

Package ‘wateRmelon’

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Type Package

Title Illumina 450 methylation array normalization and metrics

Version 1.34.0

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

License GPL-3

Depends R (>= 2.10), Biobase, limma, methods, matrixStats, methylumi, lumi, ROC, IlluminaHumanMethylation450kanno.ilmn12.hg19, illuminaio

Imports Biobase

Enhances minfi

Suggests RPMM, IlluminaHumanMethylationEPICanno.ilm10b2.hg19, IlluminaHumanMethylationEPICmanifest, irlba

LazyLoad yes

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 bgeqot.R bgeqq2.R bgeqqn.R BMIQ_1.1.R combo.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 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 predictSex.R pwod.R readEPIC.R preprocessIlluminaMethylation.R referenceQuantiles.R robustQuantileNorm_Illumina450K.probeCategories.R

robustQuantileNorm_Illumina450K.R seabird.R sextest.R summits.R
 swan2.R uniqueAnnotationCategory.R qual.R AllGenerics.R
 x_methylumi.R y_minfi.R

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wateRmelon-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.

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)

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)

`agep`*Age Prediction from methylomic expression data*

Description

Predict age of samples using Horvaths Coefficients

Usage

```
agep(betas, coeff=NULL, method=c('horvath','hannum') ,...)
```

Arguments

<code>betas</code>	Matrix of betas or MethyLumiSet or MethylSet object.
<code>coeff</code>	If NULL, will default to Horvaths coefficients. Otherwise can be supplied with vector of own coefficients with an intercept as first element.
<code>method</code>	currently either "horvath" or "hannum". If using "hannum" remember to call <code>data(hannumCoef)</code> before analysis and supply <code>coeff</code> with the hannum coefficients, otherwise will default to "horvath".
<code>...</code>	To pass to arguments to downstream functions to specify adult.age

Value

Returns matrix of predicted ages per sample.

Author(s)

Original Functions: Steve Horvath

watermelon Implementation: 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)
data(coef)
agep(melon,coeff=coef)
# or
agep(betas(melon),coeff=coef)
```

as.methylumi-methods *Methods for Function as.methylumi*

Description

Returns a MethyLumiSet object populated 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, ...)
```

Arguments

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
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 getColumn()

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

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

x 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 (fucs)

Description

Internal functions for peak.correction

Usage

Beta2M(B)

Arguments

B 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 exprmethy450 objects

Arguments

bn	An exprmethy450 object.
fudge	value added to total intensity to prevent denominators close to zero when calculating betas

Details

betaqn quantile normalizes betas

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,
## S4 method for signature 'MethyLumiSet'
BMIQ(beta.v, nL=3, doH=TRUE, nfit=50000, 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

<code>beta.v</code>	vector consisting of beta-values for a given sample, or a <code>MethyLumiSet</code> or <code>MethylSet</code> containing multiple samples. For the <code>MethyLumiSet</code> and <code>MethylSet</code> methods, this is the only required argument, and the function will be run on each sample.
<code>design.v</code>	corresponding vector specifying probe design type (1=type1,2=type2). This must be of the same length as <code>beta.v</code> and in the same order.
<code>nL</code>	number of states in beta mixture model. 3 by default. At present BMIQ only works for <code>nL=3</code> .
<code>doH</code>	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.
<code>nfit</code>	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.
<code>th1.v</code>	thresholds used for the initialisation of the EM-algorithm, they should represent best guesses for calling type1 probes hemi-methylated and methylated, and will be refined by the EM algorithm. Default values work well in most cases.
<code>th2.v</code>	thresholds used for the initialisation of the EM-algorithm, they should represent best 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 <code>th1.v</code> and a modified PBC correction method.
<code>niter</code>	maximum number of EM iterations to do. This number should be large enough to yield good fits to the type1 distribution. By default 5.
<code>tol</code>	tolerance convergence threshold for EM algorithm. By default 0.001.
<code>plots</code>	logical specifying whether to plot the fits and normalised profiles out. By default TRUE.

sampleID	the ID of the sample being normalised.
pri	logical: print verbose progress information?
pnbeta.v	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

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	<i>Calculate bisulphite conversion</i>
-------	--

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 bisulfite 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.

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
```

colnames-methods	<i>Methods for Function colnames in Package wateRmelon</i>
------------------	--

Description

Methods for function colnames in package **wateRmelon**.

Methods:

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

combo	<i>Combine MethyLumiSet objects</i>
-------	-------------------------------------

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>

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
melon <- melon
sampleNames(melon) <- gsub('^6', 7, sampleNames(melon))
combo(melon, melon)

```

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

```

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 Satrix 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

<code>bn, data</code>	matrix of precalculated betas, each column representing a sample
<code>onetwo</code>	character vector or factor of length <code>nrow(mn)</code> indicating assay type 'I' or 'II'
<code>pn</code>	matrix of detection p-values, each column representing a sample
<code>da, anno</code>	annotation data frame, such as <code>x@featureData@data</code> #methylumi package. If NULL, the swan method requires the <code>IlluminaHumanMethylation450kmanifest</code> package.
<code>qc</code>	control probe intensities: list of 2 matrices, Cy3 and Cy5, with rownames, such as produced by <code>intensitiesByChannel(QCdata(x))</code> #methylumi package
<code>fudge</code>	value added to total intensity to prevent denominators close to zero when calculating betas
<code>return.MethylSet</code>	if TRUE, returns a MethylSet object instead of a naked matrix of betas.
<code>ret2</code>	if TRUE, returns a list of intensities and betas instead of a naked matrix of betas.
<code>...</code>	additional argument <code>roco</code> for <code>dfsfit</code> giving Sentrix rows and columns. This allows a background gradient model to be fit. This is split from data column names by default. <code>roco=NULL</code> disables model fitting (and speeds up processing), otherwise <code>roco</code> 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)

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

nasen same as `naten` but type I and type II 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)
```

dasen-methods

Calculate normalized betas from MethyLumiSets of 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.

Arguments

mn, mns, data, bn

A MethyLumiSet object. Sample names names are used to get Satrix row and column by default, see '...'.
 fudge value added to total intensity to prevent denominators close to zero when calculating betas
 ... additional argument roco for dfsfit giving Satrix 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

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 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 and 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 (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 ) nanet ( mn, 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.

Arguments

<code>mn, mns</code>	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 <code>'...'</code> .
<code>un, uns</code>	matrix of unmethylated signal intensities, each column representing a sample (default method) or NULL when <code>mn</code> is an object containing methylated and unmethylated values
<code>bn, data</code>	matrix of precalculated betas, each column representing a sample
<code>onetwo</code>	character vector or factor of length <code>nrow(mn)</code> indicating assay type 'I' or 'II'
<code>da, anno</code>	annotation data frame, such as <code>x@featureData@data</code> #methylumi package
<code>qc</code>	control probe intensities: list of 2 matrices, Cy3 and Cy5, with rownames, such as produced by <code>intensitiesByChannel(QCdata(x))</code> #methylumi package
<code>fudge</code>	value added to total intensity to prevent denominators close to zero when calculating betas
<code>...</code>	additional argument <code>roco</code> for <code>dfsfit</code> giving Sentrix rows and columns. This allows a background gradient model to be fit. This is split from data column names by default. <code>roco=NULL</code> disables model fitting (and speeds up processing), otherwise <code>roco</code> 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

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 equalised first.

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

danen background equalisation only, no normalization

daten1 same as `naten`, except type I 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 and 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) nanes (mns, uns, onetwo, fudge = 100) danes (mn, un, onetwo, fudge = 100, ...) danet (mn, un, onetwo, fudge = 100, ...) danen (mns, uns, onetwo, fudge = 100, ...) daten1(mn, un, onetwo, fudge = 100, ...) daten2(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

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)

```

Arguments

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'
roco	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' (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())
```

Arguments

<code>betas</code>	a matrix of betas (default method), a <code>MethylumiSet</code> object (<code>methylumi</code> package), a <code>MethylSet</code> or <code>RGChannelSet</code> object (<code>minfi</code> package) or a <code>exprmethy450</code> object (<code>IMA</code> package).
<code>idmr</code>	a character vector of iDMR probe names such as returned by <code>iDMR()</code>

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)
```

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`, `MethylumiSe`, or `RGChannelSet`) and calculate the metric.

estimateCellCounts

*Cell Proportion Estimation using wateRmelon***Description**

Estimates relative proportion of pure cell types within a sample, mostly identical to [estimateCellCounts](#). Currently, only a reference data-set exists for 450k arrays. As a result, if performed on EPIC data, function will convert methylumi object to 450k array dimensions.

Usage

```
estimateCellCounts.wmln(
  object,
  platform = c("450k", "EPIC", "27k"),
  mn = NULL,
  un = NULL,
  bn = NULL,
  perc = 1,
  compositeCellType = "Blood",
  probeSelect = "auto",
  cellTypes = c("CD8T", "CD4T", "NK", "Bcell", "Mono", "Gran"),
  referencePlatform = c("IlluminaHumanMethylation450k",
    "IlluminaHumanMethylationEPIC",
    "IlluminaHumanMethylation27k"),
  returnAll = FALSE,
  meanPlot = FALSE,
  verbose=TRUE,
  ...)
```

Arguments

<code>object</code>	An object of class <code>methylumisest</code> , which contains (un)normalised methylated and unmethylated intensities
<code>platform</code>	Which micro-array platform was used to analysed samples

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 identical dimensions to object.
bn	if NULL will call betas(object), otherwise can be given matrix of identical dimensions 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 increase in performance
compositeCellType	Which composite cell type is being deconvoluted. Should be either "Blood", "CordBlood", or "DLPCF"
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 deconvolution? 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 discriminating 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 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.

genki	<i>SNP derived performance metrics for Illumina 450K DNA methylation arrays.</i>
-------	--

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)

Examples

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

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

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`, `MethyLumiSe`, 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))
```

Arguments

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

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)

iDMR	<i>Imprinting differentially methylated region probes of Illumina 450 arrays</i>
------	--

Description

A character vector of 227 probes 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/>

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	<i>Small MethyLumi data set for examples and testing</i>
-------	--

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

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 <code>sextest</code> , one per row of betas
X	a logical vector of the same length as <code>pv</code> , 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 <code>iDMR()</code> or <code>data(iDMR)</code>
subset	index or character vector giving a subset of betas to be tested

Value

dmrse_row	see <code>dmrse_row</code>
dmrse_col	see <code>dmrse_col</code>
dmrse	see <code>dmrse</code>
gcoms_a	see <code>genki</code>
gcose_a	see <code>genki</code>
gcoms_b	see <code>genki</code>
gcose_b	see <code>genki</code>
gcoms_c	see <code>genki</code>
gcose_c	see <code>genki</code>
seabird	see <code>seabi</code>

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)
```

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

```
outlyx(x, iqr=TRUE, iqrP=2, pc=1,
       mv=TRUE, mvP=0.15, plot=TRUE, ...)
```

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
mvP	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.

Note

May perform poorly on normalized data

Author(s)

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

Examples

```
library(wateRmelon)
data(melon)
outliers <- outlyx(melon,iqr=TRUE, iqrP=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 RGGchannelSetExtended rather than an RGChannelSet.
un	matrix of unmethylated signal intensities, each column representing a sample (default method) or NULL when mn is a MethyLumiSet or RGChannelSetExtended object
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 each 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 RGChannelSetExtended methods - 12/10/2015 Now outputs a MethylSet object using preprocessRaw from minfi.

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)

Examples

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

predictSex	<i>Sex Prediction of samples assayed on the Illumina Infinium BeadChip platform</i>
------------	---

Description

Predicts the sex of samples

Usage

```
predictSex(x, x.probes, pc = 2, plot=TRUE, irlba=TRUE, center=FALSE, scale. = FALSE)
```

Arguments

x	A matrix containing values equivalent to betas. Can be the full set of CpG features or pre-filtered for X-chromosomes.
x.probes	Default NULL, assuming that x is a beta matrix consisting of only X-chromosome probes, otherwise supply with vector containing logical or numeric index of probes on the X-chromosome.
pc	The principal component to guess sex across (usually the 2nd one in most cases)
plot	Logical, indicated whether or not to plot the prediction
irlba	Logical, indicate whether or not to use the faster method to generate principal components
center	Logical, indicate whether or not to center data around 0
scale.	Logical, indicate whether or not to scale data prior to prcomp

Details

Function attempts to predictSex based on the colMeans of the X chromosome and principal components. Assumes that there are two sexes to predict and as a result will perform poorly on data-sets consisting of a single sex.

Value

Character Vector of predicted sex for each sample, samples that were not guess are assigned 'Unknown'

Author(s)

Tyler J. Gorrie-Stone Who to contact: <tgorri@essex.ac.uk>

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.
mul	Number 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

Examples

```
library(watermelon)
data(melon)
cattle <- betas(melon)
new.betas <- pwod(cattle, mul=4)
```

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

<code>norm</code>	Matrix of normalized betas
<code>raw</code>	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

Examples

```
library(watermelon)
data(melon)
d.melon <- dasen(melon)
raw.bet <- betas(melon)
norm.bet <- betas(d.melon)
qual(norm=norm.bet, raw=raw.bet)
```

`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.
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

Examples

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

seabi	<i>Calculate a performance metric based on male-female differences for Illumina methylation 450K arrays</i>
-------	---

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)
X	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)

Examples

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

# 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

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	<i>Test Illumina methylation 450K array probes for sex difference (internal function for calculating seabi performance metric)</i>
---------	--

Description

This is a wrapper for `lm` 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 <code>lm</code>

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](#)

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)
```

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 =
```

Arguments

mats

chans

parallel

protocol.data

IDAT

force

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