# Package 'coMethDMR'

October 10, 2019

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

Version 0.0.0.9001

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 (>= 3.5.0),
    IlluminaHumanMethylation450kanno.ilmn12.hg19,
    IlluminaHumanMethylationEPICanno.ilm10b2.hg19

License GPL-3

Encoding UTF-8

LazyData true

RoxygenNote 6.1.1

Imports BiocParallel,
    bumphunter,
    GenomicRanges,
    IRanges,
```

utils **Suggests** knitr, testthat

lmerTest, stats,

biocViews DNAMethylation,

Epigenetics, MethylationArray, DifferentialMethylation, GenomeWideAssociation

VignetteBuilder knitr

## R topics documented:

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AnnotateResults

Annotate coMethDMR Pipeline Results

### Description

Given a data frame with regions in the genome, add gene symbols, UCSC reference gene accession, and IDs of probes in the region.

### Usage

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

### Arguments

1mmRes\_df

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

Optionally, the data frame can also has regionType, which is a character string marking the type of genomic region tested. See details below.

arrayType

Type of array: 450k or EPIC

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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**

```
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).

#### **Examples**

```
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\_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

```
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.

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#### **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

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

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

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

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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

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

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, arrayType = c("450k", "EPIC"),
  maxGap = 200, minCpGs = 3)
```

#### **Arguments**

CpGs\_char a list of CpG IDs

arrayType Type of array, 450k or EPIC

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 clsuter 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)

```
CpGs_char <- c(
    "cg02505293", "cg03618257", "cg04421269", "cg17885402", "cg19890033",
    "cg20566587", "cg27505880"
)
cluster_ls <- CloseBySingleRegion(
    CpGs_char, arrayType = "450k", maxGap = 100, minCpGs = 3
)</pre>
```

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CoMethAllRegions	Extract contiguous co-methylated genomic regions from a list of pre-
	defined 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"), rDropThresh_num = 0.4, minCpGs = 3,
   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_nu	m
	thershold for min correlation between a cpg with sum of the rest of the CpGs
minCpGs	mininum number of CpGs to be considered a "region". Only regions with more than minCpGs will be returned.
arrayType	Type of array, can be "450k" or "EPIC"
CpGs_1s	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 inputing pre-defined region, returnAllCpGs = 1 indicates outputting all the CpGs in the input regions, while returnAllCpGs = 0 indicates not returning any CpG.
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\_ls argument (2) use file argument

examples of these files are at https://github.com/lissettegomez/coMethDMRdata

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#### 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, betaToM = TRUE,
  rDropThresh_num = 0.4, method = c("pearson", "spearman"),
  minCpGs = 3, arrayType = c("450k", "EPIC"), returnAllCpGs = FALSE)
```

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#### **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

thershold 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 mininum number of CpGs to be considered a "region". Only regions with more

than minCpGs will be returned.

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

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",
  "cg09033563", "cg10150615", "cg18538332", "cg20007245", "cg23131131",
  "cg25703541"
CoMethSingleRegion(
  CpGs_char = CpGsChr22_char,
  dnam = betasChr22_df
data(betaMatrix_ex3)
CpGsEx3_char <- c(</pre>
  "cg14221598", "cg02433884", "cg07372974", "cg13419809", "cg26856676",
  "cg25246745"
CoMethSingleRegion(
  CpGs_char = CpGsEx3_char,
  dnam = t(betaMatrix_ex3),
  returnAllCpGs = TRUE
)
```

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

### **Description**

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

#### Usage

```
CpGsInfoAllRegions(AllRegionNames_char, betas_df, pheno_df, contPheno_char,
  covariates_char, 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"
betas df
                  data frame of beta values for all genomic regions, 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.
                  character string of the continuous phenotype, to be tested against methylation
contPheno_char
                  values
covariates_char
                  character vector of covariate variables names
                  Type of array, can be "450k" or "EPIC"
arrayType
```

#### Value

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

```
data(betasChr22_df)
data(pheno_df)
AllRegionNames_char <- c("chr22:18267969-18268249", "chr22:18531243-18531447")
CpGsInfoAllRegions(
   AllRegionNames_char,
   betas_df = betasChr22_df,
   pheno_df, contPheno_char = "stage",
   covariates_char = c("age.brain", "sex")
)</pre>
```

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

#### **Description**

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

### Usage

```
CpGsInfoOneRegion(regionName_char, betas_df, pheno_df, contPheno_char,
  covariates_char, arrayType = c("450k", "EPIC"))
```

#### **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 regions

```
data(betasChr22_df)
data(pheno_df)

CpGsInfoOneRegion(
  regionName_char = "chr22:19709548-19709755",
  betas_df = betasChr22_df,
  pheno_df, contPheno_char = "stage",
  covariates_char = c("age.brain", "sex"),
```

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```
arrayType = "450k"
)

# not adjusting for covariates
CpGsInfoOneRegion(
  regionName_char = "chr22:18267969-18268249",
  betas_df = betasChr22_df,
  pheno_df, contPheno_char = "stage",
  covariates_char = NULL
)
```

CreateCpGsRegions

Create a list object with class CpGsRegions

### Description

Create a list object with class CpGsRegions

### Usage

```
CreateCpGsRegions(CpGs_ls)
```

### **Arguments**

CpGs\_ls

a list where each item is a character vector of CpGs IDs in a region. Each vector should be named with the region name.

### Value

A list object with class CpGsRegions

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CreateParallelWorkers Create a Parallel Computing Cluster

### **Description**

This function is a wrapper for the SnowParam and MulticoreParam constructor functions.

### Usage

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

#### **Arguments**

nCores The number of computing cores to amke 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

#### **Examples**

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

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"))
```

### **Arguments**

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

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

#### **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

### **Examples**

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

FindComethylatedRegions

 $Find\ contiguous\ comethy lated\ regions\ based\ on\ output\ file\ from\ function\ {\tt MarkComethylatedCpGs}$ 

### **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, indicates minimum nubmer of CpGs for output genomic regions

### Value

A data frame with variables ProbeID and Subregion (index for each output contiguous comethylated regions)

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

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GetCpGsInRegion	Extract probe IDs for CpGs located in a genomic region	

### **Description**

Extract probe IDs for CpGs located in a genomic region

### Usage

```
GetCpGsInRegion(regionName_char, arrayType = c("450k", "EPIC"))
```

### **Arguments**

regionName\_char

character string with location information for one region in this format: "chrxx:xxxxxxx-

xxxxxx"

arrayType Type of array, 450k or EPIC

#### Value

vector of CpG probe IDs mapped to the genomic region

### **Examples**

```
GetCpGsInRegion(
  regionName_char = "chr22:18267969-18268249",
  arrayType = "450k"
)
```

GetResiduals

Get Residuals

### Description

Get Residuals

### Usage

```
GetResiduals(dnam, betaToM = TRUE, pheno_df, covariates_char,
    nCores_int = 1L, ...)
```

#### **Arguments**

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

names = sample IDs. This is often the genome-wide array data. Note that if beta values are the input here, then betaToM should be set to TRUE. If mvalues are the

input, then betaToM should be set to FALSE

betaToM indicates if converting methylation beta values to mvalues

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

sample IDs.

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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.

#### Value

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

#### **Examples**

```
data(betasChr22_df)

data(pheno_df)

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

1mmTest

Fit mixed model to methylation values in one genomic region

### Description

Fit mixed model to methylation values in one genomic region

#### Usage

```
lmmTest(betaOne_df, pheno_df, contPheno_char, covariates_char,
  modelType = c("randCoef", "simple"), arrayType = c("450k", "EPIC"),
  outLogFile = NULL)
```

#### **Arguments**

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

umn 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.

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

outLogFile Name of log file for messages of mixed model analysis

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#### **Details**

This function implements a mixed model to test association between methylation values in a genomic region with a continuous phenotype.

When randCoef is selected, the model is

methylation M value ~ contPheno\_char + covariates\_char + (1|Sample) + (contPheno\_char|CpG). The last term specifies random intercept and slope for each CpG.

When simple is selected, the model is

```
methylation M value ~ contPheno_char + covariates_char + (1|Sample)
```

In our simulation studies, we found both models are conservative, so p-values are estimated from normal distributions instead of t-distributions.

#### Value

A dataframe with one row for association result of one region: Estimate, StdErr, and pvalue for the association of methylation values in the genomic region tested vs. continuous phenotype 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]], ]</pre>
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"
```

lmmTestAllRegions

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

#### **Description**

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

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#### Usage

```
lmmTestAllRegions(betas, region_ls, pheno_df, contPheno_char,
  covariates_char, modelType = c("randCoef", "simple"),
  arrayType = c("450k", "EPIC"), outFile = NULL, outLogFile = NULL,
  nCores_int = 1L, ...)
```

#### **Arguments**

٠	,	
	betas	data frame or matrix of beta values for all genomic regions, with row names = CpG IDs, 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 region. 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.
	arrayType	Type of array, can be "450k" or "EPIC"
	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 $\mbox{CreateParallelWorkers}$ for more information.

#### **Details**

This function implements a mixed model to test association between methylation values in a genomic region with a continuous phenotype.

When randCoef is selected, the model is

methylation M value  $\sim$  contPheno\_char + covariates\_char + (1|Sample) + (contPheno\_char|CpG). The last term specifies both random intercept and slope for each CpG.

When simple is selected, the model is

```
methylation M value ~ contPheno_char + covariates_char + (1|Sample)
```

In our simulation studies, we found both models are conservative, so p-values are estimated from normal distributions instead of t-distributions.

For the results of mixed models, note that

- (1) When mixed model failed to converge, p-value for mixed model is set to 1.
- (2) When 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 variabilities in data to estimate the random effects. In this case, mixed model reduces to a fixed effects model. The p-values for these regions are still valid.

#### Value

- (1) output file: a .csv file with 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).
- (2) log file: a .txt file that includes messages for mixed model fitting

#### **Examples**

```
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 <- lmmTestAllRegions(</pre>
  betas = betasChr22_df,
  region_ls = coMeth_ls,
  pheno_df,
  contPheno_char = "stage",
  covariates_char = "age.brain",
 modelType = "randCoef",
 arrayType = "450k",
  # generates a log file in the current directory
  outLogFile = paste0("lmmLog_", Sys.Date(), ".txt")
)
```

 ${\tt Mark CpGs} \quad \textit{Mark CpGs in contiguous and co-methylated region}$ 

### **Description**

Mark CpGs in contiguous and co-methylated region

#### **Usage**

```
MarkComethylatedCpGs(betaCluster_mat, betaToM = TRUE,
    rDropThresh_num = 0.4, method = c("pearson", "spearman"))
```

### **Arguments**

betaCluster\_mat

matrix of beta values, with rownames = sample ids, 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 converting to mvalues before computing correlations

rDropThresh\_num

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

method correlation method, can be pearson or spearman

#### **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

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

data(betaMatrix_ex2)
MarkComethylatedCpGs(betaCluster_mat = betaMatrix_ex2, method = "pearson")

data(betaMatrix_ex3)
MarkComethylatedCpGs(betaCluster_mat = betaMatrix_ex3, method = "pearson")

data(betaMatrix_ex4)
MarkComethylatedCpGs(betaCluster_mat = betaMatrix_ex4, rDropThresh_num = 0.6, method = "pearson")
```

NameRegion 21

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, in the format of "chrxx:xxxxxx-xxxxxx"

#### **Examples**

```
CpGs_char <- c("cg04677227", "cg07146435", "cg11632906", "cg20214853")
CpGsOrdered_df <- OrderCpGsByLocation(CpGs_char, arrayType=c("EPIC"), output = "dataframe")
NameRegion(CpGsOrdered_df)</pre>
```

OrderCpGsByLocation

Order CpGs by genomic location

### **Description**

Order CpGs by genomic location

#### Usage

```
OrderCpGsByLocation(CpGs_char, arrayType = c("450k", "EPIC"),
  output = c("vector", "dataframe"))
```

### Arguments

CpGs\_char vector of CpGs

arrayType Type of array, 450k or EPIC

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)

22 RegionsToRanges

#### **Examples**

```
CpGs_char <- c("cg04677227", "cg07146435", "cg11632906", "cg20214853")
OrderCpGsByLocation(CpGs_char, arrayType=c("EPIC"), 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.

### Usage

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

```
{\tt RegionsToRanges(regionName\_char)}
```

#### **Arguments**

```
regionName_char
```

a character vector of regions, in this format: "chrxx:xxxxxx-xxxxxx"

#### Value

genomic regions in GRanges format

```
regions = c("chr22:19709548-19709755", "chr2:241721922-241722113")
RegionsToRanges (regions)
```

WriteCloseByAllRegions

Extract clusters of CpG probes located closely

### **Description**

Extract clusters of CpG probes located closely

### Usage

```
WriteCloseByAllRegions(fileName, regions, arrayType = c("450k", "EPIC"), 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
arrayType	Type of array, can be "450k" or "EPIC"
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 already calculated the clusters of CpGs. They are saved in folder /inst/extdata/.

### Value

a file with the genomic regions containing CpGs located closely within each inputing pre-defined genomic region

```
data(regions)
WriteCloseByAllRegions(
   regions = regions, arrayType = "EPIC", maxGap = 50,
   minCpGs = 3, fileName = "closeByRegions.rds"
)
```

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