duplicate probes are omitted.

## Usage

```
## Default S3 method:
epicv2clean(x)
```

#### **Details**

EPICv2 manifests contain a few thousand probes with up to 10 replicate syntheses. To accommodate this a modified naming scheme is used, so none of the probe names match those on the EPIC and previous arrays (even though most of the probes are the same sequence and presumably similar performance).

This simple function relies on the rowname and subsetting methods and will work for matrix, dataframe, MethyLumiSet, or MethylSet objects, and there is a method for gds (bigmelon) objects.

estimateCellCounts

Cell Proportion Estimation using wateRmelon

## **Description**

Estimates relative proportion of pure cell types within a sample, mostly identical to estimateCellCounts. References for both 450k and EPIC array are available. However 450k reference can be used on EPIC data by specifying the reference platform. Additionally a measure of error is calculated as a means of quality control.

## Usage

```
estimateCellCounts.wmln(
   object,
   referencePlatform = c("IlluminaHumanMethylation450k",
        "IlluminaHumanMethylationEPIC",
        "IlluminaHumanMethylation27k"),
   mn = NULL,
   un = NULL,
   bn = NULL,
   perc = 1,
   compositeCellType = "Blood",
   probeSelect = "auto",
   cellTypes = c("CD8T","CD4T","NK","Bcell","Mono","Gran"),
```

28 estimateCellCounts

```
returnAll = FALSE,
meanPlot = FALSE,
verbose=TRUE,
...)
```

#### **Arguments**

object An object of class methylumiset, which contains (un)normalised methylated and

unmethylated intensities

mn if NULL will call methylated(object), otherwise can be given matrix of identical

dimensions to object.

un if NULL will call unmethylated(object), otherwise can be given matrix of iden-

tical dimensions to object.

bn if NULL will call betas(object), otherwise can be given matrix of identical di-

mensions to object.

perc Percentage of query-samples to use to normalise reference dataset. This should

be 1 unless using a very large data-set then lowering this will allow for an in-

crease in performance

compositeCellType

Which composite cell type is being deconvoluted. Should be either "Blood",

"CordBlood", or "DLPFC"

probeSelect How should probes be selected to distinguish cell types? Options include "both",

which selects an equal number (50) of probes (with F-stat p-value < 1E-8) with the greatest magnitude of effect from the hyper- and hypo-methylated sides, and "any", which selects the 100 probes (with F-stat p-value < 1E-8) with the greatest magnitude of difference regardless of direction of effect. Default input "auto" will use "any" for cord blood and "both" otherwise, in line with previous versions of this function and/or our recommendations. Please see the references

for more details.

cellTypes Which cell types, from the reference object, should be we use for the deconvo-

lution? See details.

referencePlatform

The platform for the reference dataset; if the input rgSet belongs to another

platform, it will be converted using convertArray.

returnAll Should the composition table and the normalized user supplied data be return?

verbose Should the function be verbose?

meanPlot Whether to plots the average DNA methylation across the cell-type discrimating

probes within the mixed and sorted samples.

... Other arguments, i.e arguments passed to plots

## **Details**

See estimateCellCounts for more information regarding the exact details. estimateCellCounts.wmln differs slightly, as it will impose the quantiles of type I and II probes onto the reference Dataset rather than normalising the two together. This is 1) More memory efficient and 2) Faster - due to

estimateSex 29

not having to normalise out a very small effect the other 60 samples from the reference set will have on the remaining quantiles.

Optionally, a proportion of samples can be used to derive quantiles when there are more than 1000 samples in a dataset, this will further increase performance of the code at a cost of precision. If data is pre-normalised a minimum of two samples are required.

estimateSex

Predict sex by using robust sex-related CpG sites on ChrX and ChrY

## **Description**

Predict sex by using robust sex-related CpG sites on ChrX and ChrY

#### Usage

```
estimateSex(betas, do_plot = FALSE)
```

#### **Arguments**

betas A matrix with sample IDs as column names, and probe names as row names,

ideally: beta = M / (M + U + 100). Take a look at an example betas with:

"data(melon); print(betas(melon)[1:10, 1:3])".

do\_plot logical. Should plot the predicted results? Default: FALSE

## Value

dataframe contains predicted sex information.

## Author(s)

Wang, Yucheng, et al. "DNA methylation-based sex classifier to predict sex and identify sex chromosome aneuploidy." BMC genomics 22.1 (2021): 1-11.

## **Examples**

```
data(melon)
pred_XY <- estimateSex(betas(melon), do_plot=TRUE)</pre>
```

30 genki

genki	SNP derived performance metrics for Illumina 450K DNA methylation
	arrays.

## **Description**

A very simple genotype calling by one-dimensional K-means clustering is performed on each SNP, and for those SNPs where there are three genotypes, the squared deviations are summed for each genotype (similar to a standard deviation for each of allele A homozygote, heterozygote and allele B homozygote). By default these are further divided by the square root of the number of samples to get a standard error-like statistic.

## Usage

```
genki(bn, g = getsnp(rownames(bn)), se = TRUE)
```

## **Arguments**

bn	a matrix of beta values(default method), a MethyLumiSet object (methylumi package), a MethylSet or RGChannelSet object (minfi package) or a exprmethy450 object (IMA package).
g	vector of SNP names
se	TRUE or FALSE specifies whether to calculate the standard error-like statistic

## **Details**

There are 65 well-behaved SNP genotyping probes included on the array. These each produce a distribution of betas with tight peaks for the three possible genotypes, which will be broadened by technical variation between samples. The spread of the peaks is thus usable as a performance metric.

#### Value

a vector of 3 values for the dispersion of the three genotype peaks (AA, AB, BB : low, medium and high beta values)

## Note

Corrected RGChannelSet methods - 12/10/2015

#### Author(s)

Leonard.Schalkwyk@kcl.ac.uk

## References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

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## **Examples**

```
#MethyLumiSet method
  data(melon)
  genki(melon)

#MethyLumiSet method after normalization
  melon.dasen <- dasen(melon)
  genki(melon.dasen)</pre>
```

genki-methods

Methods for Function genki in Package wateRmelon

## **Description**

Methods for function genki in package **wateRmelon**. Please see genki for details of the calculation of this standard-error performance metric.

## **Methods:**

signature(betas = "exprmethy450") all of the methods simply extract betas from the data object (which can be a exprmethy450, MethylSet, MethylLumiSe, or RGChannelSet) and calculate the metric.

genkme

Internal functions for genotype-based normalization metrics

## **Description**

```
genkme - genotype calling with 1d k-means
genkus - apply genkme to available SNPs
getsnp - grep the rs-numbered probes
gcose - calculate between-sample SNP standard error
gcoms - calculate between-sample SNP mean-squared deviation
```

## Usage

```
genkme(y, peaks = c(0.2, 0.5, 0.8))
```

```
y a vector or matrix of numeric values (betas, between 0 and 1) peaks initial values for cluster positions
```

32 got

## **Details**

see genki

#### Value

see genki

## Author(s)

Leonard.Schalkwyk@kcl.ac.uk

#### References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

got

Internal functions for Illumina i450 normalization functions

## Description

got and fot find the annotation column differentiating type I and type II assays in MethylSet (got) or MethyLumiSet (fot) objects. pop extracts columns from IlluminaHumanMethylation450k.db

## Usage

```
got(obj)
fot(x)
```

## **Arguments**

```
x a MethyLumiSet obj a MethylSet
```

## **Details**

got returns a character vector of 'I' and 'II', fot returns the index of the relevant column. pop returns a data frame

## Author(s)

Leonard.Schalkwyk@kcl.ac.uk

## References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

idet 33

idet

idet - identify idats by a hash of the addresses

## **Description**

idet - identify idats by a hash of the addresses

## Usage

idet(idat)

## **Arguments**

idat

can be an idat file or the list produced by reading one with readIDAT()

#### **Details**

this function is a response to the fact that IlluminaHumanMethylationEPIC and EPICv2 idats both have the ChipType "BeadChip 8x5" but different manifests. They did have different numbers of addresses. Subsequently we have had some confusion.... This hash (certainly taken together with the ChipType) should be bomb proof as an identifier. Warning: this is slow.

## Value

three strings: ChipType, an md5 hash of the MidBlock (address vector) and if known, the annotation name

iDMR

Imprinting differentially methylated region probes of Illumina 450 arrays

## Description

A character vector of 227 propes on the Illumina 450k methylation array

## Usage

data(iDMR)

#### **Format**

The format is: chr [1:227] "cg00000029" "cg00155882" "cg00576435" "cg00702231" "cg00765653" "cg00766368" ...

## Source

DMR coordinates from https://atlas.genetics.kcl.ac.uk/

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## References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

## **Examples**

```
data(iDMR)
## maybe str(iDMR) ; plot(iDMR) ...
```

melon

Small MethyLumi data set for examples and testing

## **Description**

This object was derived using methylumiR on an edited GenomeStudio file containing a small subset of features. It works with all of the wateRmelon package beta functions (see dasen and metrics (see genki, seabi, and dmrse\_col) except for swan.

## Usage

```
data(melon)
```

#### **Format**

MethyLumiSet with assayData containing 3363 features, 12 samples

## **Source**

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

## **Examples**

```
library(methylumi)
data(melon)
boxplot(log(methylated(melon)), las=2)
## maybe str(melon) ; plot(melon) ...
```

metrics 35

metrics (	Calculate a full set of 450K normalization/performance metrics
-----------	--

## Description

Calculate X-chromosome, SNP and imprinting DMR metrics for a matrix of betas from an Illumina 450K Human DNA methylation array. Requires precalculated t-test p-values for sex differences, a list of X-chromosome features and of imprinting DMR features.

## Usage

```
metrics(betas, pv, X, idmr = iDMR, subset = NULL)
```

## **Arguments**

betas	a matrix of betas, each row representing a probe, each column a sample
pv	a vector of p-values such as produced by sextest, one per row of betas
X	a logical vector of the same length as pv, indicating whether each probe is mapped to the X-chromosome
idmr	a character vector of probe names known to be in imprinting DMRs. Can be obtained with iDMR() or data(iDMR)
subset	index or character vector giving a subset of betas to be tested

## Value

dmrse_row	see dmrse_rov
dmrse_col	see dmrse_co
dmrse	see dmrse
gcoms_a	see genki
gcose_a	see genki
gcoms_b	see genki
gcose_b	see genki
gcoms_c	see genki
gcose_c	see genki
seabird	see seabi

## Author(s)

Leonard.Schalkwyk@kcl.ac.uk

## References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

## **Examples**

```
data(melon)
melon.dasen <- dasen(melon)
bn <-betas(melon.dasen)
X <- melon.dasen@featureData@data$CHR=='X'
data(iDMR)
sex <- pData(melon.dasen)$sex
pv <- sextest(bn,sex)
melon.metrics <- metrics(bn, pv, X, idmr = iDMR, subset = NULL)</pre>
```

## NChannelSetToMethyLumiSet2

For internal use, is read using minfi-like machinery and then preprocessed into the more flexible and convenient methylumi object used by wateRmelon/bigmelon

## **Description**

For internal use, is read using minfi-like machinery and then preprocessed into the more flexible and convenient methylumi object used by wateRmelon/bigmelon

## Usage

```
NChannelSetToMethyLumiSet2(
   NChannelSet,
   parallel = F,
   pval = 0.05,
   n = F,
   n.sd = F,
   oob = T,
   to = TRUE
)
```

## **Arguments**

NChannelSet an NChannelSet (raw red and green values not yet mapped to Illumina IDs/CpG names

parallel no effect, included for future parallelisation

pval detection pval threshold for filtering. Inactivated.

n keep nbeads data (min of m & u)

n.sd process SD of U and M (not currently implemented)

oob keep out-of-band signals

to does chip have type I and II probes?

## Value

A methylumi object with betas, U and M, optionally additional data

outlyx 37

outlyx	Identify Outliers within Methylumi and Minfi packaged objects

# Description

Seeks to identify outliers based on multiple (currently 2) outlier detection methods for methylumi and minfi packaged objects.

# Usage

## **Arguments**

X	A MethyLumiSet, MethylSet, RGChannelSet object or matrix containing raw betas.
iqr	If TRUE, the outliers based on interquartile ranges will be determined
iqrP	The number of interquartile ranges outliers are to be identified from designated principle component.
pc	Desired principal component for outlier identification - only used if other principal components want to be discriminated, only used for IQR outlier detection.
mv	If TRUE, the outliers will detected using pcout
m∨P	Arbitrary cut-off point for identifying outliers via pcout
plot	If TRUE, alongside regular output, a plot will be constructed displaying relative 'location' of each sample. Outliers are those that fall within the highlighted regions.
	Additional arguments passed to pcout

# Value

Returns a dataframe of TRUE/FALSE per sample where TRUE is outlying. Dataframe contains 3 columns, the first column (iqr) denotes samples which are outlying according to IQR on Principal component 1, the second column (mv) denotes outliers according to mahalanobis distances. And the third column (outliers) denotes samples that are TRUE in the first two columns.

## Note

May perform poorly on normalized data

## Author(s)

Tyler Gorrie-Stone - tgorri@essex.ac.uk

38 pfilter

#### **Examples**

```
library(wateRmelon)
data(melon)
outliers <- outlyx(melon,igr=TRUE, igrP=2, pc=1,
                   mv=TRUE, mvP=0.15, plot=TRUE)
```

outlyx-methods

Methods for Function outlyx in Package wateRmelon

## **Description**

Methods for function outlyx, please see outlyx for details of how function performs.

#### Methods

signature(x = "MethyLumiSet") all of the methods simply extract betas from the data object (which can be a MethylSet, MethyLumiSet, or RGChannelSet) and calculates the outliers.

pfilter

Basic data filtering for Illumina 450 methylation data

## **Description**

The pfilter function filters data sets based on bead count and detection p-values. The user can set their own thresholds or use the default pfilter settings. pfilter will take data matrices of beta values, signal intensities and annotation data, but will also take methylumi (MethyLumiSet) or minfi (RGChannelSetExtended) objects. However it has come to our attention that data read in using the various packages and input methods will give subtly variable data output as they calculate detection p-value and beta values differently, and do/don?t give information about beadcount. The pfilter function does not correct for this, but simply uses the detection p-value and bead count provided by each package.

#### **Usage**

```
pfilter(mn, un, bn, da, pn, bc, perCount=NULL, pnthresh = NULL, perc = NULL,
pthresh = NULL,logical.return=FALSE)
```

# **Arguments**

mn

matrix of methylated signal intensities, each column representing a sample (default method), or an object for which a method is available e.g MethyLumiSet or RGChannelSetExtended. N.B. Bead count filtering will not work unless data read in as an RGGhannelSetExtended rather than an RGChannelSet.

matrix of unmethylated signal intensities, each column representing a sample (default method) or NULL when mn is a MethyLumiSet or RGChannelSetExtended object

un

pfilter 39

bn	matrix of precalculated betas, each column representing a sample, or NULL when mn is a MethyLumiSet or RGChannelSetExtended object
da	annotation data frame, such as $x@featureData@data \#methylumi package$ , or NULL when mn is a MethyLumiSet or RGChannelSetExtended object
pn	matrix of detection p-values, each column representing a sample, a MethyLumiSet or RGChannelSetExtended object
bc	matrix of arbitrary values, each column representing a sample and eeach row representing a probe, in which "NA" represents beadcount <3, or NULL when mn is a MethyLumiSet or RGChannelSetExtended object
perCount	remove sites having this percentage of samples with a beadcount $<$ 3, default set to 5
pnthresh	cutoff for detection p-value, default set to 0.05
perc	remove samples having this percentage of sites with a detection p-value greater than pnthresh, default set to 1
pthresh	remove sites having this percentage of samples with a detection p-value greater than pnthresh, default set to 1
logical.return	If it is TRUE, FALSE or TRUE is returned to indicate success

#### Value

a filtered MethyLumiSet or a list of the filtered matrices: mn: methylated intensities un: unmethylated intensities

bn: betas

da: feature data

or

a filtered MethylSet object.

#### Methods

signature(mn = "MethyLumiSet") This is used for performing the pfilter method on MethyLumiSet objects produced by methylumiR.

signature(mn = "RGChannelSetExtended") This is used for performing the pfilter method on RGChannelSetExtended objects produced by minfi.

# Note

 $Adjusted \ RGChannel SetExtended \ methods - 12/10/2015 \ Now \ outputs \ a \ Methyl Set \ object \ using \ preprocess Raw \ from \ minfi.$ 

# Author(s)

Ruth.Pidsley@kcl.ac.uk

40 pwod

#### References

[1] Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

## **Examples**

```
# MethyLumiSet method
data(melon)
melon.pf <- pfilter(melon)</pre>
```

pwod

Probe-Wise Outlier Detection

#### **Description**

'P'robe-'W'ise 'O'utlier 'D'etection via interquartile ranges.

#### Usage

```
pwod(object, mul=4)
```

#### **Arguments**

object MethyLumiSet, RGChannelSet, MethylSet object or matrix containing betas.

Mumber of interquartile ranges used to determine outlying probes. Default is 4

to ensure only very obvious outliers are removed.

#### **Details**

Detects outlying probes across arrays in methylumi and minfi objects. Outliers are probable low MAF/SNP heterozygotes.

#### Value

Returns supplied beta matrix with outlying probes coerced to NA

#### Author(s)

Tyler Gorrie-Stone - tgorri@essex.ac.uk

```
library(wateRmelon)
data(melon)
cattle <- betas(melon)
new.betas <- pwod(cattle, mul=4)</pre>
```

pwod-methods 41

pwod-methods

Methods for Function pwod in Package wateRmelon

# Description

Methods for function pwod, please see pwod for details of how function performs.

#### Methods

signature(object = "MethyLumiSet") all of the methods simply extract betas from the data object (which can be a MethylSet, MethyLumiSet, or RGChannelSet) and calculates the outliers.

qual

A measure of Normalization Violence

# **Description**

Calculates 4 metrics to assess the degree of difference between normalized and raw betas.

# Usage

```
qual(norm, raw)
```

# Arguments

norm Matrix of normalized betas

raw Matrix of raw betas

#### Value

Returns data.frame containing rmsd, sdd, sadd and srms for each sample (columns) in supplied matrices.

# Author(s)

Leo Schalkwyk

```
library(wateRmelon)
data(melon)
d.melon <- dasen(melon)
raw.bet <- betas(melon)
norm.bet <- betas(d.melon)
qual(norm=norm.bet, raw=raw.bet)</pre>
```

42 readEPIC

read.manifest

read.manifest - read in csv format Illumina chip manifest files

# Description

read.manifest - read in csv format Illumina chip manifest files

# Usage

```
read.manifest(file)
```

# Arguments

file

# **Details**

This function is probably not much use for calling directly. It mostly exists to be called by canno.

# Value

a list of of dataframes of data prepared for making IlluminaMethylationManifest

readEPIC

readEPIC

# Description

Reads Epic arrays from raw idats into MethyLumiSet objects from directory.

# Usage

```
readEPIC(idatPath, barcodes=NULL, pdat=NULL,parallel=F,n=T,oob=F,force=F, ...)
```

# Arguments

idatPath	Path directory where .idat files are located. readEPIC looks in the specified path and converts all .idats within path to relevant barcodes, which is then passed to a modified version of methylumIDAT to parse all idats present in the specified directory.
barcodes	If NULL, function will search supplied argument in "idatPath" for all idats within directory. If given a vector of barcodes, readEPIC will search for those specific barcodes within the idatPath supplied.
parallel	If TRUE, an attempt will be made to process using multiple cores on a multicore machine

readEPIC 43

pdat	A data frame describing the samples. A special column named "barcodes" can be used to specify the barcodes to be read when using methylumIDATepic. See methylumIDAT for usage
n	If TRUE, beadcounts from .idats will be included in final object
oob	If TRUE, out-of-band (OOB) or opposite-channel signals will be kept
force	If TRUE, will combine EPIC IDATs read with differing dmaps
	Additional arguments passed to methylumIDAT

#### **Details**

Read a set of .idat files within a file directory and return a MethylumiSet object.

## Value

A MethyLumiSet object.

#### Note

Contains heavily modified version of methylumIDAT and other accessory functions used to construct a MethylumiSet object, specifically tailored for EPIC arrays. readEPIC can also handle 450k and 27k arrays as methylumIDAT functionality for these platforms remains unchanged.

Alternatively it is possible to invoke methylumIDATepic to use the modified version methylumIDAT, which has similar usage.

EPIC manifest has since been updated to B4 version, which has notably fewer probes than previous manifests. It is entirely possible that we will migrate to the manifest packages available on BioConductor and allow for versioning control.

# Author(s)

Tyler Gorrie-Stone - tgorri@essex.ac.uk

#### References

methylumi

```
#Ficticious file pathway
# path <- "Data/Experiment/Idatlocation"
# data <- readEPIC(path, barcodes = NULL oob=F, n=T)</pre>
```

readPepo

readPepo	readPepo - read (any kind of) Illumina DNA methylation array idat
·	files into a methylumi object

# Description

Pepo is a botanical term for any melon-like fruit. Given the appropriate manifest file, this function should be able to read any of the Illumina Infinium DNAm arrays including 450k, EPIC, EPIC2, MSA and Mouse.

# Usage

```
readPepo(
  idatdir = ".",
  filelist = NULL,
  barcodelist = NULL,
  manifest = NULL,
  parallel = F,
  n = F,
  pdat = NULL,
  oob = F,
  two = TRUE
)
```

# Arguments

idatdir	the directory with the idatfiles. Currently only handle one directory.
filelist	optional list of idat files to process.
barcodelist	optional list of barcodes to process.
manifest	name of a IlluminaMethylationManifest object or a csv format manifest. If missing, will run $idet()$ on one of the idat files.
parallel	try to use multiple cores.
n	keep beadcounts.
pdat	optional data.frame describing the samples.
two	are there two different assay types (true of human methylation arrays except $27k) \\$
keep	out-of-band (OOB) or opposite-channel signals

# Value

A 'MethyLumiSet' object.

seabi 45

Calculate a performance metric based on male-female differences for
Illumina methylation 450K arrays

#### **Description**

Calculates an area under ROC curve - based metric for Illumina 450K data using a t-test for male-female difference as the predictor for X-chromosome location of probes. The metric is 1-area so that small values indicate good performance, to match our other, standard error based metrics gcose and dmrse. Note that this requires both male and female samples of known sex and can be slow to compute due to running a t-test on every probe.

# Usage

```
seabi(bn, stop = 1, sex, X)
```

# **Arguments**

bn	a matrix of betas (default method) or an object containing betas i.e. a MethyLumiSet object (methylumi package), a MethylSet or RGChannelSet object (minfi package) or a exprmethy450 object (IMA package).
stop	partial area under curve is calculated if stop value <1 is provided
sex	a factor giving the sex of each sample (column)
Х	a logical vector of length equal to the number of probes, true for features mapped to X-chromosome

## Value

a value between 0 and 1. values close to zero indicate high data quality as judged by the ability to discriminate male from female X-chromosome DNA methylation.

# Author(s)

leonard.schalkwyk@kcl.ac.uk

#### References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

```
library(methylumi)
data(melon)
sex <- pData(melon)$sex
X <- melon@featureData@data$CHR=='X'
seabi(betas(melon), sex=sex, X=X)</pre>
```

46 seabird

```
# methylumi method
    seabi(melon, sex=sex, X=X)
```

seabi-methods

Methods for Function seabi in Package wateRmelon

## Description

Methods for function seabi in package **wateRmelon**. Please see seabi for details of the calculation of this ROC AUC performance metric.

#### Methods

signature(betas = "exprmethy450") all of the methods simply extract betas from the data object (which can be a exprmethy450, MethylSet, MethyLumiSe, or RGChannelSet) and calculate the metric. All the methods also require a factor differentiating male from female samples.

seabird

Calculate ROC area-under-curve for X-chromosome sex differences (internal function for calculating the seabi metric)

## **Description**

This is a wrapper for the prediction and performance functions from the ROCR package that takes a vector of p-values and a vector of true or false for being on the X. See seabi function which does everything.

#### **Usage**

```
seabird(pr, stop = 1, X)
```

#### Arguments

pr a vector of p-values, such as calculated by seabird

stop fraction for partial area under curve. For example 0.1 gives you the area for the

lowest 10% of p-values.

X logical vector the same length as pv, true for features mapped to X-chromosome

#### Value

Returns an area value between 0 and 1, where 1 is the best possible performance.

## Author(s)

Leonard C Schalkwyk 2012 Leonard.Schalkwyk@kcl.ac.uk

sextest 47

#### References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

sextest Test Illumina methylation 450K array probes for sex difference (internal function for calculating seabi performance metric)

# **Description**

This is a wrapper for 1m which does the equivalent of a Student t-test for difference in betas between males and females for each row of a matrix of betas.

#### Usage

```
sextest(betas, sex, ...)
```

## **Arguments**

betas a matrix of betas, each row is a probe, each column a sample sex a factor with 2 levels for male and female additional arguments to be passed to 1m

#### Value

Returns a vector of p-values of length equal to the number of rows of betas

#### Author(s)

Leonard.Schalkwyk@kcl.ac.uk

# References

Pidsley R, Wong CCY, Volta M, Lunnon K, Mill J, Schalkwyk LC: A data-driven approach to preprocessing Illumina 450K methylation array data (submitted)

## See Also

seabi seabird

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## **Examples**

```
#MethyLumiSet method
  data(melon)
  sex <- pData(melon)$sex
  melon.sextest<-sextest(betas(melon),sex)

#MethyLumiSet method with quality control step
  data(melon)
  melon.dasen <- dasen(melon)
  sex <- pData(melon.dasen)$sex
  melon.sextest<-sextest(betas(melon.dasen),sex)</pre>
```

smokp

Smoking Prediction from methylomic expression data

# Description

Predict smoking from samples using various methods

## Usage

```
smokp(betas, method, sst)
```

#### **Arguments**

betas Matrix of betas or MethyLumiSet or MethylSet object. Rows are Illumina IDs

referring to CpG sites and Columns refer to samples or participants.

method Currently: 'AHRR', 'McCartney', 'Maas', 'Sugden', 'Teschendorff', 'Yu', 'Gao',

'Yang', 'Zhang', 'Wen', 'Langdon', 'SSt', 'Packyears', 'Cessation', and 'All'. If "All" smokp will seek to predict smoking using all methods else will use the method specified. Default is "SSt". If 'Teschendorff', 'Yu', 'Gao', 'Yang' or

'Langdon' specified then smoking status is required.

Named vector describing smoking status, coded as 'Current', 'Former', or 'Never',

of participant for each sample (where names match rownames of betas).

# Value

Returns data frame of predicted smoking per sample.

#### Author(s)

Original Functions: See References.

wateRmelon: Tyler Gorrie-Stone, Leo Schalkwyk, Louis El Khoury

smokp: Alexandria Andrayas

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#### References

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# **Examples**

```
data(melon)
# note, melon is not a complete dataset, does not work with all methods
smokp(melon, method="McCartney", sst=NULL)
```

wm\_internal

Internal functions for readEPIC and other wateRmelon functions introduced in v 1.13.1

#### **Description**

few if any functions of interest to users

#### Usage

```
DataToNChannelSet2(mats, chans = c(Cy3 = "GRN", Cy5 = "RED"), parallel = F, protocol.data = F, IDAT = TRU
```

# Arguments

```
mats
chans
parallel
protocol.data
IDAT
force
```

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