## Package: coMethDMR (via r-universe)

October 3, 2024

**Title** Accurate identification of co-methylated and differentially methylated regions in epigenome-wide association studies

Version 1.9.0

Description coMethDMR identifies genomic regions associated with continuous phenotypes by optimally leverages covariations among CpGs within predefined genomic regions. Instead of testing all CpGs within a genomic region, coMethDMR carries out an additional step that selects co-methylated sub-regions first without using any outcome information. Next, coMethDMR tests association between methylation within the sub-region and continuous phenotype using a random coefficient mixed effects model, which models both variations between CpG sites within the region and differential methylation simultaneously.

Depends R (>= 4.1) License GPL-3 Encoding UTF-8 LazyData false

RoxygenNote 7.1.2

**Imports** AnnotationHub, BiocParallel, bumphunter, ExperimentHub, GenomicRanges, IRanges, ImerTest, methods, stats, utils

**Suggests** BiocStyle, corrplot, knitr, rmarkdown, testthat, IlluminaHumanMethylation450kanno.ilmn12.hg19, IlluminaHumanMethylationEPICanno.ilm10b4.hg19

**biocViews** DNAMethylation, Epigenetics, MethylationArray, DifferentialMethylation, GenomeWideAssociation

VignetteBuilder knitr

 ${\bf BugReports}\ {\tt https://github.com/TransBioInfoLab/coMethDMR/issues}$ 

 ${\bf URL} \ \, {\tt https://github.com/TransBioInfoLab/coMethDMR}$ 

Repository https://bioc.r-universe.dev

RemoteUrl https://github.com/bioc/coMethDMR

RemoteRef HEAD

**RemoteSha** dcdc135041d03b4fdc99c59f7cc7708f2f25c7ae

2 AnnotateResults

## **Contents**

AnnotateResults	2
betaMatrix_ex1	4
betaMatrix_ex2	4
betaMatrix_ex3	5
betaMatrix_ex4	5
betasChr22_df	6
CloseBySingleRegion	6
CoMethAllRegions	7
CoMethSingleRegion	9
CpGsInfoAllRegions	11
CpGsInfoOneRegion	12
CreateParallelWorkers	14
CreateRdrop	15
FindComethylatedRegions	16
GetCpGsInRegion	16
GetResiduals	17
ImportSesameData	19
lmmTest	19
ImmTestAllRegions	21
MarkComethylatedCpGs	24
MarkMissing	25
NameRegion	27
OrderCpGsByLocation	27
pheno_df	28
RegionsToRanges	29
WriteCloseByAllRegions	30
	32

AnnotateResults

Annotate coMethDMR Pipeline Results

## Description

Index

Given a data frame with regions in the genome, add gene symbols, UCSC reference gene accession, UCSC reference gene group and relation to CpG island.

## Usage

```
AnnotateResults(lmmRes_df, arrayType = c("450k", "EPIC"), nCores_int = 1L, ...)
```

AnnotateResults 3

## Arguments

A data frame returned by lmmTestAllRegions. This data frame must contain the following columns:

• chrom: the chromosome the region is on, e.g. "chr22"

• start: the region start point

• end: the region end point

arrayType

Type of array: 450k or EPIC

Number of computing cores to be used when executing code in parallel. Defaults to 1 (serial computing).

Dots for additional arguments passed to the cluster constructor. See CreateParallelWorkers for more information.

#### **Details**

```
The region types include "NSHORE", "NSHELF", "SSHORE", "SSHELF", "TSS1500", "TSS200", "UTR5", "EXON1", "GENEBODY", "UTR3", and "ISLAND".
```

#### Value

A data frame with

- the location of the genomic region's chromosome (chrom), start (start), and end (end);
- UCSC annotation information (UCSC\_RefGene\_Group, UCSC\_RefGene\_Accession, and UCSC\_RefGene\_Name);
   and
- a list of all of the probes in that region (probes).

```
lmmResults_df <- data.frame(
  chrom = c("chr22", "chr22", "chr22", "chr22", "chr22"),
  start = c("39377790", "50987294", "19746156", "42470063", "43817258"),
  end = c("39377930", "50987527", "19746368", "42470223", "43817384"),
  regionType = c("TSS1500", "EXON1", "ISLAND", "TSS200", "ISLAND"),
  stringsAsFactors = FALSE
)

AnnotateResults(
  lmmRes_df = lmmResults_df,
  arrayType = "450k"
)</pre>
```

betaMatrix\_ex2

betaMatrix\_ex1

Alzheimer's Prefrontal Cortex (PFC) Methylation Data

#### **Description**

Subset of an Alzheimer's Disease methylation data set, with beta values for measured CpGs methylation levels.

#### Usage

```
data("betaMatrix_ex1")
```

#### **Format**

A data frame containing beta values for 4 CpGs in one CpG islands for 110 subjects. Each column is a CpG, each row is a sample.

#### **Source**

GEO accession: GSE59685

betaMatrix\_ex2

Alzheimer's Prefrontal Cortex (PFC) Methylation Data

## Description

Subset of an Alzheimer's Disease methylation data set, with beta values for measured CpGs methylation levels.

## Usage

```
data("betaMatrix_ex2")
```

#### **Format**

A data frame containing beta values for 4 CpGs in one CpG islands for 110 subjects. Each column is a CpG, each row is a sample.

#### **Source**

GEO accession: GSE59685

betaMatrix\_ex3 5

betaMatrix\_ex3

Alzheimer's Prefrontal Cortex (PFC) Methylation Data

#### **Description**

Subset of an Alzheimer's Disease methylation data set, with beta values for measured CpGs methylation levels.

#### Usage

```
data("betaMatrix_ex3")
```

#### **Format**

A data frame containing beta values for 6 CpGs in one CpG islands for 110 subjects. Each column is a CpG, each row is a sample.

#### **Source**

GEO accession: GSE59685

betaMatrix\_ex4

Alzheimer's Prefrontal Cortex (PFC) Methylation Data

## Description

Subset of an Alzheimer's Disease methylation data set, with beta values for measured CpGs methylation levels.

## Usage

```
data("betaMatrix_ex4")
```

#### **Format**

A data frame containing beta values for 52 CpGs in one CpG islands for 110 subjects. Each column is a CpG, each row is a sample.

#### **Source**

GEO accession: GSE59685

betasChr22\_df

Prefrontal Cortex (PFC) Methylation Data from Alzheimer's Disease subjects

#### **Description**

Subset of an Alzheimer's methylation dataset, with beta values for CpGs.

## Usage

```
data("betasChr22_df")
```

#### **Format**

A data frame containing beta values for 8552 CpGs in Chr22 for a subset of 20 subjects.

#### **Source**

GEO accession: GSE59685

CloseBySingleRegion

Extract clusters of CpGs located closely in a genomic region.

#### **Description**

Extract clusters of CpGs located closely in a genomic region.

## Usage

```
CloseBySingleRegion(
  CpGs_char,
  genome = c("hg19", "hg38"),
  arrayType = c("450k", "EPIC"),
  manifest_gr = NULL,
  maxGap = 200,
  minCpGs = 3
)
```

#### **Arguments**

CpGs\_char a list of CpG IDs

genome Human genome of reference hg19 or hg38

arrayType Type of array, 450k or EPIC

CoMethAllRegions 7

manifest\_gr A GRanges object with the genome manifest (as returned by ExperimentHub or by ImportSesameData). This function by default ignores this argument in favour of the genome and arrayType arguments.

maxGap an integer, genomic locations within maxGap from each other are placed into the same cluster

minCpGs an integer, minimum number of CpGs for the resulting CpG cluster

#### **Details**

Note that this function depends only on CpG locations, and not on any methylation data. The algorithm is based on the clusterMaker function in the bumphunter R package. Each cluster is essentially a group of CpG locations such that two consecutive locations in the cluster are separated by less than maxGap.

#### Value

a list, each item in the list is a character vector of CpG IDs located closely (i.e. in the same cluster)

#### **Examples**

```
CpGs_char <- c(
    "cg02505293", "cg03618257", "cg04421269", "cg17885402", "cg19890033",
    "cg20566587", "cg27505880"
)

cluster_ls <- CloseBySingleRegion(
    CpGs_char,
    genome = "hg19",
    arrayType = "450k",
    maxGap = 100,
    minCpGs = 3
)</pre>
```

CoMethAllRegions

Extract contiguous co-methylated genomic regions from a list of predefined genomic regions

## Description

Extract contiguous co-methylated genomic regions from a list of pre-defined genomic regions

## Usage

```
CoMethAllRegions(
  dnam,
  betaToM = FALSE,
  method = c("pearson", "spearman"),
```

8 CoMethAllRegions

```
rDropThresh_num = 0.4,
minCpGs = 3,
genome = c("hg19", "hg38"),
arrayType = c("450k", "EPIC"),
CpGs_ls,
file = NULL,
returnAllCpGs = FALSE,
output = c("CpGs", "dataframe"),
nCores_int = 1L,
...
)
```

#### **Arguments**

dnam matrix (or data frame) of beta values, with row names = CpG IDs, column names

= sample IDs. This is typically genome-wide methylation beta values.

betaToM indicates if converting methylation beta values to mvalues

method method for computing correlation, can be "spearman" or "pearson"

rDropThresh\_num

threshold for min correlation between a cpg with sum of the rest of the CpGs

minCpGs minimum number of CpGs to be considered a "region". Only regions with more

than minCpGs will be returned.

genome Human genome of reference hg19 or hg38 arrayType Type of array, can be "450k" or "EPIC"

CpGs\_ls list where each item is a character vector of CpGs IDs. This should be CpG

probes located closely on the array.

file an RDS file with clusters of CpG locations (i.e. CpGs located closely to each

other on the genome). This file can be generated by the WriteCloseByAllRegions

function.

returnAllCpGs When there is not a contiguous comethylated region in the inputting pre-defined

region, returnAllCpGs = TRUE indicates outputting all the CpGs in the input regions (regardless of statistical significance), while returnAllCpGs = FALSE indicates not returning any CpGs not contained in comethylated clusters. De-

faults to FALSE, and we provide this option for debugging purposes only.

output a character vector of CpGs or a dataframe of CpGs along with rDrop info

nCores\_int Number of computing cores to be used when executing code in parallel. Defaults

to 1 (serial computing).

... Dots for additional arguments passed to the cluster constructor. See CreateParallelWorkers

for more information.

#### Details

There are two ways to input genomic regions for this function: (1) use CpGs\_1s argument, or (2) use file argument. Examples of these files are in /inst/extdata/ folder of the package.

CoMethSingleRegion 9

#### Value

When output = "dataframe" is selected, returns a list of data frames, each with CpG (CpG name), Chr (chromosome number), MAPINFO (genomic position), r\_drop (correlation between the CpG with rest of the CpGs), keep (indicator for co-methylated CpG), keep\_contiguous (index for contiguous comethylated subregions).

When output = "CpGs" is selected, returns a list, each item is a list of CpGs in the contiguous co-methylated subregion.

#### **Examples**

```
data(betasChr22_df)
CpGisland_ls <- readRDS(</pre>
  system.file(
    "extdata",
    "CpGislandsChr22_ex.rds",
    package = 'coMethDMR',
    mustWork = TRUE
)
coMeth_ls <- CoMethAllRegions (</pre>
  dnam = betasChr22_df,
  betaToM = TRUE,
 method = "pearson",
 CpGs_ls = CpGisland_ls,
  arrayType = "450k",
  returnAllCpGs = FALSE,
  output = "CpGs"
```

CoMethSingleRegion

Wrapper function to find contiguous and comethyalted sub-regions within a pre-defined genomic region

#### **Description**

Wrapper function to find contiguous and comethyalted sub-regions within a pre-defined genomic region

#### Usage

```
CoMethSingleRegion(
  CpGs_char,
  dnam,
```

10 CoMethSingleRegion

```
betaToM = TRUE,
  rDropThresh_num = 0.4,
  method = c("pearson", "spearman"),
  minCpGs = 3,
  genome = c("hg19", "hg38"),
  arrayType = c("450k", "EPIC"),
  manifest_gr = NULL,
  returnAllCpGs = FALSE
)
```

#### **Arguments**

CpGs\_char vector of CpGs in the inputting pre-defined genomic region.

dnam matrix (or data frame) of beta values, with row names = CpG ids, column names

= sample ids. This should include the CpGs in CpGs\_char, as well as additional

CpGs.

betaToM indicates if converting methylation beta values mvalues

rDropThresh\_num

threshold for min correlation between a cpg with sum of the rest of the CpGs

method method for computing correlation, can be "pearson" or "spearman"

minCpGs minimum number of CpGs to be considered a "region". Only regions with more

than minCpGs will be returned.

genome Human genome of reference hg19 or hg38 arrayType Type of array, can be "450k" or "EPIC"

manifest\_gr A GRanges object with the genome manifest (as returned by ExperimentHub

or by  ${\tt ImportSesameData}).$  This function by default ignores this argument in

favour of the genome and arrayType arguments.

returnAllCpGs When there is not a contiguous comethylated region in the inputing pre-defined

region, returnAllCpGs = 1 indicates outputting all the CpGs in the input region,

while returnAllCpGs = 0 indicates not returning any CpG.

#### Value

A list with two components:

- Contiguous\_Regions: a data frame with CpG (CpG ID), Chr (chromosome number), MAPINFO (genomic position), r\_drop (correlation between the CpG with rest of the CpGs), keep (indicator for co-methylated CpG), keep\_contiguous (index for contiguous comethylated subregion)
- CpGs\_subregions : lists of CpGs in each contiguous co-methylated subregion

```
data(betasChr22_df)
CpGsChr22_char <- c(
   "cg02953382", "cg12419862", "cg24565820", "cg04234412", "cg04824771",</pre>
```

CpGsInfoAllRegions 11

```
"cg09033563", "cg10150615", "cg18538332", "cg20007245", "cg23131131",
    "cg25703541"
)
CoMethSingleRegion(
    CpGs_char = CpGsChr22_char,
    dnam = betasChr22_df
)

data(betaMatrix_ex3)
CpGsEx3_char <- c(
    "cg14221598", "cg02433884", "cg07372974", "cg13419809", "cg26856676",
    "cg25246745"
)
CoMethSingleRegion(
    CpGs_char = CpGsEx3_char,
    dnam = t(betaMatrix_ex3),
    returnAllCpGs = TRUE
)</pre>
```

CpGsInfoAllRegions

Test Associations Between Regions and Phenotype

#### **Description**

Test associations of individual CpGs in multiple genomic regions with a continuous phenotype

#### Usage

```
CpGsInfoAllRegions(
   AllRegionNames_char,
   allRegions_gr = NULL,
   betas_df,
   pheno_df,
   contPheno_char,
   covariates_char,
   genome = c("hg19", "hg38"),
   arrayType = c("450k", "EPIC")
```

#### **Arguments**

AllRegionNames\_char

vector of character strings with location info for all the genomic regions. Each region should be specified in this format: "chrxx:xxxxxx-xxxxxx"

allRegions\_gr

An object of class GRanges with location information for the regions. If this argument is NULL, then the regions in AllRegionNames\_char are used. If this argument is not NULL, then region\_gr will overwrite any supplied ranges in AllRegionNames\_char.

12 CpGsInfoOneRegion

betas\_df data frame of beta values for all genomic regions, with row names = CpG IDs

amd column names = sample IDs

pheno\_df a data frame with phenotype and covariate variables, with variable "Sample" for

sample IDs.

contPheno\_char character string of the continuous phenotype to be tested against methylation

values

covariates\_char

character vector of covariate variables names

genome human genome of reference hg19 (default) or hg38

arrayType Type of array, can be "450k" or "EPIC"

#### Value

a data frame with locations of the genomic region (Region), CpG ID (cpg), chromosome (chr), position (pos), results for testing association of methylation in individual CpGs with the continuous phenotype (slopeEstimate, slopePval), UCSC\_RefGene\_Name, UCSC\_RefGene\_Accession, and UCSC\_RefGene\_Group

#### **Examples**

```
data(betasChr22_df)
data(pheno_df)
AllRegionNames_char <- c(
    "chr22:18267969-18268249",
    "chr22:18531243-18531447"
)

CpGsInfoAllRegions(
    AllRegionNames_char,
    betas_df = betasChr22_df,
    pheno_df = pheno_df,
    contPheno_char = "stage",
    covariates_char = c("age.brain", "sex")
)</pre>
```

CpGsInfoOneRegion

Test Associations Between a Region and Phenotype

#### **Description**

Test associations of individual CpGs in a genomic region with a continuous phenotype

CpGsInfoOneRegion 13

#### Usage

```
CpGsInfoOneRegion(
  regionName_char,
  region_gr = NULL,
  betas_df,
  pheno_df,
  contPheno_char,
  covariates_char = NULL,
  genome = c("hg19", "hg38"),
  arrayType = c("450k", "EPIC"),
  manifest_gr = NULL
)
```

#### **Arguments**

regionName\_char

character string of location information for a genomic region, specified in the

format of "chrxx:xxxxxx-xxxxxx"

region\_gr An object of class GRanges with location information for one region. If this

argument is NULL, then the region in regionName\_char is used.

betas\_df data frame of beta values with row names = CpG IDs, column names = sample

IDs

pheno\_df a data frame with phenotype and covariate variables, with variable "Sample" for

sample IDs.

contPheno\_char character string of the continuous phenotype to be tested against methylation

values

covariates\_char

character vector of covariate variables names

genome human genome of reference hg19 (default) or hg38

arrayType Type of array, can be "450k" or "EPIC"

manifest\_gr A GRanges object with the genome manifest (as returned by ExperimentHub

or by ImportSesameData). This function by default ignores this argument in

favour of the genome and arrayType arguments.

#### **Details**

This function implements linear models that test association between methylation values in a genomic region with a continuous phenotype. Note that methylation M values are used as regression outcomes in these models. The model for each CpG is:

```
methylation M value ~ contPheno_char + covariates_char
```

#### Value

a data frame with location of the genomic region (Region), CpG ID (cpg), chromosome (chr), position (pos), results for testing association of methylation in individual CpGs with continuous phenotype (slopeEstimate, slopePval) and annotations for the region.

14 CreateParallelWorkers

#### **Examples**

```
data(betasChr22_df)
data(pheno_df)
myRegion_gr <- RegionsToRanges("chr22:18267969-18268249")

CpGsInfoOneRegion(
  region_gr = myRegion_gr,
  betas_df = betasChr22_df,
  pheno_df = pheno_df,
  contPheno_char = "stage",
  covariates_char = c("age.brain", "sex"),
  arrayType = "450k"
)</pre>
```

CreateParallelWorkers Create a Parallel Computing Cluster

#### **Description**

This function is an operating-system agnostic wrapper for the SnowParam and MulticoreParam constructor functions.

#### Usage

```
CreateParallelWorkers(nCores, ...)
```

#### **Arguments**

nCores The number of computing cores to make available for coMethDMR computation
... Additional arguments passed to the cluster constructors.

## **Details**

This function checks the operating system and then creates a cluster of workers using the SnowParam function for Windows machines and the MulticoreParam function for non-Windows machines.

#### Value

A parameter class for use in parallel evaluation

```
workers_cl <- CreateParallelWorkers(nCores = 4)</pre>
```

CreateRdrop 15

CreateRdrop	Computes leave-one-out correlations (rDrop) for each CpG	

#### Description

Computes leave-one-out correlations (rDrop) for each CpG

#### Usage

```
CreateRdrop(data, method = c("pearson", "spearman"), use = "complete.obs")
```

#### **Arguments**

data a dataframe with rownames = sample IDs, column names = CpG IDs.

method method for computing correlation, can be "pearson" or "spearman", and is passed

to the cor function.

use method for handling missing values when calculating the correlation. Defaults

to "complete.obs" because the option "pairwise.complete.obs" only works

for Pearson correlation.

#### **Details**

An outlier CpG in a genomic region will typically have low correlation with the rest of the CpGs in a genomic region. On the other hand, in a cluster of co-methylated CpGs, we expect each CpG to have high correlation with the rest of the CpGs. The r.drop statistic is used to identify these co-methylated CpGs here.

#### Value

A data frame with the following columns:

- CpG: CpG ID
- r\_drop: The correlation between each CpG with the sum of the rest of the CpGs

```
data(betaMatrix_ex1)
CreateRdrop(data = betaMatrix_ex1, method = "pearson")
```

16 GetCpGsInRegion

FindComethylatedRegions

Find Contiguous Co-Methylated Regions

#### Description

Find contiguous comethylated regions based on output file from function MarkComethylatedCpGs

#### Usage

```
FindComethylatedRegions(CpGs_df, minCpGs_int = 3)
```

#### **Arguments**

CpGs\_df an output dataframe from function MarkComethylatedCpGs, with variables: CpG,

keep, ind, r\_drop. See details in documentation for MarkComethylatedCpGs.

minCpGs\_int an integer indicating the minimum number of CpGs for output genomic regions

#### Value

A data frame with variables ProbeID and Subregion (index for each output contiguous comethy-lated region)

#### **Examples**

```
data(betaMatrix_ex4)
CpGs_df <- MarkComethylatedCpGs(betaCluster_mat = betaMatrix_ex4)
FindComethylatedRegions(CpGs_df)</pre>
```

GetCpGsInRegion

Extract probe IDs for CpGs located in a genomic region

#### **Description**

Extract probe IDs for CpGs located in a genomic region

GetResiduals 17

#### Usage

```
GetCpGsInRegion(
  regionName_char,
  region_gr = NULL,
  genome = c("hg19", "hg38"),
  arrayType = c("450k", "EPIC"),
  manifest_gr = NULL,
  ignoreStrand = TRUE
)
```

#### **Arguments**

regionName\_char

character string with location information for one region in the format "chrxx:xxxxx-xxxxxx"

region\_gr An object of class GRanges with location information for one region. If this

argument is NULL, then the region in regionName\_char is used.

genome human genome of reference hg19 (default) or hg38

arrayType Type of array, 450k or EPIC

manifest\_gr A GRanges object with the genome manifest (as returned by ExperimentHub

or by ImportSesameData). This function by default ignores this argument in

favour of the genome and arrayType arguments.

ignoreStrand Whether strand can be ignored, default is TRUE

#### Value

vector of CpG probe IDs mapped to the genomic region

#### **Examples**

```
myRegion_gr <- RegionsToRanges("chr22:18267969-18268249")

GetCpGsInRegion(
  region_gr = myRegion_gr,
  genome = "hg19",
  arrayType = "450k",
  ignoreStrand = TRUE
)</pre>
```

GetResiduals

Get Linear Model Residuals

#### **Description**

Remove covariate effects from methylayion values by fitting probe-specific linear models

18 GetResiduals

#### Usage

```
GetResiduals(
  dnam,
  betaToM = TRUE,
  epsilon = 1e-08,
  pheno_df,
  covariates_char,
  nCores_int = 1L,
  ...
)
```

#### **Arguments**

dnam data frame or matrix of methylation values with row names = CpG IDs and

column names = sample IDs. This is often the genome-wide array data.

betaToM indicates if methylation beta values (ranging from [0, 1]) should be converted

to M values (ranging from (-Inf, Inf)). Note that if beta values are the input to

dnam, then betaToM should be set to TRUE, otherwise FALSE.

epsilon When transforming beta values to M values, what should be done to values

exactly equal to 0 or 1? The M value transformation would yield  $\neg$ Inf or Inf which causes issues in the statistical model. We thus replace all values exactly equal to 0 with 0 + epsilon, and we replace all values exactly equal to 1 with 1

- epsilon. Defaults to epsilon = 1e-08.

pheno\_df a data frame with phenotype and covariates, with variable Sample indicating

sample IDs.

covariates\_char

character vector for names of the covariate variables

nCores\_int Number of computing cores to be used when executing code in parallel. Defaults

to 1 (serial computing).

... Dots for additional arguments passed to the cluster constructor. See CreateParallelWorkers

for more information.

#### **Details**

This function fits an ordinary linear model predicting methylation values for each probe from the specified covariates. This process will be useful in scenarios where methylation values in a region or at an individual probe are known *a priori* to have differential methylation independent of the disease or condition of interest.

#### Value

output a matrix of residual values in the same dimension as dnam

```
data(betasChr22_df)
```

ImportSesameData 19

```
data(pheno_df)

GetResiduals(
    dnam = betasChr22_df[1:10, 1:10],
    betaToM = TRUE,
    pheno_df = pheno_df,
    covariates_char = c("age.brain", "sex", "slide")
)
```

ImportSesameData

Import Illumina manifests (sesameData versions)

#### **Description**

Load either the HM540 and EPIC manifests into working memory

#### Usage

```
ImportSesameData(manifest_char)
```

#### **Arguments**

 ${\tt manifest\_char}$ 

Which manifest should be loaded? Currently, this package has been tested to work with 450k and EPIC arrays for HG19 and HG38.

#### **Details**

This function assumes that the .onLoad() function has executed properly and (therefore) that the necessary data sets are in the cache.

## **Examples**

```
\label{lem:hm450k_gr} $$ \mbox{ - ImportSesameData("HM450.hg19.manifest")} $$ \mbox{ head(hm450k_gr)} $$
```

1mmTest

Fit mixed model to methylation values in one genomic region

#### Description

Fit mixed model to methylation values in one genomic region

20 ImmTest

#### Usage

```
lmmTest(
  betaOne_df,
  pheno_df,
  contPheno_char,
  covariates_char,
  modelType = c("randCoef", "simple"),
  genome = c("hg19", "hg38"),
  arrayType = c("450k", "EPIC"),
  manifest_gr = NULL,
  ignoreStrand = TRUE,
  outLogFile = NULL
)
```

#### **Arguments**

betaOne\_df matrix of beta values for one genomic region, with row names = CpG IDs and

column names = sample IDs

pheno\_df a data frame with phenotype and covariates, with variable Sample indicating

sample IDs.

contPheno\_char character string of the main effect (a continuous phenotype) to be tested for

association with methylation values in the region

covariates\_char

character vector for names of the covariate variables

modelType type of mixed model: can be randCoef for random coefficient mixed model or

simple for simple linear mixed model.

genome Human genome of reference: hg19 or hg38

arrayType Type of array: "450k" or "EPIC"

manifest\_gr A GRanges object with the genome manifest (as returned by ExperimentHub

or by ImportSesameData). This function by default ignores this argument in

favour of the genome and arrayType arguments.

ignoreStrand Whether strand can be ignored, default is TRUE

outLogFile Name of log file for messages of mixed model analysis

#### **Details**

This function implements a mixed model to test association between methylation M values in a genomic region with a continuous phenotype. In our simulation studies, we found both models shown below are conservative, so p-values are estimated from normal distributions instead of Student's *t* distributions.

```
When modelType = "randCoef", the model is:
```

```
M ~ contPheno_char + covariates_char + (1|Sample) + (contPheno_char|CpG).
```

The last term specifies random intercept and slope for each CpG. When modelType = "simple", the model is

```
M~contPheno_char + covariates_char + (1|Sample).
```

lmmTestAllRegions 21

#### Value

A dataframe with one row for association result of one region and the following columns: Estimate, StdErr, and pvalue showing the association of methylation values in the genomic region tested with the continuous phenotype supplied in contPheno\_char

#### **Examples**

```
data(betasChr22_df)
CpGsChr22_char <- c(
  "cg02953382", "cg12419862", "cg24565820", "cg04234412", "cg04824771",
  "cg09033563", "cg10150615", "cg18538332", "cg20007245", "cg23131131",
  "cg25703541"
)
coMethCpGs <- CoMethSingleRegion(CpGsChr22_char, betasChr22_df)</pre>
# test only the first co-methylated region
coMethBeta\_df <- betasChr22\_df[coMethCpGs$CpGsSubregions[[1]], ] \\
data(pheno_df)
res <- lmmTest(</pre>
 betaOne_df = coMethBeta_df,
 pheno_df,
 contPheno_char = "stage",
  covariates_char = c("age.brain", "sex"),
  modelType = "randCoef",
  arrayType = "450k",
  ignoreStrand = TRUE
)
```

 ${\tt lmmTestAllRegions}$ 

Linear Mixed Models by Region

#### **Description**

Fit mixed model to test association between a continuous phenotype and methylation values in a list of genomic regions

## Usage

```
lmmTestAllRegions(
  betas,
  region_ls,
  pheno_df,
  contPheno_char,
  covariates_char,
```

22 ImmTestAllRegions

```
modelType = c("randCoef", "simple"),
genome = c("hg19", "hg38"),
arrayType = c("450k", "EPIC"),
ignoreStrand = TRUE,
outFile = NULL,
outLogFile = NULL,
nCores_int = 1L,
...
)
```

#### **Arguments**

betas data frame or matrix of beta values for all genomic regions, with row names =

CpG IDs and column names = sample IDs. This is often the genome-wide array

data.

region\_ls a list of genomic regions; each item is a vector of CpG IDs within a genomic re-

gion. The co-methylated regions can be obtained by function CoMethAllRegions.

pheno\_df a data frame with phenotype and covariates, with variable Sample indicating

sample IDs.

contPheno\_char character string of the main effect (a continuous phenotype) to be tested for

association with methylation values in each region

covariates\_char

character vector for names of the covariate variables

modelType type of mixed model; can be randCoef for random coefficient mixed model or

simple for simple linear mixed model.

genome Human genome of reference: hg19 or hg38

arrayType Type of array: "450k" or "EPIC"

ignoreStrand Whether strand can be ignored, default is TRUE

outFile output .csv file with the results for the mixed model analysis

outLogFile log file for mixed models analysis messages

nCores\_int Number of computing cores to be used when executing code in parallel. Defaults

to 1 (serial computing).

... Dots for additional arguments passed to the cluster constructor. See CreateParallelWorkers

for more information.

#### **Details**

This function implements a mixed model to test association between methylation M values in a genomic region with a continuous phenotype. In our simulation studies, we found both models shown below are conservative, so p-values are estimated from normal distributions instead of Student's *t* distributions.

```
When modelType = "randCoef", the model is:
```

```
M~contPheno_char + covariates_char + (1|Sample) + (contPheno_char|CpG).
```

The last term specifies random intercept and slope for each CpG. When modelType = "simple", the model is

ImmTestAllRegions 23

```
M~contPheno_char + covariates_char + (1|Sample).
```

For the results of mixed models, note that if the mixed model failed to converge, p-value for mixed model is set to 1. Also, if the mixed model is singular, at least one of the estimated variance components for intercepts or slopes random effects is 0, because there isn't enough variability in the data to estimate the random effects. In this case, the mixed model reduces to a fixed effects model. The p-values for these regions are still valid.

#### Value

If outFile is NULL, this function returns a data frame as described below. If outFile is specified, this function writes a data frame in .csv format with the following information to the disk: location of the genomic region (chrom, start, end), number of CpGs (nCpGs), Estimate, Standard error (StdErr) of the test statistic, p-value and False Discovery Rate (FDR) for association between methylation values in each genomic region with phenotype (pValue).

If outLogFile is specified, this function also writes a .txt file that includes messages for mixed model fitting to the disk.

```
data(betasChr22_df)
data(pheno_df)
CpGisland_ls <- readRDS(</pre>
  system.file(
    "extdata",
    "CpGislandsChr22_ex.rds",
    package = 'coMethDMR',
    mustWork = TRUE
 )
)
coMeth_ls <- CoMethAllRegions(</pre>
  dnam = betasChr22_df,
  betaToM = TRUE,
  CpGs_ls = CpGisland_ls,
  arrayType = "450k",
  rDropThresh_num = 0.4,
  returnAllCpGs = FALSE
results_df <- lmmTestAllRegions(</pre>
  betas = betasChr22_df,
  region_ls = coMeth_ls,
  pheno_df = pheno_df,
  contPheno_char = "stage",
  covariates_char = "age.brain",
  modelType = "randCoef",
  arrayType = "450k",
  ignoreStrand = TRUE,
  # generates a log file in the current directory
```

```
# outLogFile = paste0("lmmLog_", Sys.Date(), ".txt")
)
```

MarkComethylatedCpGs Mark CpGs in contiguous and co-methylated region

#### **Description**

Mark CpGs in contiguous and co-methylated region

#### Usage

```
MarkComethylatedCpGs(
  betaCluster_mat,
  betaToM = TRUE,
  epsilon = 1e-08,
  rDropThresh_num = 0.4,
  method = c("pearson", "spearman"),
  use = "complete.obs"
)
```

#### **Arguments**

betaCluster\_mat

matrix of beta values, with rownames = sample ids and column names = CpG ids. Note that the CpGs need to be ordered by their genomic positions, this can be accomplished by the OrderCpGbyLocation function.

betaToM

indicates if beta values should be converted to M values before computing correlations. Defaults to TRUE.

epsilon

When transforming beta values to M values, what should be done to values exactly equal to 0 or 1? The M value transformation would yield -Inf or Inf which causes issues in the statistical model. We thus replace all values exactly equal to 0 with 0 + epsilon, and we replace all values exactly equal to 1 with 1 - epsilon. Defaults to epsilon = 1e-08.

rDropThresh\_num

threshold for minimum correlation between a cpg with the rest of the CpGs. Defaults to 0.4.

method

correlation method; can be "pearson" or "spearman"

use

method for handling missing values when calculating the correlation. Defaults to "complete.obs" because the option "pairwise.complete.obs" only works for Pearson correlation.

MarkMissing 25

#### **Details**

An outlier CpG in a genomic region will typically have low correlation with the rest of the CpGs in a genomic region. On the other hand, in a cluster of co-methylated CpGs, we expect each CpG to have high correlation with the rest of the CpGs. The r.drop statistic is used to identify these co-methylated CpGs here.

#### Value

A data frame with the following columns:

- CpG: CpG ID
- keep: The CpGs with keep = 1 belong to the contiguous and co-methylated region
- ind: Index for the CpGs
- r\_drop: The correlation between each CpG with the sum of the rest of the CpGs

#### **Examples**

```
data(betaMatrix_ex1)

MarkComethylatedCpGs(
  betaCluster_mat = betaMatrix_ex1,
  betaToM = FALSE,
  method = "pearson"
)
```

MarkMissing

Return Column and Row Names of Samples and Probes under the Missingness Theshold

#### **Description**

Return Column and Row Names of Samples and Probes under the Missingness Theshold

#### **Usage**

```
MarkMissing(dnaM_df, sampMissing_p = 0.5, probeMissing_p = 0.25)
```

#### **Arguments**

dnaM\_df A data frame of DNA methylation values. Samples are columns. Row names

are probe IDs.

sampMissing\_p The maximum proportion of missingness allowed in a sample. Defaults to 50%.

probeMissing\_p The maximum proportion of missingness allowed in a probe. Defaults to 25%.

26 MarkMissing

#### **Details**

Before calculating the missing proportion of samples, probes with missingness greater than the threshold are dropped first.

#### Value

A list of four entries:

- dropSamples: the column names of samples with more than sampMissing\_p percent missing values
- keepSamples: the column names of samples with less than or equal to sampMissing\_p percent missing values
- dropProbes: the row names of probes with more than probeMissing\_p percent missing values
- keepProbes: the row names of probes with less than or equal to probeMissing\_p percent missing values

```
### Setup ###
values_num <- c(</pre>
 0.1, 0.1, 0.1, 0.1, 0.1,
 0.1, 0.1, 0.1, 0.1, NA,
 0.1, 0.1, 0.1, 0.1, NA,
 0.1, 0.1, 0.1, NA,
 0.1, 0.1, 0.1, NA,
                      NA,
  0.1, 0.1, NA, NA, NA,
  0.1, 0.1, NA, NA, NA,
  0.1, NA, NA, NA, NA,
  NA, NA, NA, NA
)
values_mat <- matrix(values_num, nrow = 9, ncol = 5, byrow = TRUE)</pre>
rownames(values_mat) <- paste0("probe_0", 1:9)</pre>
colnames(values_mat) <- paste0("sample_0", 1:5)</pre>
values_df <- as.data.frame(values_mat)</pre>
### Simple Calculations ###
MarkMissing(values_df)
MarkMissing(values_df, probeMissing_p = 0.5)
MarkMissing(values_df, sampMissing_p = 0.25)
### Using the Output ###
mark_ls <- MarkMissing(values_df, probeMissing_p = 0.5)</pre>
valuesPurged_df <- values_df[ mark_ls$keepProbes, mark_ls$keepSamples ]</pre>
valuesPurged_df
```

NameRegion 27

NameRegion

Name a region with several CpGs based on its genomic location

#### **Description**

Name a region with several CpGs based on its genomic location

#### Usage

```
NameRegion(CpGsOrdered_df)
```

## Arguments

CpGsOrdered\_df dataframe with columns for Probe IDs as character (cpg), chromosome number as character (chr), and genomic location as integer (pos)

#### Value

genome location of the CpGs formatted as "chrxx:xxxxxx-xxxxxx"

## **Examples**

```
# Consider four probe IDs:
CpGs_char <- c("cg04677227", "cg07146435", "cg11632906", "cg20214853")
# After querying these four probes against an EPIC array via the
# OrderCpGsByLocation() function, we get the following data frame:
CpGsOrdered_df <- data.frame(
    chr = c("chr10", "chr10", "chr10", "chr10"),
    pos = c(100028236L, 100028320L, 100028468L, 100028499L),
    cpg = c("cg20214853", "cg04677227", "cg11632906", "cg07146435"),
    stringsAsFactors = FALSE
)
# Now, we can name the region that contains these four probes:
NameRegion(CpGsOrdered_df)</pre>
```

OrderCpGsByLocation

Order CpGs by genomic location

#### Description

Order CpGs by genomic location

28 pheno\_df

#### Usage

```
OrderCpGsByLocation(
   CpGs_char,
   genome = c("hg19", "hg38"),
   arrayType = c("450k", "EPIC"),
   manifest_gr = NULL,
   ignoreStrand = TRUE,
   output = c("vector", "dataframe")
)
```

## **Arguments**

CpGs\_char vector of CpGs

genome Human genome of reference: hg19 or hg38

arrayType Type of array: 450k or EPIC

manifest\_gr A GRanges object with the genome manifest (as returned by ExperimentHub

or by ImportSesameData). This function by default ignores this argument in

favour of the genome and arrayType arguments.

ignoreStrand Whether strand can be ignored, default is TRUE

output vector of CpGs or dataframe with CpGs, CHR, MAPINFO

#### Value

vector of CpGs ordered by location or dataframe with CpGs ordered by location (cpg), chromosome (chr), position (pos)

#### **Examples**

```
CpGs_char <- c("cg04677227", "cg07146435", "cg11632906", "cg20214853")
OrderCpGsByLocation(
   CpGs_char,
   genome = "hg19",
   arrayType = "450k",
   ignoreStrand = TRUE,
   output = "dataframe"
)</pre>
```

pheno\_df

Example phenotype data file from Prefrontal Cortex (PFC) Methylation Data of Alzheimer's Disease subjects

#### **Description**

Subset of phenotype information for Alzheimer's methylation dataset.

RegionsToRanges 29

#### Usage

```
data("pheno_df")
```

#### **Format**

A data frame containing variables for Braak stage (stage), subject.id, Batch (slide), Sex, Sample, age of brain sample (age.brain)

#### **Source**

GEO accession: GSE59685

RegionsToRanges

Convert genomic regions in a data frame to GRanges format

#### **Description**

Convert genomic regions in a data frame to GRanges format

## Usage

```
RegionsToRanges(regionName_char)
```

## **Arguments**

```
regionName_char
a character vector of regions in the format "chrxx:xxxxxx-xxxxxx"
```

#### Value

genomic regions in GRanges format

```
\label{eq:constant} $$\text{regions} < - c("\text{chr}22:19709548-19709755", "\text{chr}2:241721922-241722113"})$$$$\text{RegionsToRanges(regions)}
```

WriteCloseByAllRegions

Extract clusters of CpG probes located closely

#### **Description**

Extract clusters of CpG probes located closely

#### Usage

```
WriteCloseByAllRegions(
  fileName,
  regions,
  genome = c("hg19", "hg38"),
  arrayType = c("450k", "EPIC"),
  ignoreStrand = TRUE,
  maxGap = 200,
  minCpGs = 3,
  ...
)
```

## **Arguments**

fileName Name of the RDS file where the output genomic regions will be saved.

regions GRanges of input genomic regions

genome Human genome of reference: hg19 or hg38

arrayType Type of array: "450k" or "EPIC"

ignoreStrand Whether strand can be ignored, default is TRUE

maxGap an integer, genomic locations within maxGap from each other are placed into

the same cluster

minCpGs an integer, minimum number of CpGs for each resulting region

... Dots for internal arguments. Currently unused.

#### **Details**

For maxGap = 200 and minCpGs = 3, we have already calculated the clusters of CpGs. They are saved in folder /inst/extdata/.

#### Value

Nothing. Instead, file with the genomic regions containing CpGs located closely within each inputting pre-defined genomic region will be written to the disk

```
regions <- GenomicRanges::GRanges(
    seqnames = c("chr4", "chr6", "chr16", "chr16", "chr22", "chr19"),
    ranges = c(
        "174202697-174203520", "28226203-28227482", "89572934-89574634",
        "67232460-67234167", "38244199-38245362", "39402823-39403373"
)

# Uncomment out the example code below:
# WriteCloseByAllRegions(
# regions = regions,
# arrayType = "EPIC",
# maxGap = 50,
# minCpGs = 3,
# fileName = "closeByRegions.rds"
# )</pre>
```

# **Index**

* datasets	MulticoreParam, <i>14</i>
betaMatrix_ex1,4	
betaMatrix_ex2,4	NameRegion, 27
betaMatrix_ex3,5	OrderCpGsByLocation, 27
betaMatrix_ex4,5	order cpdsbyLocation, 27
betasChr22_df, 6	pheno_df, 28
pheno_df, 28	pheno_dr, 20
AnnotateResults, 2	RegionsToRanges, 29
betaMatrix_ex1,4	SnowParam, 14
betaMatrix_ex1, 4 betaMatrix_ex2, 4	
betaMatrix_ex3, 5	WriteCloseByAllRegions, $8,30$
betaMatrix_ex4, 5	
betasChr22_df, 6	
50 Ca30111 22_a1, 0	
CloseBySingleRegion, 6	
clusterMaker, 7	
CoMethAllRegions, 7, 22	
CoMethSingleRegion, 9	
cor, 15	
CpGsInfoAllRegions, 11	
CpGsInfoOneRegion, 12	
CreateParallelWorkers, 3, 8, 14, 18, 22	
CreateRdrop, 15	
ExperimentHub, 7, 10, 13, 17, 20, 28	
${\tt FindComethylatedRegions}, {\tt 16}$	
GetCpGsInRegion, 16	
GetResiduals, 17	
GRanges, 11, 13, 17	
${\tt ImportSesameData}, 7, 10, 13, 17, 19, 20, 28$	
1mmTest, 19	
1mmTestAllRegions, 3, 21	
MarkComethylatedCpGs, 16, 24	
MarkMissing, 25	