Package 'epimutacions'

October 16, 2022

Title Robust outlier identification for DNA methylation data

Version 1.0.3

Description The package includes some statistical outlier detection methods for epimutations detection in DNA methylation data.

The methods included in the package are MANOVA, Multivariate linear models, isolation forest, robust mahalanobis distance, quantile and beta.

The methods compare a case sample with a suspected disease against a reference panel (composed of healthy individuals) to identify epimutations in the given case sample.

It also contains functions to annotate and visualize the identified epimutations.

biocViews DNAMethylation, BiologicalQuestion, Preprocessing,

StatisticalMethod, Normalization

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Imports minfi, bumphunter, isotree, robustbase, ggplot2,

 $Genomic Ranges, \, Genomic Features, \, IRanges, \, Summarized Experiment, \,$

stats, matrixStats, BiocGenerics, S4Vectors, utils, biomaRt,

BiocParallel

Suggests AnnotationDbi, AnnotationHub, testthat, knitr, rmarkdown,

BiocStyle, a4Base, kableExtra,

IlluminaHumanMethylation450kanno.ilmn12.hg19,

IlluminaHumanMethylation450kmanifest,

IlluminaHumanMethylationEPICmanifest,

TxDb.Hsapiens.UCSC.hg19.knownGene,

TxDb.Hsapiens.UCSC.hg18.knownGene,

TxDb.Hsapiens.UCSC.hg38.knownGene, rtracklayer, methods,

ensembldb, ExperimentHub,

IlluminaHumanMethylationEPICanno.ilm10b2.hg19, grid, gridExtra,

grDevices, tibble, Gviz, ggrepel, reshape2, purrr,

GenomeInfoDb, Homo.sapiens

VignetteBuilder knitr

Encoding UTF-8

Depends R (>= 4.2.0), epimutacionsData

Roxygen list(markdown = TRUE)

URL https://github.com/isglobal-brge/epimutacions

Maintainer Leire Abarrategui <abarrategui.leire@gmail.com>

R topics documented:

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Description

add ansamble regulatory

Add ENSEMBL regulatory regions to epimutations

Usage

```
add_ensemble_regulatory(epimutations, build = "37")
```

Arguments

epimutations	a data frame object containing the result from epimutations or epimutations_one_leave_out functions.
build	the build used to define epimutations coordinates. By default, it is '37', corresponding to Illumina annotation.

Value

The function returns a data frame object containing the results of epimutations or epimutations_one_leave_out with some additional variables describing regulatory elements from ENSEMBL.

Note that a single epimutation might overlap with more than one regulatory region. In that case, the different regulatory regions are separated by ///.

- ensembl_reg_idRegion identifier from ENSEMBL
- ensembl_reg_coordinatesCoordinates for the ENSEMBL regulatory regions
- ensembl_reg_typeType of regulatory region
- ensembl_reg_tissuesActivity of the regulatory region per tissue. The different activation states are separated by /

4 annotate_cpg

annotate_cpg Annotate the DMR resulting from epimutacions package

Description

This function annotates a differentially methylated region

Usage

```
annotate_cpg(
  data,
  db,
  split = ",",
  epi_col = "cpg_ids",
  gene_col = "GencodeBasicV12_NAME",
  feat_col = "Regulatory_Feature_Group",
  relat_col = "Relation_to_Island",
  build = "37",
  omim = TRUE
)
```

Arguments

data	DataFrame-like object.
db	a character string specifying the Database to use for annotation. E.g: 'IlluminaHumanMethylationEPICa
split	a character string containing the separator for CpG ids. Default ','.
epi_col	CpG ids, should be row names in the data base.
gene_col	column name from where to extract gene names. Default: 'GencodeBasicV12_NAME'.
feat_col	$column\ name\ from\ where\ to\ extract\ CpG\ feature\ groups.\ Default:\ \verb 'Regulatory_Feature_Group'.$
relat_col	column name from where to extract relation to island info. Default: 'Relation_to_Island'.
build	The build for bioMart. Default '37'.
omim	a boolean, if TRUE will annotate OMIMs as well. Takes a bit longer. Default TRUE.

Value

The function returns a DataFrame-like object annotated.

annotate_epimutations 5

```
annotate_epimutations Annotate\ the\ results\ of\ epimutations\ or\ epimutations\_one\_leave\_out\ functions
```

Description

Information about close genes and regulatory elements for epimutations.

Usage

```
annotate_epimutations(
  epi_results,
  db = "IlluminaHumanMethylationEPICanno.ilm10b2.hg19",
  build = "37",
  ...
)
```

Arguments

epi_results	a data frame object containing the output from epimutations or epimutations_one_leave_out functions.
db	a character string containing the Illumina annotation package used to annotate the CpGs.
build	a character string containing the genomic build where the epimutations are mapped. The default is GRCh37 (build = "37"). To use GRCh38 set built to NULL.
	Further arguments passed to annotate_cpg.

Value

The function returns the input object epi_results with additional columns containing the information about the genes or overlapping regulatory features.

See annotate_cpg and add_ensemble_regulatory for an in-depth description of these variables.

Examples

```
data(res.epi.manova)
#Annotate the epimutations
#anno_results <- annotate_epimutations(res.epi.manova)</pre>
```

6 betas_sd_mean

betas_from_bump	Obtains bumps beta values

Description

The function obtains beta values corresponding to the CpGs into DMRs.

Usage

```
betas_from_bump(bump, fd, betas)
```

Arguments

bump the result from bumphunter.

fd a data frame containing the genomic ranges for each CpGs. betas a matrix containing the beta values for all CpGs in each sample.

Value

The function returns a data frame containing the beta values for each sample and CpG into DMR.

betas_sd_mean	Computes beta values, standard deviation and mean to plot the epimu-
	tation

Description

Computes the beta values, population mean and 1, 1.5, and 2 standard deviations from the mean of the distribution necessary to plot the epimutations.

Usage

```
betas_sd_mean(gr)
```

Arguments

gr a GRanges object obtained from create_GRanges_class function.

Value

The function returns a list containing the melted beta values, the population mean and 1, 1.5, and 2 standard deviations from the mean of the distribution.

cols_names 7

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Description

Sets common column names in a given data frame containing the CpGs genomic ranges or a DMR (result of epimutations or epimutations_one_leave_out function).

Usage

```
cols_names(x, cpg_ids_col = FALSE)
```

Arguments

x a data frame containing the genomic ranges or a DMR (a row of the results of

epimutations or epimutations_one_leave_out function).

cpg_ids_col a boolean, if TRUE the input data frame contains the CpGs names column.

Value

The function returns a data frame containing the column names to carry out the analysis without any error.

```
create_GRanges_class Generates a GRanges object
```

Description

This function makes a GRanges object from a GenomicRatioSet.

Usage

```
create_GRanges_class(methy, cpg_ids)
```

Arguments

methy a GenomicRatioSet object containing the control and case samples used in

epimutations or epimutations_one_leave_out function.

cpg_ids a character string specifying the name of the CpGs in the DMR of interest.

Value

The function returns a GRanges object containing the beta values and the genomic ranges of the CpGs of interest.

8 epimutations

 ${\tt epimutations}$

Epimutations analysis based on outlier detection methods

Description

The function identifies differentially methylated regions in a case sample by comparing it against a control panel.

Usage

```
epimutations(
   case_samples,
   control_panel,
   method = "manova",
   chr = NULL,
   start = NULL,
   end = NULL,
   epi_params = epi_parameters(),
   maxGap = 1000,
   bump_cutoff = 0.1,
   min_cpg = 3,
   verbose = TRUE
)
```

Arguments

case_samples	a GenomicRatioSet object containing the case samples. See the constructor function $GenomicRatioSet$, $makeGenomicRatioSetFromMatrix$.
control_panel	a GenomicRatioSet object containing the control panel (control panel).
method	a character string naming the outlier detection method to be used. This can be set as: "manova", "mlm", "iForest", "mahdist", "quantile" and "beta". The default is "manova". For more information see Details .
chr	a character string containing the sequence names to be analysed. The default value is NULL.
start	an integer specifying the start position. The default value is NULL.
end	an integer specifying the end position. The default value is NULL.
epi_params	the parameters for each method. See the function epi_parameters.
maxGap	the maximum location gap used in bumphunter method.
bump_cutoff	a numeric value of the estimate of the genomic profile above the cutoff or below the negative of the cutoff will be used as candidate regions.
min_cpg	an integer specifying the minimum CpGs number in a DMR.
verbose	logical. If TRUE additional details about the procedure will provide to the user.

The default is TRUE.

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Details

The function compares a case sample against a control panel to identify epimutations in the given sample. First, the DMRs are identified using the bumphunter approach. After that, CpGs in those DMRs are tested in order to detect regions with CpGs being outliers. For that, different outlier detection methods can be selected:

- Multivariate Analysis of Variance ("manova"). manova
- Multivariate Linear Model ("mlm")
- Isolation Forest ("iForest") isolation.forest
- Robust Mahalanobis Distance ("mahdist") covMcd
- Quantile distribution ("quantile")
- Beta ("beta")

We defined candidate epimutation regions (found in candRegsGR) based on the 450K array design. As CpGs are not equally distributed along the genome, only CpGs closer to other CpGs can form an epimutation. More information can be found in candRegsGR documentation.

Value

The function returns an object of class tibble containing the outliers regions. The results are composed by the following columns:

- epi_id: systematic name for each epimutation identified. It provides the name of the used anomaly detection method.
- sample: the name of the sample containing the epimutation.
- chromosome, start and end: indicate the location of the epimutation.
- sz: the window's size of the event.
- cpg_n: the number of CpGs in the epimutation.
- cpg_ids: the names of CpGs in the epimutation.
- outlier_score:
 - For method manova it provides the approximation to F-test and the Pillai score, separated by /.
 - For method mlm it provides the approximation to F-test and the R2 of the model, separated by /.
 - For method iForest it provides the magnitude of the outlier score.
 - For method beta it provides the mean outlier p-value.
 - For methods quantile and mahdist it is filled with NA.
- outlier_direction: indicates the direction of the outlier with "hypomethylation" and "hypermethylation"
 - For manova, mlm, iForest, and mahdist it is computed from the values obtained from bumphunter.
 - For quantile it is computed from the location of the sample in the reference distribution (left vs. right outlier).
 - For method beta it return a NA.

- pvalue:
 - For methods manova, mlm, and iForest it provides the p-value obtained from the model.
 - For method quantile, mahdist and beta is filled with NA.
- adj_pvalue: for methods with p-value (manova and mlm adjusted p-value with Benjamini-Hochberg based on the total number of regions detected by Bumphunter.
- epi_region_id: Name of the epimutation region as defined in candRegsGR.
- CRE: cREs (cis-Regulatory Elements) as defined by ENCODE overlapping the epimutation region. Different cREs are separated by ;.
- CRE_type: Type of cREs (cis-Regulatory Elements) as defined by ENCODE. Different type are separeted by, and different cREs are separated by;.

Examples

```
data(GRset)
#Find epimutations in GSM2562701 sample of GRset dataset

case_samples <- GRset[,11]
control_panel <- GRset[,1:10]
epimutations(case_samples, control_panel, method = "manova")</pre>
```

```
epimutations_one_leave_out
```

Epimutations analysis based on outlier detection methods

Description

This function is similar to epimutations with the particularity that when is more than one case sample, the remaining case samples are included as controls.

Usage

```
epimutations_one_leave_out(
  methy,
  method = "manova",
  epi_params = epi_parameters(),
  BPPARAM = BiocParallel::SerialParam(),
  verbose = TRUE,
  ...
)
```

Arguments

methy	a GenomicRatioSet object containing the samples for the analysis. See the constructor function GenomicRatioSet, makeGenomicRatioSetFromMatrix.
method	a character string naming the outlier detection method to be used. This can be set as: "manova", "mlm", "iForest", "mahdist", "barbosa" and beta. The default is "manova". For more information see Details .
epi_params	the parameters for each method. See the function epi_parameters.
BPPARAM	("BiocParallelParam") BiocParallelParam object to configure parallelization execution. By default, execution is non-parallel.
verbose	logical. If TRUE additional details about the procedure will provide to the user. The default is TRUE.
	Further parameters passed to epimutations

Details

The function compares a case sample against a control panel to identify epimutations in the given sample. First, the DMRs are identified using the bumphunter approach. After that, CpGs in those DMRs are tested in order to detect regions with CpGs being outliers. For that, different anomaly detection methods can be selected:

- Multivariate Analysis of Variance ("manova"). manova
- Multivariate Linear Model ("mlm")
- Isolation Forest ("iForest") isolation.forest
- Robust Mahalanobis Distance ("mahdist") covMcd
- Barbosa ("barbosa")

Value

The function returns an object of class tibble containing the outliers regions. The results are composed by the following columns:

- epi_id: the name of the anomaly detection method that has been used to detect the epimutation
- sample: the name of the sample where the epimutation was found.
- chromosome, start and end: indicate the location of the epimutation.
- sz: the number of base pairs in the region.
- cpg_n: number of CpGs in the region.
- cpg_ids: differentially methylated CpGs names.
- outlier_score:
 - For method manova it provides the approximation to F-test and the Pillai score, separated by /.
 - For method mlm it provides the approximation to F-test and the R2 of the model, separated by /.
 - For method iForest it provides the magnitude of the outlier score.

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- For methods barbosa and mahdist is filled with NA.
- outlier_significance:
 - For methods manova, mlm, and iForest it provides the p-value obtained from the model.
 - For method barbosa and mahdist is filled with NA.
- outlier_direction: indicates the direction of the outlier with "hypomethylation" and "hypermethylation"
 - For manova, mlm, iForest, and mahdist it is computed from the values obtained from bumphunter.
 - For barbosa it is computed from the location of the sample in the reference distribution (left vs. right outlier).

Examples

epi_beta

Identifies epimutations based on a beta distribution.

Description

epi_beta method models the DNA methylation data using a beta distribution. First, the beta distribution parameters of the reference population are precomputed and passed to the method. Then, we compute the probability of observing the methylation values of the case from the reference beta distribution. CpGs with p-values smaller than a threshold pvalue_threshold and with a methylation difference with the mean reference methylation higher than diff_threshold are defined as outlier CpGs. Finally, epimutations are defined as a group of contiguous outlier CpGs.

Usage

```
epi_beta(
   beta_params,
   beta_mean,
   betas_case,
   case,
   controls,
   betas,
   annot,
   pvalue_threshold,
   diff_threshold,
   min_cpgs = 3,
   maxGap
)
```

epi_iForest 13

Arguments

beta_params matrix with the parameters of the reference beta distributions for each CpG in

the dataset.

beta_mean beta values mean.

betas_case matrix with the methylation values for a case.

case case sample name.

controls control samples names.

betas a matrix containing the beta values for all samples.

annot annotation of the CpGs.

pvalue_threshold

minimum p-value to consider a CpG an outlier.

diff_threshold minimum methylation difference between the CpG and the mean methylation to

consider a position an outlier.

min_cpgs minimum number of CpGs to consider an epimutation.

maxGap maximum distance between two contiguous CpGs to combine them into an

epimutation.

Value

The function returns a data frame with the candidate regions to be epimutations.

Description

This function identifies regions with CpGs being outliers using isolation.forest approach.

Usage

```
epi_iForest(mixture, case_id, ntrees)
```

Arguments

mixture beta values matrix. Samples in columns and CpGs in rows.

case_id a character string specifying the name of the case sample.

ntrees number of binary trees to build for the model. Default is 100.

Value

The function returns the outlier score for the given case sample.

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$_{ t mahdist}$

Identifies epimutations using Robust Mahalanobis distance

Description

This function identifies regions with CpGs being outliers using the Minimum Covariance Determinant (MCD) estimator (covMcd) to compute the Mahalanobis distance.

Usage

```
epi_mahdist(mixture, nsamp = c("best", "exact", "deterministic"))
```

Arguments

mixture beta values matrix. Samples in columns and CpGs in rows.

nsamp the number of subsets used for initial estimates in the MCD. It can be set as:

"best", "exact", or "deterministic".

Details

The implementation of the method here is based on the discussion in this thread of Cross Validated

Value

The function returns the computed Robust Mahalanobis distance.

eni	_manova
CPI-	_1111011014

Identifies epimutations using MANOVA

Description

This function identifies regions with CpGs being outliers using manova approach.

Usage

```
epi_manova(mixture, model, case_id)
```

Arguments

mixture beta values matrix. Samples in columns and CpGs in rows.

model design (or model) matrix.

case_id a character string specifying the name of the case sample.

Value

The function returns the F statistic, Pillai and P value.

epi_mlm 15

epi_mlm

Detects epimutations using Multivariate Linear Model (MLM)

Description

Identifies CpGs with outlier methylation values using methylated Multivariate Linear Model

Usage

```
epi_mlm(mixture, model)
```

Arguments

mixture beta values matrix. Samples in columns and CpGs in rows.

model design (or model) matrix.

Value

The function returns the F statistic, R2 test statistic and Pillai.

```
epi_parameters Settings for parameters of epimutations and epimutations_one_leave_out functions
```

Description

Allow the user to set the values of the parameters to compute the functions epimutations and epimutations_one_leave_out.

Usage

```
epi_parameters(
  manova = list(pvalue_cutoff = 0.05),
  mlm = list(pvalue_cutoff = 0.05),
  iForest = list(outlier_score_cutoff = 0.7, ntrees = 100),
  mahdist = list(nsamp = "deterministic"),
  quantile = list(window_sz = 1000, offset_abs = 0.15, qsup = 0.995, qinf = 0.005),
  beta = list(pvalue_cutoff = 1e-06, diff_threshold = 0.1)
)
```

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Arguments

manova, mlm, iForest, mahdist, quantile, beta

method selected in the function epimutations.

pvalue_cutoff the threshold p value to select which CpG regions are outliers in manova, mlm

and beta methods.

outlier_score_cutoff

The outlier score threshold to identify outliers CpGs in isolation forest (iForest)

method. Default is 0.5.

ntrees number of binary trees to build for the model build by isolation forest (iForest)

method. Default is 100.

nsamp the number of subsets used for initial estimates in the Minimum Covariance De-

terminant which is used to compute the Robust Mahalanobis distance (mahdist). It can be set as: "best", "exact", or "deterministic". For nsamp = "best" exhaustive enumeration is done, as long as the number of trials does not exceed 100'000. For nsamp = "exact" exhaustive enumeration will be attempted however many samples are needed. In this case, a warning message may be displayed saying that the computation can take a very long time. For nsamp = "deterministic". For more information see covMcd. Default is "deterministic".

window sz

the maximum distance between CpGs to be considered in the same DMR. This

parameter is used in quantile (default: 1000).

qsup, qinf, offset_abs

The upper and lower quantiles (threshold) to consider a CpG an outlier when using quantile method, as well as the offset to consider (defaults: 0.005, 0.995,

0.15).

diff_threshold Minimum methylation difference between the CpG and the mean methylation

to consider a position an outlier.

Details

Invoking epi_parameters() with no arguments returns return a list with the default values.

Value

the function returns a list of all set parameters for each method used in epimutations and epimutations one leave out functions.

Examples

```
#Default set of parameters
epi_parameters()
#change p value for manova method
epi_parameters(manova = list("pvalue_cutoff" = 0.01))
```

epi_preprocess 17

epi_preprocess

Preprocess methylation array

Description

The epi_preprocess function reads Illumina methylation sample sheet for case samples and it merges them with RGChannelSet reference panel. The final dataset is normalized using minfi package preprocess methods.

Usage

```
epi_preprocess(
  cases_dir,
  reference_panel,
  pattern = "csv$",
  normalize = "raw",
  norm_param = norm_parameters(),
  verbose = FALSE
)
```

Arguments

cases_dir

reference_panel

an RGChannelSet object containing the reference panel (controls) samples.

pattern

What pattern is used to identify a sample sheet file.

normalize

a character string specifying the selected preprocess method. For more information see Details or minfi package user's Guide. It can be set as: "raw", "illumina", "swan", "quantile", "noob" or "funnorm".)

norm_param

the parameters for each preprocessing method. See the function norm_parameters.

verbose

logical. If TRUE additional details about the procedure will provide to the user. The default is FALSE.

Details

The epi_preprocess function reads Illumina methylation sample sheet for case samples and it merges them with RGChannelSet reference panel. The final dataset is normalized using different minfi package preprocess methods:

```
"raw": preprocessRaw
"illumina": preprocessIllumina
"swan": preprocessSWAN
"quantile": preprocessQuantile
"noob": preprocessNoob
"funnorm": preprocessFunnorm
```

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Value

epi_preprocess function returns a GenomicRatioSet object containing case and control (reference panel) samples.

Examples

epi_quantile

Identifies epimutations using quantile distribution

Description

Identifies CpGs with outlier methylation values using a sliding window approach to compare individual methylation profiles of a single case sample against all other samples from reference panel (controls)

Usage

```
epi_quantile(
  case,
  fd,
  bctr_pmin,
  bctr_pmax,
  controls,
  betas,
  window_sz = 1000,
  N = 3,
  offset_abs = 0.15
)
```

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Arguments

case	beta values for a single case (data.frame). The samples as single column and CpGs in rows (named).
fd	feature description as data.frame having at least chromosome and position as columns and and CpGs in rows (named).
bctr_pmin	Beta value observed at 0.01 quantile in controls. A beta values has to be lower or equal to this value to be considered an epimutation.
bctr_pmax	Beta value observed at 0.99 quantile in controls. A beta values has to be higher or equal to this value to be considered an epimutation.
controls	control samples names.
betas	a matrix containing the beta values for all samples.
window_sz	Maximum distance between a pair of CpGs to defined an region of CpGs as epimutation (default: 1000).
N	Minimum number of CpGs, separated in a maximum of window_sz bass, to defined an epimutation (default: 3).
offset_abs	Extra enforcement defining an epimutation based on beta values at 0.005 and 0.995 quantiles (default: 0.15).

Value

The function returns a data frame with the regions candidates to be epimutations.

getBetaParams	Model methylation as a beta distribution	
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Description

Model methylation as a beta distribution

Usage

getBetaParams(x)

Arguments

x Matrix of methylation expressed as a beta. CpGs are in columns and samples in rows.

Value

Beta distribution.

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get_candRegsGR

Candidate regions to be epimutations

Description

Load candidate regions to be epimutations from epimutacionsData package in ExperimentHub.

Usage

```
get_candRegsGR()
```

Value

The function returns a GRanges object containing the candidate regions.

get_ENSEMBL_data

Get ENSEMBL regulatory features overlapping a genomic region

Description

This function queries for ENSEMBL regulatory features and collapse them to return a single record.

Usage

```
get_ENSEMBL_data(chromosome, start, end, mart)
```

Arguments

chromosome Chromosome of the region

start Start of the region end End of the region

mart Mart object to perform the ENSEMBL query

Value

data.frame of one row with the ENSEMBL regulatory regions overlapping the genomic coordinate.

GRset 21

GRset GRset

Description

A small GenomicRatioSet object to use in the functions examples containing 10 control samples and a case sample.

Usage

```
data(GRset)
```

Format

A GenomicRatioSet object with 4243 CpGs and 11 variables

Examples

```
data(GRset)
```

merge_records

Merge records for the same ENSEMBL regulatory element

Description

This function collapses the activity status of a given an ENSEMBL regulatory element in different tissues. Notice that tissues identified as inactive will not be reported.

Usage

```
merge_records(tab)
```

Arguments

tab

Results from biomaRt::getBM for the same regulatory element

Value

data.frame of one row after collapsing the

22 mlm

mlm

Non-parametric, Asymptotic P-values for Multivariate Linear Models

Description

Fits a multivariate linear model and computes test statistics and asymptotic P-values for predictors in a non-parametric manner.

Usage

```
mlm(
  formula,
  data,
  transform = "none",
  contrasts = NULL,
  subset = NULL,
  fit = FALSE
)
```

Arguments

formula	object of class "formula" (or one that can be coerced to that class): a symbolic description of the model to be fitted.
data	an optional data frame, list or environment (or object coercible by as.data.frame to a data frame) containing the variables in the model. If not found in data, the variables are taken from environment(formula), typically the environment from which mlm is called.
transform	transformation of the response variables: "none", "sqrt" or "log". Default is "none".
contrasts	an optional list. See contrasts.arg in model.matrix.default. Default is "contr.sum" for ordered factors and "contr.poly" for unordered factors. Note that this is different from the default setting in options("contrasts").
subset	subset of predictors for which summary statistics will be reported. Note that this is different from the "subset" argument in 1m.
fit	logical. If TRUE the multivariate fit on transformed and centered responses is returned.

Details

A Y matrix is obtained after transforming (optionally) and centering the original response variables. Then, the multivariate fit obtained by lm can be used to compute sums of squares, pseudo-F statistics and asymptotic P-values for the terms specified by the formula in a non-parametric manner.

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Value

mlm returns an object of class "MLM", a list containing:

call the matched call.

aov.tab ANOVA table with Df, Sum Sq, Mean Sq, F values, partial R-squared and P-

values.

precision the precision in P-value computation.

transform the transformation applied to the response variables.

na.omit incomplete cases removed (see na.omit).

fit if fit = TRUE the multivariate fit done on the transformed and centered response

variables is also returned.

Author(s)

Diego Garrido-Martín

See Also

1m, Anova

norm_parameters

Settings for parameters of epi_preprocess function

Description

norm_parameters function allows the user to set the values of the parameters to compute the functions epi_preprocess.

Usage

```
norm_parameters(
  illumina = list(bg.correct = TRUE, normalize = c("controls", "no"), reference = 1),
  quantile = list(fixOutliers = TRUE, removeBadSamples = FALSE, badSampleCutoff = 10.5,
   quantileNormalize = TRUE, stratified = TRUE, mergeManifest = FALSE, sex = NULL),
  noob = list(offset = 15, dyeCorr = TRUE, dyeMethod = c("single", "reference")),
  funnorm = list(nPCs = 2, sex = NULL, bgCorr = TRUE, dyeCorr = TRUE, keepCN = FALSE)
)
```

Arguments

```
illumina, quantile, noob, funnorm
```

preprocess method selected in the function epi_preprocess.

bg.correct logical. If TRUE background correction will be performed in "illumina"

method. Default is TRUE.

normalize logical. If TRUE control normalization will be performed in "illumina" method.

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reference numeric. The reference array for control normalization in "illumina" method. fixOutliers logical. If TRUE low outlier Meth and Unmeth signals will be fixed in "quantile"

method. Default is TRUE.

removeBadSamples

logical. If TRUE bad samples will be removed.

badSampleCutoff

a numeric specifying the cutoff to label samples as 'bad' in "quantile" method.

Default is 10.5.

quantileNormalize

logical. If TRUE quantile normalization will be performed in "quantile"

method. Default is TRUE.

stratified logical. If TRUE quantile normalization will be performed within region strata

in "quantile" method. Default is TRUE.

mergeManifest logical. If TRUE the information in the associated manifest package will be

merged into the output object in "quantile" method. Default is FALSE.

offset a numeric specifying an offset for the normexp background correction in "noob"

method. Default is 15.

dyeCorr logial. Dye correction will be done in "noob" and "funnorm" methods. Default

is TRUE.

dyeMethod specify the dye bias correction to be done, single sample approach or a reference

array in "noob" method.

nPCs numeric specifying the number of principal components from the control probes

PCA in "funnorm" method. Default is 2.

sex an optional numeric vector containing the sex of the samples in "quantile" and

"funnorm" methods.

bgCorr logical. If TRUE NOOB background correction will be done prior to functional

normalization. in "funnorm" method. Default is TRUE.

keepCN logical. If TRUE copy number estimates will be kept in "funnorm" method.

Default is FALSE.

Details

Invoking epi_parameters() with no arguments returns a list with the default values for each normalization parameter.

Value

the function returns a list of all set parameters for each normalization method used in epi_peprocess.

Examples

```
#Default set of parameters
norm_parameters()
#change p value for manova method
norm_parameters(illumina = list("bg.correct" = FALSE))
```

plot_epimutations 25

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Plot a given epimutation and locate it along the genome

Description

This function plots a given epimutation and UCSC annotations for the specified genomic region.

Usage

```
plot_epimutations(
  dmr,
  methy,
  genome = "hg19",
  genes_annot = FALSE,
  regulation = FALSE,
  from = NULL,
  to = NULL
)
```

Arguments

dmr	epimutation obtained as a result of epimutations function.
methy	a GenomicRatioSet object containing the information of control and case samples used for the analysis in the epimutations function. See the constructor function GenomicRatioSet, makeGenomicRatioSetFromMatrix.
genome	a character string specifying the genome of reference. It can be set as "hg38", "hg19" and "hg18". The default is "hg19".
genes_annot	a boolean. If TRUE gene annotations are plotted. Default is FALSE.
regulation	a boolean. If TRUE UCSC annotations for CpG Islands, H3K27Ac, H3K4Me3 and H3K27Me3 are plotted. The default is FALSE. The running process when regulation is TRUE can take several minutes.
from, to	scalar, specifying the range of genomic coordinates for the plot of gene annotation region. If NULL the plotting ranges are derived from the individual track. Note that from cannot be larger than to.

Details

The tracks are plotted vertically. Each track is separated by different background colour and a section title. The colours and titles are preset and cannot be set by the user.

Note that if you want to see the UCSC annotations maybe you need to take a bigger genomic region.

Value

The function returns a plot divided in two parts:

- ggplot graph including the individual with the epimutation in red, the control samples in dashed black lines and population mean in blue. Grey shaded regions indicate 1, 1.5 and 2 standard deviations from the mean of the distribution.
- UCSC gene annotations for the specified genomic region (if genes == TRUE)
- UCSC annotations for CpG Islands, H3K27Ac, H3K4Me3 and H3K27Me3 (if regulation == TRUE)

Examples

```
data(GRset)
data(res.epi.manova)
plot_epimutations(res.epi.manova[1,], GRset)
```

```
process_ENSEMBL_results
```

Process data from ENSEMBL

Description

Process data from ENSEMBL to combine results from the same regulatory elements in a unique record.

Usage

```
process_ENSEMBL_results(ensembl_res)
```

Arguments

```
ensembl_res Results from biomaRt::getBM
```

Value

data.frame of one row after collapsing the input ENSEMBL regulatory regions

res.epi.manova 27

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res.epi.manova

Description

A data frame containing the results of epimutations function using "manova" methods for GRset dataset. For more information see the example of epimutations function.

Usage

```
data(res.epi.manova)
```

Format

A data frame with 16 variables and 6 epimutations.

Examples

```
data(res.epi.manova)
```

res_iForest

Creates a data frame containing the results obtained from Isolation Forest

Description

Creates a data frame containing the genomic regions, statistics and direction for the DMRs.

Usage

```
res_iForest(bump, sts, outlier_score_cutoff)
```

Arguments

```
bump a DMR obtained from bumphunter (i.e. a row from bumphunter method result).

sts the outlier score from epi_iForest function results.

outlier_score_cutoff

numeric specifying the outlier score cut off
```

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Value

The function returns a data frame containing the following information for each DMR:

- genomic ranges
- DMR base pairs
- number and name of CpGs in DMR
- statistics:
 - Outlier score
 - Outlier significance
 - Outlier direction
- Sample name

For more information about the output see epimutations.

res_mahdist Creates a data frame containing the results obtained from Robust Mahalanobis distance

Description

Creates a data frame containing the genomic regions, statistics and direction for the DMRs.

Usage

```
res_mahdist(case, bump, outliers)
```

Arguments

case a character string specifying the case sample name.

bump a DMR obtained from bumphunter (i.e. a row from bumphunter method result).

outliers the robust distance computed by epi_mahdist function results.

Value

The function returns a data frame containing the following information for each DMR:

- genomic ranges
- DMR base pairs
- number and name of CpGs in DMR
- statistics:
 - Outlier score
 - Outlier significance
 - Outlier direction
- Sample name

For more information about the output see epimutations.

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res_manova

Creates a data frame containing the results obtained from MANOVA

Description

Creates a data frame containing the genomic regions, statistics and direction for the DMRs.

Usage

```
res_manova(bump, sts)
```

Arguments

bump a DMR obtained from bumphunter (i.e. a row from bumphunter method result).

sts F statistic, Pillai and P value from epi_manova function results.

Value

The function returns a data frame containing the following information for each DMR:

- genomic ranges
- DMR base pairs
- number and name of CpGs in DMR
- statistics:
 - Outlier score
 - Outlier significance
 - Outlier direction
- Sample name

For more information about the output see epimutations.

res_mlm

Creates a data frame containing the results obtained from MLM

Description

Creates a data frame containing the genomic regions, statistics and direction for the DMRs.

Usage

```
res_mlm(bump, sts)
```

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Arguments

bump a DMR obtained from bumphunter (i.e. a row from bumphunter method result).

sts the F statistic, R2 test statistic and Pillai obtained as a result of epi_mlm function.

Value

The function returns a data frame containing the following information for each DMR:

- genomic ranges
- DMR base pairs
- number and name of CpGs in DMR
- statistics:
 - Outlier score
 - Outlier significance
 - Outlier direction
- Sample name

For more information about the output see epimutations.

UCSC_annotation

UCSC gene annotations

Description

UCSC gene annotations for a given genome assembly.

Usage

```
UCSC_annotation(genome = "hg19")
```

Arguments

genome genome asambly. Can be set as: 'hg38', 'hg19' and 'hg18'.

Value

The function returns gene annotations for the specified genome assembly.

UCSC_regulation 31

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Description

UCSC annotations for CpG Islands, H3K27Ac and H3K4Me3 for a given genome assembly and genomic coordinates.

Usage

```
UCSC_regulation(genome, chr, from, to)
```

Arguments

genome genome asambly. Can be set as: 'hg38', 'hg19' and 'hg18'.

chr a character string containing the sequence names to be analysed.

from, to scalar, specifying the range of genomic coordinates. Note that from cannot be

larger than to.

Value

UCSC_regulation returns a list containing CpG Islands, H3K27Ac and H3K4Me3 tacks.

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