# Package 'ELMER'

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**Title** Inferring Regulatory Element Landscapes and Transcription Factor Networks Using Cancer Methylomes

Version 2.33.1

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**Description** ELMER is designed to use DNA methylation and gene expression from a large number of samples to infere regulatory element landscape and transcription factor network in primary tissue.

**Depends** R (>= 3.4.0), ELMER.data (>= 2.9.3)

License GPL-3

LazyData true

VignetteBuilder knitr

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addDistNearestTSS

Calculate the distance between probe and gene TSS

# **Description**

Calculate the distance between probe and gene TSS

# Usage

```
addDistNearestTSS(data, NearGenes, genome, met.platform, cores = 1)
```

# Arguments

data A multi Assay Experiment with both DNA methylation and gene Expression

objects

NearGenes A list or a data frame with the pairs gene probes

genome Which genome build will be used: hg38 (default) or hg19.

met.platform DNA methyaltion platform to retrieve data from: EPIC or 450K (default)

cores Number fo cores to be used. Deafult: 1

```
## Not run:
data <- ELMER:::getdata("elmer.data.example")
NearbyGenes <- GetNearGenes(
  data = data,
  probes = c("cg15924102", "cg24741609"),
  numFlankingGenes = 20
)
NearbyGenes <- addDistNearestTSS(data = data, NearGenes = NearbyGenes)
## End(Not run)</pre>
```

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addMutCol

Adds mutation information to MAE

#### **Description**

Adds mutation information to MAE

#### Usage

```
addMutCol(
  data,
  disease,
  genes,
  mutant_variant_classification = c("Frame_Shift_Del", "Frame_Shift_Ins",
    "Missense_Mutation", "Nonsense_Mutation", "Splice_Site", "In_Frame_Del",
    "In_Frame_Ins", "Translation_Start_Site", "Nonstop_Mutation")
)
```

### **Arguments**

data MAE object

disease TCGA disease (LUSC, GBM, etc)
genes list of genes to add information

mutant\_variant\_classification

List of mutant\_variant\_classification that will be consider a sample mutant or not.

# **Examples**

```
## Not run:
data <- ELMER:::getdata("elmer.data.example") # Get data from ELMER.data
data <- ELMER:::addMutCol(data, "LUSC", "TP53")
## End(Not run)</pre>
```

calcDistNearestTSS

Calculate distance from region to nearest TSS

# **Description**

Idea For a given region R linked to X genes G merge R with nearest TSS for G (multiple) this will increse nb of lines i.e R1 - G1 - TSS1 - DIST1 R1 - G1 - TSS2 - DIST2 To vectorize the code: make a granges from left and onde from right and find distance collapse the results keeping min distance for equals values

#### Usage

```
calcDistNearestTSS(links, TRange, tssAnnot)
```

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#### **Arguments**

links Links to calculate the distance

TRange Genomic coordinates for Tartget region

tssAnnot TSS annotation

#### Author(s)

Tiago C. Silva

# **Examples**

```
## Not run:
data <- ELMER:::getdata("elmer.data.example")
NearbyGenes <- GetNearGenes(
   data = data,
   probes = c("cg15924102", "cg24741609"),
   numFlankingGenes = 20
)

NearbyGenes <- ELMER:::calcDistNearestTSS(
   links = NearbyGenes,
   tssAnnot = getTSS(genome = "hg38"),
   TRange = rowRanges(getMet(data))
)

## End(Not run)</pre>
```

calculateEnrichement Calculate motif Erichment

# **Description**

Calculates fisher exact test

#### Usage

calculateEnrichement(foreground, background)

# Arguments

foreground A nsparseMatrix object in each 1 means the motif is found in a region, 0 not.

A nsparseMatrix object in each 1 means the motif is found in a region, 0 not.

A nsparseMatrix object in each 1 means the motif is found in a region, 0 not.

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```
createBigWigDNAmetArray
```

Create a bigwig file for IGV visualization of DNA methylation data (Array)

# **Description**

Create a bigwig for IGV visualization of DNA methylation data (Array)

# Usage

```
createBigWigDNAmetArray(
  data = NULL,
  genome = "hg38",
  met.platform = "450K",
  track.names = NULL,
  dir = "IGV_tracks"
)
```

# **Arguments**

data A matrix
genome Which genome build will be used: hg38 (default) or hg19.

met.platform DNA methyaltion platform to retrieve data from: EPIC or 450K (default)

track.names Provide a list of track names (.bw) otherwise the deault is the will be samples.bw

dir Which directory files will be saved

### Author(s)

Tiago Chedraoui Silva (tiagochst at gmail.com)

# **Examples**

```
## Not run:
data <- assay(getMet(ELMER:::getdata("elmer.data.example")))
createBigWigDNAmetArray(data = data, met.platform = "450K", genome = "hg38")
## End(Not run)</pre>
```

createIGVtrack

Create a junction track for IGV visualization of interection

# Description

Create a junction track for IGV visualization of interection

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#### Usage

```
createIGVtrack(
  pairs,
  met.platform = "450K",
  genome = "hg38",
  filename = "ELMER_interactions.bed",
  color.track = "black",
  track.name = "junctions",
  gene.symbol = NULL,
  all.tss = TRUE
)
```

### **Arguments**

pairs A data frame output from getPairs function

met.platform DNA methyaltion platform to retrieve data from: EPIC or 450K (default)

genome Which genome build will be used: hg38 (default) or hg19.

filename (".bed")

color.track A color for the track (i.e blue, red,#272E6A)

track.name Track name

gene. symbol Filter pairs to a single gene.

all.tss A logical. If TRUE it will link probes to all TSS of a gene (transcript level), if

FALSE it will link to the promoter region of a gene (gene level).

# Author(s)

Tiago Chedraoui Silva (tiagochst at gmail.com)

```
## Not run:
data <- ELMER:::getdata("elmer.data.example")</pre>
nearGenes <-GetNearGenes(TRange=getMet(data)[c("cg00329272","cg10097755"),],</pre>
                          geneAnnot=getExp(data))
Hypo.pair <- get.pair(data=data,</pre>
                        nearGenes=nearGenes,
                        permu.size=5,
                        group.col = "definition",
                        group1 = "Primary solid Tumor",
                        group2 = "Solid Tissue Normal",
                        raw.pvalue = 0.2,
                        Pe = 0.2,
                        dir.out="./"
                        label= "hypo")
createIGVtrack(Hypo.pair,met.platform = "450K", genome = "hg38")
## End(Not run)
```

createMAE	Construct a Multi Assay Experiment for ELMER analysis
-----------	---

# Description

This function will receive a gene expression and DNA methylation data objects and create a Multi Assay Experiment.

# Usage

```
createMAE(
   exp,
   met,
   colData,
   sampleMap,
   linearize.exp = FALSE,
   filter.probes = NULL,
   met.na.cut = 0.2,
   filter.genes = NULL,
   met.platform = "450K",
   genome = NULL,
   save = TRUE,
   save.filename,
   TCGA = FALSE
)
```

# Arguments

ехр	A Summarized Experiment with one assay, or a matrix or path of rda file only containing the data. Rownames should be either Ensembl gene id (ensembl_gene_id) or gene symbol (external_gene_name)
met	A Summarized Experiment with one assay containing beta-values, a matrix or path of rda file only containing the data.
colData	A DataFrame or data.frame of the phenotype data for all participants. Must have column primary (sample ID).
sampleMap	A DataFrame or data.frame of the matching samples and colnames of the gene expression and DNA methylation matrix. This should be used if your matrix have different columns names. This object must have following columns: assay ("DNA methylation" and "Gene expression"), primary (sample ID) and colname (names of the columns of the matrix).
linearize.exp	Take log2(exp + 1) in order to linearize relation between methylation and expression
filter.probes	A GRanges object contains the coordinate of probes which locate within promoter regions or distal feature regions such as union enhancer from REMC and FANTOM5. See get.feature.probe function.
met.na.cut	Define the percentage of NA that the line should have to remove the probes for humanmethylation platforms.
filter.genes	List of genes ensemble ids to filter from object
met.platform	DNA methylation platform "450K" or "EPIC"

genome Which is the default genome to make gene information. Options hg19 and hg38 save If TRUE, MAE object will be saved into a file named as the argument save.file if this was set, otherwise as mae\_genome\_met.platform.rda.

Name of the rda file to save the object (must end in .rda)

A logical. FALSE indicate data is not from TCGA (FALSE is default). TRUE indicates data is from TCGA and sample section will automatically filled in.

#### Value

A MultiAssayExperiment object

```
# NON TCGA example: matrices has different column names
gene.exp <- S4Vectors::DataFrame(</pre>
   sample1.exp = c("ENSG00000141510"=2.3,"ENSG00000171862"=5.4),
   sample2.exp = c("ENSG00000141510"=1.6,"ENSG00000171862"=2.3)
)
dna.met <- S4Vectors::DataFrame(</pre>
   sample1.met = c("cg14324200"=0.5,"cg23867494"=0.1),
   sample2.met = c("cg14324200"=0.3,"cg23867494"=0.9)
sample.info <- S4Vectors::DataFrame(</pre>
  primary = c("sample1", "sample2"),
  sample.type = c("Normal", "Tumor")
sampleMap <- S4Vectors::DataFrame(</pre>
    assay = c("Gene expression", "DNA methylation", "Gene expression", "DNA methylation"),
    primary = c("sample1", "sample1", "sample2"),
    colname = c("sample1.exp", "sample1.met", "sample2.exp", "sample2.met")
)
mae <- createMAE(</pre>
    exp = gene.exp,
    met = dna.met,
    sampleMap = sampleMap,
    met.platform ="450K",
    colData = sample.info,
    genome = "hg38"
)
# You can also use sample Mapping and Sample information tables from a tsv file
# You can use the createTSVTemplates function to create the tsv files
readr::write_tsv(as.data.frame(sampleMap), path = "sampleMap.tsv")
readr::write_tsv(as.data.frame(sample.info), path = "sample.info.tsv")
mae <- createMAE(</pre>
   exp = gene.exp,
   met = dna.met,
   sampleMap = "sampleMap.tsv",
   met.platform ="450K",
   colData = "sample.info.tsv",
   genome = "hg38"
)
```

```
# NON TCGA example: matrices has same column names
gene.exp <- S4Vectors::DataFrame(sample1 = c("ENSG00000141510"=2.3,"ENSG00000171862"=5.4),</pre>
                  sample2 = c("ENSG00000141510"=1.6,"ENSG00000171862"=2.3))
dna.met <- S4Vectors::DataFrame(sample1 = c("cg14324200"=0.5,"cg23867494"=0.1),</pre>
                        sample2= c("cg14324200"=0.3,"cg23867494"=0.9))
sample.info <- S4Vectors::DataFrame(primary = c("sample1","sample2"),</pre>
                                     sample.type = c("Normal", "Tumor"))
sampleMap <- S4Vectors::DataFrame(</pre>
     assay = c("Gene expression","DNA methylation","Gene expression","DNA methylation"),
    primary = c("sample1", "sample1", "sample2"),
    colname = c("sample1", "sample1", "sample2", "sample2")
)
mae <- createMAE(</pre>
  exp = gene.exp,
  met = dna.met,
  sampleMap = sampleMap,
  met.platform ="450K",
  colData = sample.info,
  genome = "hg38"
## Not run:
   # TCGA example using TCGAbiolinks
   # Testing creating MultyAssayExperiment object
   # Load library
   library(TCGAbiolinks)
   library(SummarizedExperiment)
   samples <- c(</pre>
      "TCGA-BA-4074", "TCGA-BA-4075", "TCGA-BA-4077", "TCGA-BA-5149",
       "TCGA-UF-A7JK", "TCGA-UF-A7JS", "TCGA-UF-A7JV"
   )
   #1) Get gene expression matrix
   query.exp <- GDCquery(</pre>
       project = "TCGA-HNSC",
       data.category = "Transcriptome Profiling",
       data.type = "Gene Expression Quantification",
       workflow.type = "STAR - Counts",
       barcode = samples
   GDCdownload(query.exp)
   exp.hg38 <- GDCprepare(query = query.exp)</pre>
   # DNA Methylation
   query.met <- GDCquery(</pre>
      project = "TCGA-HNSC",
      data.category = "DNA Methylation",
      data.type = "Methylation Beta Value",
      barcode = samples,
      platform = "Illumina Human Methylation 450"
   )
   GDCdownload(query.met)
   met <- GDCprepare(query = query.met)</pre>
```

```
distal.enhancer <- get.feature.probe(</pre>
      genome = "hg19"
      met.platform = "450k"
   mae.hg38 <- createMAE(</pre>
     exp = exp.hg38, met = met,
     TCGA = TRUE, genome = "hg38",
     filter.probes = distal.enhancer
   values(getExp(mae.hg38))
   # Consisering it is TCGA and not SE
   mae.hg19.test <- createMAE(</pre>
     exp = assay(exp.hg19), met = assay(met),
     TCGA = TRUE, genome = "hg19",
     filter.probes = distal.enhancer
   mae.hg38 <- createMAE(</pre>
     exp = assay(exp.hg38),
     met = assay(met),
     TCGA = TRUE,
     genome = "hg38",
     filter.probes = distal.enhancer
   values(getExp(mae.hg38))
   # Consisering it is not TCGA and SE
   # DNA methylation and gene expression Objects should have same sample names in columns
   not.tcga.exp <- exp.hg19</pre>
   colnames(not.tcga.exp) <- substr(colnames(not.tcga.exp),1,15)</pre>
   not.tcga.met <- met</pre>
   colnames(not.tcga.met) <- substr(colnames(not.tcga.met),1,15)</pre>
   phenotype.data <- data.frame(</pre>
      row.names = colnames(not.tcga.exp),
      primary = colnames(not.tcga.exp),
      samples = colnames(not.tcga.exp),
      group = c(rep("group1",4),rep("group2",4))
   distal.enhancer <- get.feature.probe(genome = "hg19",met.platform = "450k")</pre>
   mae.hg19 <- createMAE(</pre>
      exp = not.tcga.exp,
      met = not.tcga.met,
      TCGA = FALSE,
      filter.probes = distal.enhancer,
      genome = "hg19",
      colData = phenotype.data
   )
## End(Not run)
createMAE
```

createMotifRelevantTfs

Get family of transcription factors

# Description

This will output a list each TF motif and TFs that binding the motis. Multiple TFs may recognize a same motif such as TF family. The association between each motif famil and transcription factor was created using the (HOCOMOCO)[https://hocomoco11.autosome.org/human/mono?full=true] which TF structural families was created according to TFClass [@wingender2014tfclass] This data is stored as a list whose elements are motifs and contents for each element are TFs which recognize the same motif that is the name of the element. This data is used in function get.TFs in **ELMER** to identify the real regulator TF whose motif is enriched in a given set of probes and expression associate with average DNA methylation of these motif sites.

#### Usage

```
createMotifRelevantTfs(classification = "family")
```

# **Arguments**

classification Select if we will use Family classification or sub-family

#### Value

A list of TFs and its family members

createSummaryDocument Create summary document for TCGA.pipe function

# Description

This function will create a text file with the date of the last run, which aanalysis were performed, the values of the arguments so the user can keep track

# Usage

```
createSummaryDocument(
   analysis = "all",
   argument.values = "defaults",
   genome = NULL,
   mae.path = NULL,
   mode = NULL,
   direction = NULL,
   group.col = NULL,
   group1 = NULL,
   group2 = NULL,
   results.path = NULL
)
```

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#### **Arguments**

analysis Which analysis were performed

argument.values

Other argument values changed

genome Genome of reference hg38 and hg19

mae.path Where mae is stored

mode Mode "supervised" or "unsupervised" used in the analysis

direction Hypo or hyper direction

group.col Group col group1 Group 1 group2 Group 2

results.path Path where the results were saved

createMAE function

#### **Description**

This function will receive the DNA methylation and gene expression matrix and will create some examples of table for the argument colData and sampleMap used in ceeateMae function.

# Usage

```
createTSVTemplates(met, exp)
```

# **Arguments**

met DNA methylation matrix or Summarized Experiment exp Gene expression matrix or Summarized Experiment

#### **Examples**

ELMER

ELMER (Enhancer Linking by Methylation/Expression Relationships)

### **Description**

ELMER is designed to use DNA methylation and gene expression from a large number of samples to infere regulatory element landscape and transcription factor network in primary tissue.

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findMotifRegion

Use Hocomoco motif and homer to identify motifs in a given region

### **Description**

To find for each probe the know motif we will use HOMER software (http://homer.salk.edu/homer/). Homer and genome should be installed before this function is executed Step: 1 - get DNA methylation probes annotation with the regions 2 - Make a bed file from it 3 - Execute section: Finding Instance of Specific Motifs from http://homer.salk.edu/homer/ngs/peakMotifs.html to the HOCO-MOCO TF motifs Also, As HOMER is using more RAM than the available we will split the files in to 100k probes. Obs: for each probe we create a winddow of 500 bp (-size 500) around it. This might lead to false positives, but will not have false negatives. The false posives will be removed latter with some statistical tests.

### Usage

```
findMotifRegion(
  regions,
  output.filename = "mapped_motifs_regions.txt",
  region.size = NULL,
  genome = "hg38",
 nstep = 10000,
  cores = 1
)
```

#### **Arguments**

regions A GRanges object. Names will be used as the identifier.

output.filename

Final file name

region.size If NULL the motif will be mapped to the region. If set a window around its

center will be considered. For example if region.size is 500, then +-250bp round

it will be searched.

Homer genome (hg38, hg19) genome

Number of regions to evaluate in homer, the bigger, more memory it will use at nstep

each step.

A interger which defines the number of cores to be used in parallel process. cores

Default is 1: no parallel process.

```
## Not run:
 # use the center of the region and +-250bp around it
 gr0 <- GRanges(Rle(c("chr2", "chr2", "chr1", "chr3"),</pre>
                     c(1, 3, 2, 4)
                     ),
               IRanges(1:10, width=10:1)
names(gr0) \leftarrow paste0("ID",c(1:10))
 findMotifRegion(regions = gr0, region.size = 500, genome = "hg38", cores = 1)
```

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```
# use the region size itself
 gr1 <- GRanges(Rle(c("chr2", "chr2", "chr1", "chr3"), c(1, 3, 2, 4)),
                IRanges(1:10, width=sample(200:1000,10)))
 names(gr1) \leftarrow paste0("ID",c(1:10))
 findMotifRegion(regions = gr0, genome = "hg38", cores = 1)
## End(Not run)
```

get.diff.meth

Identify hypo/hyper-methylated CpG sites between two groups (i.e. normal vs tumor samples, treated vs untreated).

#### **Description**

get.diff.meth applys one-way t-test to identify the CpG sites that are significantly hypo/hypermethyalated using proportional samples (defined by minSubgroupFrac option) from group 1 and group 2. The P values will be adjusted by Benjamini-Hochberg method. Option pvalue and sig.dif will be the criteria (cutoff) for selecting significant differentially methylated CpG sites. If save is TURE, two getMethdiff.XX.csv files will be generated (see detail).

#### Usage

```
get.diff.meth(
  data,
  diff.dir = "hypo",
  cores = 1,
  mode = "unsupervised",
  minSubgroupFrac = 0.2,
  pvalue = 0.01,
  group.col,
  min.samples = 5,
  group1,
  group2,
  test = t.test,
  sig.dif = 0.3,
  dir.out = "./",
  save = TRUE
)
```

# **Arguments**

diff.dir

data A multiAssayExperiment with DNA methylation and Gene Expression data. See createMAE function.

A character can be "hypo", "hyper" or "both", showing differential methylation direction. It can be "hypo" which is only selecting hypomethylated probes (one tailed test); "hyper" which is only selecting hypermethylated probes (one tailed test); or "both" which are probes differenly methylated (two tailed test).

A interger which defines the number of cores to be used in parallel process. Default is 1: no parallel process.

cores

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mode

A character. Can be "unsupervised" or "supervised". If "supervised", the min-SubgroupFrac argument will be set to 1 to use all samples from both groups to find the differently methylated regions. The supervised mode should be used when all samples from both groups are considered homogenous (i.e. treated vs untreated, molecular subtype A vs molecular subtype B), while unsupervised mode should be used when there is at least one group with heterogenous samples (i.e tumor samples).

minSubgroupFrac

A number ranging from 0 to 1, specifying the fraction of extreme samples from group 1 and group 2 that are used to identify the differential DNA methylation. The default is 0.2 because we typically want to be able to detect a specific (possibly unknown) molecular subtype among tumor; these subtypes often make up only a minority of samples, and 20% was chosen as a lower bound for the purposes of statistical power. If you are using pre-defined group labels, such as treated replicates vs. untreated replicated, use a value of 1.0 (Supervised mode)

pvalue A number specifies the significant P value (adjusted P value by BH) threshold

Limit for selecting significant hypo/hyper-methylated probes. Default is  $0.01\ \mathrm{If}$ 

pvalue is smaller than pvalue than it is considered significant.

group.col A column defining the groups of the sample. You can view the available columns

using: colnames(MultiAssayExperiment::colData(data)).

min.samples Minimun number of samples to use in the analysis. Default 5. If you have 10

samples in one group, minSubgroupFrac is 0.2 this will give 2 samples in the

lower quintile, but then 5 will be used.

group1 A group from group.col. ELMER will run group1 vs group2. That means, if

direction is hyper, get probes hypermethylated in group 1 compared to group 2.

group2 A group from group.col. ELMER will run group1 vs group2. That means, if

direction is hyper, get probes hypermethylated in group 1 compared to group 2.

test Statistical test to be used. Options: t.test (DEFAULT), wilcox.test

sig.dif A number specifies the smallest DNA methylation difference as a cutoff for

selecting significant hypo/hyper-methylated probes. Default is 0.3.

dir.out A path specify the directory for outputs. Default is is current directory.

save A logic. When TRUE, two getMethdiff.XX.csv files will be generated (see de-

tail)

# Details

save: When save is TRUE, function will generate two XX.csv files. The first one is named getMethd-iff.hypo.probes.csv (or getMethdiff.hyper.probes.csv depends on diff.dir). The first file contains all statistic results for each probe. Based on this file, user can change different P value or sig.dir cutoff to select the significant results without redo the analysis. The second file is named getMethd-iff.hypo.probes.significant.csv (or getMethdiff.hyper.probes.significant.csv depends on diff.dir). This file contains statistic results for the probes that pass the significant criteria (P value and sig.dir). When save is FALSE, a data frame R object will be generate which contains the same information with the second file.

# Value

Statistics for all probes and significant hypo or hyper-methylated probes.

get.enriched.motif

#### References

Yao, Lijing, et al. "Inferring regulatory element landscapes and transcription factor networks from cancer methylomes." Genome biology 16.1 (2015): 1.

# **Examples**

get.enriched.motif

get.enriched.motif to identify the overrepresented motifs in a set of probes (HM450K) regions.

#### **Description**

get.enriched.motif is a function make use of Probes.motif data from **ELMER.data** package to calculate the motif enrichment Odds Ratio and 95% confidence interval for a given set of probes using fisher test function, after performing the Fisher's exact test, the results for all transcription factors are corrected for multiple testing with the Benjamini-Hochberg procedure. If save is TURE, two output files will be saved: getMotif.XX.enriched.motifs.rda and getMotif.XX.motif.enrichment.csv (see detail).

# Usage

#### **Arguments**

data A multi Assay Experiment from createMAE function. If set and probes.motif/background

probes are missing this will be used to get this other two arguments correctly. This argument is not require, you can set probes.motif and the backaground.probes

manually.

probes.motif A matrix contains motifs occurrence within probes regions. Probes.motif in

**ELMER.data** will be used if probes.motif is missing (detail see Probes.motif.hg19.450K

in ELMER.data).

probes A vector lists the name of probes to define the set of probes in which motif

enrichment OR and confidence interval will be calculated.

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min.motif.quality

Minimum motif quality score to consider. Possible valules: A, B, C, D, AS (A and S), BS (A, B and S), CS (A, B, C and S), DS (all - default) Description: Each PWM has a quality rating from A to D where A represents motifs with the highest confidence, and D motifs only weakly describe the pattern with a limited applications for quantitative analyses. Special S quality marks the single-box motifs (secondary motif). Source: http://hocomoco.autosome.ru/help#description\_quality\_score More information: http://nar.oxfordjournals.org/content/44/D1/D116.

More information: http://nar.oxfordjournals.org/content/

full#sec-8

background.probes

A vector lists name of probes which are considered as background for mo-

tif.enrichment calculation (see detail).

pvalue FDR P-value cut off (default 0.05)

lower .OR A number specifies the smallest lower boundary of 95% confidence interval for

Odds Ratio. The motif with higher lower boundary of 95% confidence interval for Odds Ratio than the number are the significantly enriched motifs (detail see

reference).

min.incidence A non-negative integer specifies the minimum incidence of motif in the given

probes set. 10 is default.

dir.out A path. Specifies the directory for outputs. Default is current directory

label A character. Labels the outputs such as "hypo", "hyper"

save If save is TURE, two files will be saved: getMotif.XX.enriched.motifs.rda and

getMotif.XX.motif.enrichment.csv (see detail).

plot.title Plot title. Default: no title.

#### **Details**

background.probes: For enhancer study, it is better to use probes within distal enhancer probes as background.probes. For promoter study, it is better to use probes within promoter regions as background.probes. Because enhancer and promoter have different CG content and harbors different clusters of TFs motif.

save: if save is TRUE, two files will be save on the disk. The first file is getMotif.XX.motif.enrichment.csv (XX depends on option label). This file reports the Odds Ratio and 95% confidence interval for these Odds Ratios which pass the significant cutoff (lower.OR and min.incidence). The second file is get-Motif.XX.enriched.motifs.rda (XX depends on option lable). This file contains a list R object with enriched motifs as name and probes containing the enriched motif as contents. This object will be used in get.TFs function. if save is FALSE, the function will return a R object which is the same with second file.

#### Value

A list contains enriched motifs with the probes regions harboring the motif.

A list (R object) with enriched motifs as name and probes containing the enriched motif as contents. And hypo.motif.enrichment.pdf plot will be generated.

# Author(s)

Lijing Yao (creator: lijingya@usc.edu)

get.feature.probe

#### References

Yao, Lijing, et al. "Inferring regulatory element landscapes and transcription factor networks from cancer methylomes." Genome biology 16.1 (2015): 1.

### **Examples**

```
probes <- c("cg00329272","cg10097755","cg08928189", "cg17153775","cg21156590",
"cg19749688", "cg12590404", "cg24517858", "cg00329272", "cg09010107",
"cg15386853", "cg10097755", "cg09247779", "cg09181054", "cg19371916")
  data <- tryCatch(ELMER:::getdata("elmer.data.example"), error = function(e) {</pre>
  data(elmer.data.example, envir = environment())
 })
bg <- rownames(getMet(data))</pre>
data(Probes.motif.hg38.450K,package = "ELMER.data")
enriched.motif <- get.enriched.motif(</pre>
   probes.motif = Probes.motif.hg38.450K,
   probes = probes,
   background.probes = bg,
   pvalue = 1,
   min.incidence = 2,
   label = "hypo"
# If the MAE is set, the background and the probes.motif will
# be automatically set
enriched.motif <- get.enriched.motif(</pre>
    data = data,
    min.motif.quality = "DS",
    probes=probes,
    pvalue = 1,
    min.incidence=2,
    label="hypo"
)
```

get.feature.probe

get.feature.probe to select probes within promoter regions or distal regions.

#### **Description**

get.feature.probe is a function to select the probes falling into distal feature regions or promoter regions.

This function selects the probes on HM450K that either overlap distal biofeatures or TSS promoter.

### Usage

```
get.feature.probe(
  feature = NULL,
  TSS,
  genome = "hg38",
  met.platform = "450K",
  TSS.range = list(upstream = 2000, downstream = 2000),
```

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```
promoter = FALSE,
rm.chr = NULL
)
```

#### **Arguments**

feature A GRange object containing biofeature coordinate such as enhancer coordinates.

If NULL only distal probes (2Kbp away from TSS will be selected) feature

option is only usable when promoter option is FALSE.

TSS A GRange object contains the transcription start sites. When promoter is FALSE,

Union.TSS in **ELMER.data** will be used for default. When promoter is TRUE, UCSC gene TSS will be used as default (see detail). User can specify their own

preference TSS annotation.

genome Which genome build will be used: hg38 (default) or hg19.

met.platform DNA methyaltion platform to retrieve data from: EPIC or 450K (default)

TSS. range A list specify how to define promoter regions. Default is upstream =2000bp and

downstream=2000bp.

promoter A logical.If TRUE, function will outut the promoter probes. If FALSE, function

will ouput the distal probes overlaping with features. The default is FALSE.

rm.chr A vector of chromosome need to be remove from probes such as chrX chrY or

chrM

#### **Details**

In order to get real distal probes, we use more comprehensive annotated TSS by both GENCODE and UCSC. However, to get probes within promoter regions need more accurate annotated TSS such as UCSC. Therefore, there are different settings for promoter and distal probe selection. But user can specify their own favorable TSS annotation. Then there won't be any difference between promoter and distal probe selection. @return A GRanges object contains the coordinate of probes which locate within promoter regions or distal feature regions such as union enhancer from REMC and FANTOM5. @usage get.feature.probe( feature, TSS, TSS.range = list(upstream = 2000, downstream = 2000), promoter = FALSE, rm.chr = NULL )

#### Value

A GRange object containing probes that satisfy selecting critiria.

```
# get distal enhancer probe
## Not run:
Probe <- get.feature.probe()

## End(Not run)
# get promoter probes
## Not run:
Probe <- get.feature.probe(promoter=FALSE)

## End(Not run)
# get distal enhancer probe remove chrX chrY
Probe2 <- get.feature.probe(rm.chr=c("chrX", "chrY"))</pre>
```

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get.pair

get.pair to predict enhancer-gene linkages.

# **Description**

get.pair is a function to predict enhancer-gene linkages using associations between DNA methylation at enhancer CpG sites and expression of 20 nearby genes of the CpG sites (see reference). Two files will be saved if save is true: getPair.XX.all.pairs.statistic.csv and getPair.XX.pairs.significant.csv (see detail).

#### Usage

```
get.pair(data,
         nearGenes,
         minSubgroupFrac = 0.4,
         permu.size = 10000,
         permu.dir = NULL,
         raw.pvalue = 0.001,
         Pe = 0.001,
         mode = "unsupervised",
         diff.dir = NULL,
         dir.out = "./",
         diffExp = FALSE,
         group.col,
         group1 = NULL,
         group2 = NULL,
         cores = 1,
         correlation = "negative",
         filter.probes = TRUE,
         filter.portion = 0.3,
         filter.percentage = 0.05,
         label = NULL,
         addDistNearestTSS = FALSE,
         save = TRUE)
```

#### **Arguments**

data

A multiAssayExperiment with DNA methylation and Gene Expression data. See createMAE function.

nearGenes

Can be either a list containing output of GetNearGenes function or path of rda file containing output of GetNearGenes function.

minSubgroupFrac

A number ranging from 0 to 1, specifying the fraction of extreme samples that define group U (unmethylated) and group M (methylated), which are used to link probes to genes. The default is 0.4 (the lowest quintile of samples is the U group and the highest quintile samples is the M group) because we typically want to be able to detect a specific (possibly unknown) molecular subtype among tumor; these subtypes often make up only a minority of samples, and 20% was chosen as a lower bound for the purposes of statistical power. If you are using predefined group labels, such as treated replicates vs. untreated replicated, use a value of 1.0 (Supervised mode).

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A number specify the times of permuation used in the unsupervised mode. De-

fault is 10000. A path where the output of permutation will be. permu.dir raw.pvalue A number specify the raw p-value cutoff for defining significant pairs. Default is 0.001. It will select the significant P value cutoff before calculating the empirical p-values. Pe A number specify the empirical p-value cutoff for defining significant pairs. Default is 0.001 mode A character. Can be "unsupervised" or "supervised". If unsupervised is set the U (unmethylated) and M (methylated) groups will be selected among all samples based on methylation of each probe. Otherwise U group and M group will set as the samples of group1 or group2 as described below: If diff.dir is "hypo, U will be the group 1 and M the group2. If diff.dir is "hyper" M group will be the group1 and U the group2. diff.dir A character can be "hypo" or "hyper", showing differential methylation direction in group 1. It can be "hypo" which means the probes are hypomethylated in group1; "hyper" which means the probes are hypermethylated in group1; This argument is used only when mode is supervised nad it should be the same value from get.diff.meth function. dir.out A path specify the directory for outputs. Default is current directory diffExp A logic. Default is FALSE. If TRUE, t test will be applied to test whether putative target gene are differentially expressed between two groups. A column defining the groups of the sample. You can view the available columns group.col using: colnames(MultiAssayExperiment::colData(data)). group1 A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2. group2 A group from group.col. ELMER will run group1 vs group2. That means, if

A interger which defines number of core to be used in parallel process. Default

is 1: don't use parallel process.

Type of correlation to evaluate (negative or positive). Negative (default) checks if hypomethylated region has a upregulated target gene. Positive checks if region hypermethylated has a upregulated target gene.

Should filter probes by selecting only probes that have at least a certain number of samples below and above a certain cut-off. See preAssociationProbeFiltering function.

direction is hyper, get probes hypermethylated in group 1 compared to group 2.

filter.portion A number specify the cut point to define binary methylation level for probe loci. Default is 0.3. When beta value is above 0.3, the probe is methylated and vice versa. For one probe, the percentage of methylated and unmethylated samples should be above filter.percentage value. Only used if filter.probes is TRUE. See preAssociationProbeFiltering function.

filter.percentage

filter.probes

cores

permu.size

Minimun percentage of samples to be considered in methylated and unmethylated for the filter.portion option. Default 5%. Only used if filter.probes is TRUE. See preAssociationProbeFiltering function.

 $\label A \ character \ labels \ the \ outputs.$  addDistNearestTSS

Calculated distance to the nearest TSS instead of gene distance. Having to calculate the distance to nearest TSS will take some time.

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save

Two files will be saved if save is true: getPair.XX.all.pairs.statistic.csv and get-Pair.XX.pairs.significant.csv (see detail).

#### Value

Statistics for all pairs and significant pairs

# Author(s)

Lijing Yao (creator: lijingya@usc.edu) Tiago C Silva (maintainer: tiagochst@usp.br)

#### References

Yao, Lijing, et al. "Inferring regulatory element landscapes and transcription factor networks from cancer methylomes." Genome biology 16.1 (2015): 1.

# **Examples**

```
data <- ELMER:::getdata("elmer.data.example")</pre>
nearGenes <- GetNearGenes(TRange=getMet(data)[c("cg00329272","cg10097755"),],</pre>
                          geneAnnot=getExp(data))
Hypo.pair <- get.pair(data=data,</pre>
                        nearGenes=nearGenes,
                        permu.size=5,
                        group.col = "definition",
                        group1 = "Primary solid Tumor",
                        group2 = "Solid Tissue Normal",
                        raw.pvalue = 0.2,
                        Pe = 0.2,
                        dir.out="./"
                        label= "hypo")
Hypo.pair <- get.pair(data = data,</pre>
                       nearGenes = nearGenes,
                       permu.size = 5,
                       raw.pvalue = 0.2,
                       Pe = 0.2,
                       dir.out = "./",
                       diffExp = TRUE,
                       group.col = "definition",
                       group1 = "Primary solid Tumor",
                       group2 = "Solid Tissue Normal",
                       label = "hypo")
```

get.permu

get.permu to generate permutation results for calculation of empirical P values for each enhancer-gene linkage.

# Description

get.permu is a function to use the same statistic model to calculate random enhancer-gene pairs. Based on the permutation value, empirical P value can be calculated for the real enhancer-gene pair (see reference).

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#### **Usage**

# Arguments

data A multiAssayExperiment with DNA methylation and Gene Expression data. See

createMAE function.

geneID A vector lists the genes' ID.

methy Index of M (methylated) group.
unmethy Index of U (unmethylated) group.

percentage A number ranges from 0 to 1 specifying the percentage of samples of group 1

and group 2 groups used to link probes to genes. Default is 0.2.

rm. probes A vector lists the probes name.

correlation Type of correlation to identify. Default is negative: look for hypomethylation

and increase target expression.

permu. size A number specify the times of permuation. Default is 10000.

permu.dir A path where the output of permuation will be.

cores A interger which defines number of core to be used in parallel process. Default

is 1: don't use parallel process.

# Value

Permutations

#### Note

Permutation is the most time consuming step. It is recommended to use multiple cores for this step. Default permutation time is 1000 which may need 12 hrs by 4 cores. However 10,000 permutations is recommended to get high confidence results. But it may cost 2 days.

### Author(s)

Lijing Yao (creator: lijingya@usc.edu) Tiago C Silva (maintainer: tiagochst@usp.br)

### References

Yao, Lijing, et al. "Inferring regulatory element landscapes and transcription factor networks from cancer methylomes." Genome biology 16.1 (2015): 1.

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# **Examples**

Get.Pvalue.p

Calculate empirical Pvalue

# **Description**

Calculate empirical Pvalue

# Usage

```
Get.Pvalue.p(U.matrix, permu)
```

# **Arguments**

U.matrix A data.frame of raw pvalue from U test. Output from .Stat.nonpara data frame of permutation. Output from .Stat.nonpara.permu

### Value

A data frame with empirical Pvalue.

get.tab

summarize MR TF as a binary table with 1 if TF was found in the analysis, 0 if not

# **Description**

summarize MR TF as a binary table with 1 if TF was found in the analysis, 0 if not

#### Usage

```
get.tab(dir, classification, top = TRUE)
```

# Arguments

dir Directory with ELMER results
classification Which columns to retrieve family or subfamily
top Consider only top 1 within each (sub)family

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#### **Examples**

```
## Not run:
dir.create("out")
dir.create("out2")
data <- tryCatch(</pre>
  ELMER:::getdata("elmer.data.example"),
  error = function(e) {
    message(e)
     data(elmer.data.example, envir = environment())
enriched.motif <- list("P53_HUMAN.H11MO.1.A"= c("cg00329272", "cg10097755", "cg08928189", "cg17153775", "cg21156590", "cg19749688", "cg12590404",
                                   "cg24517858", "cg00329272", "cg09010107", "cg15386853",
                                   "cg10097755", "cg09247779", "cg09181054"))
TF <- get.TFs(data,
               enriched.motif,
               group.col = "definition",
               group1 = "Primary solid Tumor",
               group2 = "Solid Tissue Normal",
               TFs = data.frame(
                      external_gene_name=c("TP53","TP63","TP73"),
                      ensembl_gene_id= c("ENSG00000141510",
                                           "ENSG00000073282",
                                           "ENSG00000078900"),
                      stringsAsFactors = FALSE),
                      dir.out = "out",
              label="hypo")
TF <- get.TFs(data,
               enriched.motif,
               group.col = "definition",
               group1 = "Primary solid Tumor",
               group2 = "Solid Tissue Normal",
               TFs = data.frame(
                       external_gene_name=c("TP53","TP63","TP73"),
                      ensembl_gene_id= c("ENSG00000141510",
                                           "ENSG00000073282",
                                           "ENSG00000078900"),
                      stringsAsFactors = FALSE),
                      dir.out = "out2",
             label="hypo")
 ta.family <- get.tab(dir = c("out","out2"),classification = "family")</pre>
 ta.subfamily <- get.tab(dir = c("out","out2"),classification = "subfamily")</pre>
 unlink("out")
 unlink("out2")
## End(Not run)
```

get.tabs

Creating matrix for MR TF heatmap

# **Description**

Code used to create matrix for MR TF heatmap

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# Usage

```
get.tabs(dir, classification = "family", top = TRUE)
```

# Arguments

```
dir Vector ofr directory with results
classification Consider family or subfamily
top Consider only top 1 within each (sub)family
```

# **Examples**

```
## Not run:
elmer.results <- dirname(
dir(path = "analysis",
    pattern = "*.hypo.pairs.significant.csv",
    recursive = T,
    full.names = T,
    all.files = T))
tabs <- get.tabs(dir = elmer.results, classification = "subfamily")
## End(Not run)</pre>
```

get.TFs

get.TFs to identify regulatory TFs.

#### **Description**

get.TFs is a function to identify regulatory TFs based on motif analysis and association analysis between the probes containing a particular motif and expression of all known TFs. If save is true, two files will be saved: getTF.XX.significant.TFs.with.motif.summary.csv and getTF.hypo.TFs.with.motif.pvalue.rda (see detail).

### Usage

```
get.TFs(data,
          enriched.motif,
          TFs,
          group.col,
          group1,
          group2,
          mode = "unsupervised",
          correlation = "negative",
          diff.dir = NULL,
          motif.relevant.TFs,
          minSubgroupFrac = 0.4,
          dir.out = "./",
          label = NULL,
          save.plots = FALSE,
          cores = 1,
          topTFper = 0.05,
          save = TRUE)
```

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#### **Arguments**

data A multiAssayExperiment with DNA methylation and Gene Expression data. See

createMAE function.

enriched.motif A list containing output of get.enriched.motif function or a path of XX.rda file

containing output of get.enriched.motif function.

TFs A data.frame containing TF GeneID and Symbol or a path of XX.csv file con-

taining TF GeneID and Symbol. If missing, human.TF list will be used (human.TF data in ELMER.data). For detail information, refer the reference paper.

group.col A column defining the groups of the sample. You can view the available columns

using: colnames(MultiAssayExperiment::colData(data)).

group1 A group from group.col. ELMER will run group1 vs group2. That means, if

direction is hyper, get probes hypermethylated in group 1 compared to group 2.

group2 A group from group.col. ELMER will run group1 vs group2. That means, if

direction is hyper, get probes hypermethylated in group 1 compared to group 2.

mode A character. Can be "unsupervised" or "supervised". If unsupervised is set the U

(unmethylated) and M (methylated) groups will be selected among all samples based on methylation of each probe. Otherwise U group and M group will set as the samples of group1 or group2 as described below: If diff.dir is "hypo, U will be the group 1 and M the group2. If diff.dir is "hyper" M group will be the

group1 and U the group2.

correlation Type of correlation to evaluate (negative or positive). Negative checks if hy-

 $pomethylated\ is\ upregulated.\ Positive\ if\ hypermethylated\ is\ upregulated.$ 

diff.dir A character can be "hypo" or "hyper", showing differential methylation direction

in group 1. It can be "hypo" which means the probes are hypomethylated in group1; "hyper" which means the probes are hypermethylated in group1; This argument is used only when mode is supervised nad it should be the same value

from get.diff.meth function.

motif.relevant.TFs

A list containing motif as names and relavent TFs as contents for each list element or a path of XX.rda file containing a list as above. If missing, motif.relavent.TFs will be used (motif.relavent.TFs data in ELMER.data). For de-

tail information, refer the reference paper.

minSubgroupFrac

save.plots

A number ranging from 0 to 1 specifying the percentage of samples used to create the groups U (unmethylated) and M (methylated) used to link probes to TF expression. Default is 0.4 (lowest quintile of all samples will be in the U

group and the highest quintile of all samples in the M group).

dir.out A path specifies the directory for outputs of get.pair function. Default is current

directory

label A character labels the outputs.

cores A interger which defines the number of cores to be used in parallel process.

Default is 1: no parallel process.

Create TF ranking plots?

topTFper Top ranked TF to be retrieved (default "0.05" - 5 percent)

save A logic. If save is ture, two files will be saved: getTF.XX.significant.TFs.with.motif.summary.csv

and getTF.hypo.TFs.with.motif.pvalue.rda (see detail). If save is false, a data

frame contains the same content with the first file.

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#### **Details**

save: If save is ture, two files will be saved. The first file is getTF.XX.significant.TFs.with.motif.summary.csv (XX depends on option lable). This file contain the regulatory TF significantly associate with average DNA methylation at particular motif sites. The second file is getTF.hypo.TFs.with.motif.pvalue.rda (XX depends on option label). This file contains a matrix storing the statistic results for significant associations between TFs (row) and average DNA methylation at motifs (column). If save is false, a data frame which contains the same content with the first file will be reported.

#### Value

Potential responsible TFs will be reported in a dataframe with 4 columns:

- motif: the names of motif.
- top.potential.TF.family: the highest ranking upstream TFs which are known recognized the motif. First item in potential.TFs.family
- top.potential.TF.subfamily: the highest ranking upstream TFs which are known recognized the motif. First item in potential.TFs.subfamily
- potential.TFs.family: TFs which are within top 5% list and are known recognized the motif (considering family classification).
- potential.TFs.subfamily: TFs which are within top 5% list and are known recognized the motif (considering subfamily classification).
- top\_5percent: all TFs which are within top 5% list.

#### Author(s)

Lijing Yao (creator: lijingya@usc.edu) Tiago C Silva (maintainer: tiagochst@usp.br)

# References

Yao, Lijing, et al. "Inferring regulatory element landscapes and transcription factor networks from cancer methylomes." Genome biology 16.1 (2015): 1.

```
data <- tryCatch(</pre>
  ELMER:::getdata("elmer.data.example"),
  error = function(e) {
    message(e)
     data(elmer.data.example, envir = environment())
  })
enriched.motif <- list(</pre>
   "P53_HUMAN.H11MO.1.A"= c(
      "cg00329272", "cg10097755", "cg08928189",
      "cg17153775", "cg21156590", "cg19749688", "cg12590404"
      "cg24517858", "cg00329272", "cg09010107", "cg15386853",
      "cg10097755", "cg09247779", "cg09181054"
TF <- get.TFs(
   data,
   enriched.motif,
   group.col = "definition",
   group1 = "Primary solid Tumor",
```

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```
group2 = "Solid Tissue Normal",
   TFs = data.frame(
         external_gene_name=c("TP53","TP63","TP73"),
         ensembl_gene_id= c(
           "ENSG00000141510",
           "ENSG00000073282"
           "ENSG00000078900"
           ),
           stringsAsFactors = FALSE
           ),
label = "hypo"
# This case will use Uniprot dabase to get list of Trasncription factors
TF <- get.TFs(
  data,
  group.col = "definition",
  group1 = "Primary solid Tumor",
  group2 = "Solid Tissue Normal",
  enriched.motif,
  label = "hypo"
```

get450K

get450K to download HM40K DNA methylation data for certain cancer types from TCGA website. @description get450K is a function to download latest version of HM450K DNA methylation for all samples of certain cancer types from GDC website.

# **Description**

get450K to download HM40K DNA methylation data for certain cancer types from TCGA website. @description get450K is a function to download latest version of HM450K DNA methylation for all samples of certain cancer types from GDC website.

### Usage

```
get450K(disease, basedir="./Data",filter=0.2, genome = "hg38")
```

# Arguments

disease	A character specifies the disease to download from TCGA such as BLCA
basedir	A path. Shows where the data will be stored.
filter	For each probe, the percentage of NA among the all the samples should smaller than filter.
genome	Data aligned against which genome of reference. Options: "hg38" (default)

### Value

Download all DNA methylation from HM450K level 3 data for the specified disease.

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getClinic to download clinic data for certain cancer types from TCGA website.
---

# Description

getClinic is a function to download latest version of clinic data for all samples of certain cancer types from TCGA website.

# Usage

```
getClinic(disease, basedir = "./Data")
```

# **Arguments**

disease A character specifies the disease to download from TCGA such as BLCA

basedir A path shows where the data will be stored.

# Value

Download all clinic information for the specified disease.

getExp	Get Gene expression object from MAE	
--------	-------------------------------------	--

# Description

Get Gene expression object from MAE

# Usage

```
getExp(data)
```

# **Arguments**

data A multiAssayExperiment with DNA methylation and Gene Expression data. See

 ${\tt createMAE}\ function.$ 

getGeneID

 ${\tt getExpSamples}$ 

Get Gene expression object samples from MAE

# Description

Get Gene expression object samples from MAE

# Usage

```
getExpSamples(data)
```

# Arguments

data

 $\label{lem:condition} A \ multiAssay Experiment \ with \ DNA \ methylation \ and \ Gene \ Expression \ data. \ See \ create \ MAE \ function.$ 

getGeneID

getGeneID to report gene id from symbol

# Description

getGeneID to report gene id from symbol

# Usage

```
getGeneID(data, symbol)
```

# Arguments

data A multiAssayExperiment with DNA methylation and Gene Expression data. See

 ${\tt createMAE}\ function.$ 

symbol A vector of characters which are gene symbols

# Value

The gene ID for these gene symbols

```
data <- ELMER:::getdata("elmer.data.example")
getGeneID(data, symbol="ZNF697")</pre>
```

getMet 33

getMet

Get DNA methylation object from MAE

# **Description**

Get DNA methylation object from MAE

# Usage

```
getMet(data)
```

#### **Arguments**

data

A multiAssayExperiment with DNA methylation and Gene Expression data. See createMAE function.

getMetSamples

Get DNA methylation object samples from MAE

# **Description**

Get DNA methylation object samples from MAE

# Usage

```
getMetSamples(data)
```

# Arguments

data

 $A \ multi Assay Experiment \ with \ DNA \ methylation \ and \ Gene \ Expression \ data. \ See \ create \ MAE \ function.$ 

GetNearGenes

GetNearGenes to collect nearby genes for one locus.

# **Description**

GetNearGenes is a function to collect equal number of gene on each side of one locus. It can receite either multi Assay Experiment with both DNA methylation and gene Expression matrix and the names of probes to select nearby genes, or it can receive two granges objects TRange and geneAnnot.

# Usage

```
GetNearGenes(
  data = NULL,
  probes = NULL,
  geneAnnot = NULL,
  TRange = NULL,
  numFlankingGenes = 20
)
```

34 getRandomPairs

#### **Arguments**

data A multi Assay Experiment with both DNA methylation and gene Expression

objects

probes Name of probes to get nearby genes (it should be rownames of the DNA methy-

lation object in the data argument object)

geneAnnot A GRange object or Summarized Experiment object that contains coordinates

of promoters for human genome.

TRange A GRange object or Summarized Experiment object that contains coordinates

of a list of targets loci.

numFlankingGenes

A number determines how many gene will be collected totally. Then the number devided by 2 is the number of genes collected from each side of targets (number

shoule be even) Default to 20.

#### Value

A data frame of nearby genes and information: genes' IDs, genes' symbols, distance with target and side to which the gene locate to the target.

#### References

Yao, Lijing, et al. "Inferring regulatory element landscapes and transcription factor networks from cancer methylomes." Genome biology 16.1 (2015): 1.

### **Examples**

```
geneAnnot <- getTSS(genome = "hg38")
probe <- GenomicRanges::GRanges(seqnames = c("chr1","chr2"),
range=IRanges::IRanges(start = c(16058489,236417627), end= c(16058489,236417627)),
name= c("cg18108049","cg17125141"))
names(probe) <- c("cg18108049","cg17125141")
NearbyGenes <- GetNearGenes(numFlankingGenes = 20,geneAnnot=geneAnnot,TRange=probe)</pre>
```

getRandomPairs

Get random pairs

# **Description**

This function will receive a pair gene probes and will return a random object with the following pattern, if a probe is linked to R1 and L3 genes the random pairs will be a random probes (a distal probe not in the input pairs) also linked to its R1 and L3 gene.

# Usage

```
getRandomPairs(pairs, genome = "hg38", met.platform = "450K", cores = 1)
```

getRegionNearGenes 35

### **Arguments**

pairs A data frame with probe, gene and side information. See example below.

genome Which genome build will be used: hg38 (default) or hg19.

met.platform DNA methyaltion platform to retrieve data from: EPIC or 450K (default)

cores A interger which defines the number of cores to be used in parallel process.

Default is 1: no parallel process.

#### Value

A data frame with the random linkages

# **Examples**

```
## Not run:
data <- ELMER:::getdata("elmer.data.example")</pre>
nearGenes <- GetNearGenes(TRange=getMet(data)[c("cg00329272","cg10097755"),],</pre>
                             geneAnnot=getExp(data))
pair <- get.pair(data = data,</pre>
                   group.col = "definition",
                   group1 = "Primary solid Tumor",
                   group2 = "Solid Tissue Normal",
                   mode = "supervised",
                   diff.dir = "hypo",
                   nearGenes = nearGenes,
                   permu.size = 5,
                   raw.pvalue = 0.001,
                   Pe = 0.2,
                   dir.out="./",
                   permu.dir = "permu_test",
                   label = "hypo")
## End(Not run)
pair <- data.frame(Probe = rep("cg00329272",3),</pre>
                     GeneID = c("ENSG00000116213", "ENSG00000130762", "ENSG00000149527"),
                     Sides = c("R5", "R2", "L4"))
 getRandomPairs(pair)
```

getRegionNearGenes

Identifies nearest genes to a region

# Description

Auxiliary function for GetNearGenes This will get the closest genes (n=numFlankingGenes) for a target region (TRange) based on a genome of reference gene annotation (geneAnnot). If the transcript level annotation (tssAnnot) is provided the Distance will be updated to the distance to the nearest TSS.

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#### **Usage**

```
getRegionNearGenes(
  TRange = NULL,
  numFlankingGenes = 20,
  geneAnnot = NULL,
  tssAnnot = NULL
)
```

# **Arguments**

TRange A GRange object contains coordinate of targets.

numFlankingGenes

A number determine how many gene will be collected from each

geneAnnot A GRange object contains gene coordinates of for human genome.

tssAnnot A GRange object contains tss coordinates of for human genome.

#### Value

A data frame of nearby genes and information: genes' IDs, genes' symbols,

#### Author(s)

Tiago C Silva (maintainer: tiagochst@usp.br)

### **Examples**

getRNAseq

getRNAseq to download all RNAseq data for a certain cancer type from TCGA.

# Description

getRNAseq is a function to download RNAseq data for all samples of a certain cancer type from TCGA

# Usage

```
getRNAseq(disease, basedir = "./Data", genome = "hg38")
```

getSymbol 37

#### **Arguments**

disease A character specifies disease in TCGA such as BLCA

basedir Download all RNA seq level 3 data for the specified disease.

genome Data aligned against which genome of reference. Options: "hg38" (default)

#### Value

Download all RNA seq level 3 data for the specified disease.

getSymbol getSymbol to report gene symbol from id

## **Description**

getSymbol to report gene symbol from id

## Usage

```
getSymbol(data, geneID)
```

## **Arguments**

data A multiAssayExperiment with DNA methylation and Gene Expression data. See

createMAE function.

geneID A character which is the ensembl\_gene\_id

## Value

The gene symbol for input genes.

## **Examples**

```
data <- ELMER:::getdata("elmer.data.example")
getSymbol(data, geneID="ENSG00000143067")</pre>
```

getTCGA to download DNA methylation, RNA expression and clinic

data for all samples of certain cancer type from TCGA.

## **Description**

getTCGA is a function to download DNA methylation, RNA expression and clinic data for all samples of certain cancer type from TCGA website. And downloaded data will be transform to matrixes or data frame for further analysis.

```
getTCGA(disease, Meth=TRUE, RNA=TRUE, Clinic=TRUE, basedir="./Data", genome = "hg38")
```

getTF

## Arguments

disease	A character specifies the disease to download in TCGA such as BLCA
Meth	A logic if TRUE HM450K DNA methylation data will download.
RNA	A logic if TRUE RNA-seq Hiseq-V2 from TCGA level 3 will be download.
Clinic	A logic if TRUE clinic data will be download for that disease.
basedir	A path shows where the data will be stored.
genome	Data aligned against which genome of reference. Options: "hg38" (default)

## Value

Download DNA methylation (HM450K)/RNAseq(HiseqV2)/Clinic data for the specified disease from TCGA.

# **Examples**

```
getTCGA(
    disease = "BRCA",
    Meth = FALSE,
    RNA = FALSE,
    Clinic = TRUE,
    basedir = tempdir(),
    genome = "hg38"
)
```

getTF

Get human TF list from the UNiprot database

# Description

This function gets the last version of human TF list from the UNiprot database

# Usage

```
getTF()
```

## Value

A data frame with the ensemble gene id.

getTFBindingSites 39

getTFBindingSites

Get MR TF binding regions infered by ELMER

## **Description**

Saves a bed file with the unmethylated probes  $(+-250 \mathrm{bp})$  regions that was infered to be bound by a given TF

## Usage

```
getTFBindingSites(
  tf = NULL,
  results.dir = NULL,
  genome = "hg38",
  met.platform = "450K"
)
```

## **Arguments**

tf TF name

results.dir path to the directory with the results (i.e. analysis/unsupervised/definition-Primary.solid.Tumor\_vs\_S

genome Human genome (hg38, hg19)

met.platform DNA Methylation Array platform (EPIC, 450K)

## **Examples**

getTFtargets

Get TF target genes

## **Description**

This function uses ELMER analysis results and summarizes the possible genes targets for each TF

```
getTFtargets(
  pairs,
  enriched.motif,
  TF.result,
  dmc.analysis,
  mae,
  save = TRUE,
  dir.out = "./",
```

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```
classification = "family",
  cores = 1,
  label = NULL
)
```

## **Arguments**

pairs Output of get.pairs function: dataframe or file path

enriched.motif List of probes for each enriched motif: list of file path. The file created by

ELMER is getMotif...enriched.motifs.rda

TF.result Output get.TF function: dataframe or file path

dmc.analysis DMC results file or data frame

mae A multiAssayExperiment outputed from createMAE function

save A logic. If save is true, a files will be saved: getTFtarget.XX..csv If save is false,

only a data frame contains the same content with the first file.

dir.out A path specifies the directory for outputs of get.pair function. Default is current

directory

classification use family or subfamily classification to consider potential TF

cores Number of cores to be used in parallel

label A character labels the outputs.

#### **Examples**

getTSS

getTSS to fetch GENCODE gene annotation (transcripts level) from Bioconductor package biomaRt If upstream and downstream are specified in TSS list, promoter regions of GENCODE gene will be generated.

## **Description**

getTSS to fetch GENCODE gene annotation (transcripts level) from Bioconductor package biomaRt If upstream and downstream are specified in TSS list, promoter regions of GENCODE gene will be generated.

heatmapGene 41

#### Usage

```
getTSS(genome = "hg38", TSS = list(upstream = NULL, downstream = NULL))
```

#### **Arguments**

genome Which genome build will be used: hg38 (default) or hg19.

TSS A list. Contains upstream and downstream like TSS=list(upstream, downstream).

When upstream and downstream is specified, coordinates of promoter regions

with gene annotation will be generated.

#### Value

GENCODE gene annotation if TSS is not specified. Coordinates of GENCODE gene promoter regions if TSS is specified.

#### Author(s)

```
Lijing Yao (maintainer: lijingya@usc.edu)
```

#### **Examples**

```
# get GENCODE gene annotation (transcripts level)
## Not run:
    getTSS <- getTSS()
    getTSS <- getTSS(genome.build = "hg38", TSS=list(upstream=1000, downstream=1000))
## End(Not run)</pre>
```

heatmapGene

Heatmap for correlation between probes DNA methylation and a single gene expression.

# Description

This heatmap will sort samples by their gene expression and show the DNA methylation levels of the paired probes to that gene. If no pairs are given, nearest probes will be selected. To use this function you MAE object (input data) will need all probes and not only the distal ones. This plot can be used to evaluate promoter, and intro, exons regions and closer distal probes of a gene to verify if their DNA methylation level is affecting the gene expression

```
heatmapGene(
  data,
  group.col,
  group1,
  group2,
  pairs,
  GeneSymbol,
  scatter.plot = FALSE,
  correlation.method = "pearson",
```

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```
correlation.table = FALSE,
annotation.col = NULL,
met.metadata = NULL,
exp.metadata = NULL,
dir.out = ".",
filter.by.probe.annotation = TRUE,
numFlankingGenes = 10,
width = 10,
height = 10,
scatter.plot.width = 10,
scatter.plot.height = 10,
filename = NULL
```

#### **Arguments**

data A MultiAssayExperiment with a DNA methylation SummarizedExperiment (all

probes) and a gene Expression SummarizedExperiment.

group.col A column from the sample matrix from the MultiAssayExperiment object. Ac-

cessed with colData(mae)

group1 A group from group.col. ELMER will run group1 vs group2. That means, if

direction is hyper, get probes hypermethylated in group 1 compared to group 2.

group2 A group from group.col. ELMER will run group1 vs group2. That means, if

direction is hyper, get probes hypermethylated in group 1 compared to group 2.

pairs List of probe and pair genes

Gene Symbol Gene Symbol scatter.plot Plot scatter plots

correlation.method

Correlation method: Pearson or sperman

correlation.table

save table with spearman correlation analysis?

 $annotation.col \ \ A\ vector\ of\ columns\ from\ the\ sample\ matrix\ from\ the\ MultiAssayExperiment$ 

object. Accessed with colData(mae) to be added as annotation to the heatmap

met.metadata A vector of metdatada columns available in the DNA methylation GRanges to

should be added to the heatmap.

exp.metadata A vector of metdatada columns available in the Gene expression GRanges to

should be added to the heatmap.

dir.out Where to save the plots

filter.by.probe.annotation

Filter probes to plot based on probes annotation

numFlankingGenes

numFlankingGenes to plot.

width Figure width height Figure height

scatter.plot.width

Scatter plot width

scatter.plot.height

Scatter plot height

filename File names (.pdf) to save the file (i.e. "plot.pdf"). If NULL return plot.

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#### Value

A heatmap

#### Author(s)

Tiago Chedraoui Silva (tiagochst at gmail.com)

#### **Examples**

```
## Not run:
  data <- ELMER:::getdata("elmer.data.example")</pre>
  group.col <- "subtype_Expression.Subtype"</pre>
  group1 <- "classical"</pre>
  group2 <- "secretory"</pre>
  pairs <- data.frame(ID = c("cg15924102","cg19403323", "cg22396959"),
                    GeneID = c("ENSG00000196878", "ENSG00000009790", "ENSG00000009790"),
                       Symbol = c("TRAF3IP3","LAMB3","LAMB3"),
                       Side = c("R1","L1","R3"),
                       Distance = c(6017, 168499, 0),
                       stringsAsFactors = FALSE)
 heatmapGene(data = data,
             group.col = group.col,
             group1 = group1,
             group2 = group2,
             pairs = pairs,
             GeneSymbol = "LAMB3",
             height = 5,
             annotation.col = c("ethnicity", "vital_status"),
             filename = "heatmap.pdf")
 \dontrun{
     heatmapGene(data = data,
                 group.col = group.col,
                 group1 = group1,
                 group2 = group2,
                  GeneSymbol = "ACP6",
                  annotation.col = c("ethnicity", "vital_status"),
                  filename = "heatmap_closer_probes.pdf")
 }
## End(Not run)
```

heatmapPairs

Heatmap of pairs gene and probes anti-correlated

## Description

Heatmp plot of pairs gene and probes anti-correlated

```
heatmapPairs(
  data,
  group.col,
```

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```
group1,
group2,
pairs,
subset = FALSE,
cluster.within.groups = TRUE,
plot.distNearestTSS = FALSE,
annotation.col = NULL,
met.metadata = NULL,
exp.metadata = NULL,
width = 10,
height = 7,
filename = NULL
```

## **Arguments**

data

	data	probes) and a gene Expression SummarizedExperiment.
	group.col	A column from the sample matrix from the MultiAssayExperiment object. Accessed with colData(mae)
	group1	A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2.
	group2	A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2.
	pairs	List of probe and pair genes
	subset	Subset MAE object to keep only groups compared ?
cluster.within.groups		
		Cluster columns based on the groups
plot.distNearestTSS		
		Plot track with distNearestTSS?
	annotation.col	A vector of columns from the sample matrix from the MultiAssayExperiment object. Accessed with colData(mae) to be added as annotation to the heatmap.
	met.metadata	A vector of metdatada columns available in the DNA methylation GRanges to should be added to the heatmap.
	exp.metadata	A vector of metdatada columns available in the Gene expression GRanges to

File names (.pdf) to save the file (i.e. "plot.pdf"). If NULL return plot.

A MultiAssayExperiment with a DNA methylation SummarizedExperiment (all

# Value

A heatmap

width

height

filename

# Author(s)

Tiago Chedraoui Silva (tiagochst at gmail.com)

Figure width

Figure height

should be added to the heatmap.

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#### **Examples**

```
## Not run:
  data <- ELMER:::getdata("elmer.data.example")</pre>
  group.col <- "subtype_Expression.Subtype"</pre>
  group1 <- "classical"</pre>
  group2 <- "secretory"</pre>
  pairs <- data.frame(Probe = c("cg15924102","cg19403323", "cg22396959"),</pre>
                     GeneID = c("ENSG00000196878", "ENSG00000009790", "ENSG00000009790"),
                        Symbol = c("TRAF3IP3","LAMB3","LAMB3"),
                        Distance = c(6017, 168499, 0),
                        Raw.p = c(0.001, 0.00001, 0.001),
                        Pe = c(0.001, 0.00001, 0.001))
heatmapPairs(
   data = data, group.col = group.col,
               group1 = group1, group2 = group2,
               annotation.col = c("ethnicity","vital_status","age_at_diagnosis"),
pairs, filename = "heatmap.pdf",
               height = 4, width = 11
  )
## End(Not run)
```

lm\_eqn

lable linear regression formula

## **Description**

lable linear regression formula

#### Usage

```
lm_eqn(df, Dep, Exp)
```

## **Arguments**

df A data frame object contains two variables: dependent variable (Dep) and ex-

planation variable (Exp).

Dep A character specify dependent variable. The first column will be dependent

variable as default.

Exp A character specify explanation variable. The second column will be explana-

tion variable as default.

#### Value

A linear regression formula

46 metBoxPlot

metBoxPlot	scatter.plot to plot scatter plots between gene expression and DNA methylation.

## **Description**

scatter.plot is a function to plot various scatter plots between gene expression and DNA methylation. When byPair is specified, scatter plot for individual probe-gene pairs will be generated. When byProbe is specified, scatter plots for one probes with nearby 20 gene pairs will be generated. When byTF is specified, scatter plot for TF expression and average DNA methylation at certain motif sites will be generated.

# Usage

```
metBoxPlot(
  data,
  group.col,
  group1,
  group2,
  probe,
  min.samples = 5,
  minSubgroupFrac = 0.2,
  diff.dir = "hypo",
  legend.col = NULL,
  title = NULL,
  filename = NULL,
  save = TRUE
)
```

#### **Arguments**

data	A multiAssayExperiment with DNA methylation and Gene Expression data. See createMAE function.
group.col	A column defining the groups of the sample. You can view the available columns using: colnames(MultiAssayExperiment::colData(data)).
group1	A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2.
group2	A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2.
probe	Character with probe name (i.e. "cg24517858")
min.samples	Minimun number of samples to use in the analysis. Default 5. If you have 10 samples in one group, percentage is 0.2 this will give 2 samples in the lower quintile, but then 5 will be used.
minSubgroupFra	c

A number ranges from 0 to 1 specifying the percentage of samples from group1 and group2 that are used to identify the differential methylation. Default is 0.2 because we did not expect all cases to be from a single molecular subtype.But, If you are working with molecular subtypes please set it to 1.

motif.enrichment.plot 47

diff.dir A character can be "hypo" or "hyper", showing differential methylation direction. It can be "hypo" which is only selecting hypomethylated probes; "hyper" which is only selecting hypermethylated probes;

legend.col legend title title plot title

filename File names (.png) to save the file (i.e. "plot.png")

save Save plot as PNG

#### Value

Box plot

#### Author(s)

Tiago Chedraoui Silva (tiagochst at gmail.com)

## **Examples**

motif.enrichment.plot motif.enrichment.plot to plot bar plots showing motif enrichment ORs and 95% confidence interval for ORs

## **Description**

motif.enrichment.plot to plot bar plots showing motif enrichment ORs and 95% confidence interval for ORs. Option motif.enrichment can be a data frame generated by get.enriched.motif or a path of XX.csv saved by the same function.

#### **Arguments**

motif.enrichment

A data frame or a file path of get.enriched.motif output motif.enrichment.csv

file.

significant A list to select subset of motif. Default is NULL.

dir.out A path specify the directory to which the figures will be saved. Current directory

is default.

save A logic. If true (default), figure will be saved to dir.out.

label A character. Labels the outputs figure.

title Plot title. Default: no title

width Plot width

height Plot height. If NULL a default value will be calculated

summary Create a summary table along with the plot, it is necessary to add two new

columns to object (NumOfProbes and PercentageOfProbes)

#### **Details**

motif.enrichment If input data.frame object, it should contain "motif", "OR", "lowerOR", "upperOR" columns. motif specifies name of motif; OR specifies Odds Ratio, lowerOR specifies lower boundary of OR (95 upperOR specifies upper boundary of OR(95

significant A list used to select subset of motif.enrichment by the cutoff of OR, lowerOR, upperOR. significant=list(OR=1). More than one cutoff can be specified such as significant = list(OR=1, lowerOR=1, upperOR=4)

#### Value

A figure shows the enrichment level for selected motifs.

#### Author(s)

Lijing Yao (creator: lijingya@usc.edu)

#### References

Yao, Lijing, et al. "Inferring regulatory element landscapes and transcription factor networks from cancer methylomes." Genome biology 16.1 (2015): 1.

## **Examples**

```
\label{eq:motifenerichment} $$ - \text{data.frame}(\text{motif} = \text{c}(\text{"TP53"}, \text{"NR3C1"}, \text{"E2F1"}, \text{"EBF1"}, \text{"RFX5"}, \text{"ZNF143"}, \text{"CTCF"}), \\ & \text{OR} = \text{c}(19.33, 4.83, 1, \ 4.18, \ 3.67, 3.03, 2.49), \\ & \text{lowerOR} = \text{c}(10, 3, 1.09, 1.9, 1.5, 1.9, \ 0.82), \\ & \text{upperOR} = \text{c}(23, 5, 3, 7, 6, 5, 5), \\ & \text{stringsAsFactors} = \text{FALSE}) \\ \\ \text{motif.enrichment.plot}(\text{motif.enrichment} = \text{motif.enrichment}, \\ & \text{significant} = \text{list}(\text{OR} = 3), \\ & \text{label} = \text{"hypo"}, \text{ save} = \text{FALSE}) \\ \\ \text{motif.enrichment.plot}(\text{motif.enrichment} = \text{motif.enrichment}, \\ & \text{significant} = \text{list}(\text{OR} = 3), \\ & \text{label} = \text{"hypo"}, \\ & \text{title} = \text{"OR} \text{ for paired probes hypomethylated in Mutant vs WT"}, \\ \end{aligned}
```

```
save = FALSE)
motif.enrichment <- data.frame(motif = c("TP53","NR3C1","E2F1","EBF1","RFX5","ZNF143", "CTCF"),</pre>
                                OR = c(19.33, 4.83, 1, 4.18, 3.67, 3.03, 2.49),
                                lowerOR = c(10,3,1.09,1.9,1.5,1.5, 0.82),
                                upperOR = c(23,5,3,7,6,5,5)
                                NumOfProbes = c(23,5,3,7,6,5,5),
                              PercentageOfProbes = c(0.23, 0.05, 0.03, 0.07, 0.06, 0.05, 0.05),
                                stringsAsFactors=FALSE)
motif.enrichment.plot(motif.enrichment = motif.enrichment,
                       significant = list(OR = 3),
                       label = "hypo", save = FALSE)
motif.enrichment.plot(motif.enrichment = motif.enrichment,
                       significant = list(OR = 3),
                       label = "hypo",
                       summary = TRUE,
                       title = "OR for paired probes hypomethylated in Mutant vs WT",
                       save = TRUE)
```

preAssociationProbeFiltering

Filtering probes

#### **Description**

This function has some filters to the DNA methylation data in each it selects probes to avoid correlations due to non-cancer contamination and for additional stringency.

• Filter 1: We usually call locus unmethylated when the methylation value < 0.3 and methylated when the methylation value > 0.3. Therefore Meth\_B is the percentage of methylation value > K. Basically, this step will make sure we have at least a percentage of beta values lesser than K and n percentage of beta values greater K. For example, if percentage is 5%, the number of samples 100 and K = 0.3, this filter will select probes that we have at least 5 (5% of 100%) samples have beta values > 0.3 and at least 5 samples have beta values < 0.3. This filter is importante as true promoters and enhancers usually have a pretty low value (of course purity can screw that up). we often see lots of PMD probes across the genome with intermediate values like 0.4. Choosing a value of 0.3 will certainly give some false negatives, but not compared to the number of false positives we thought we might get without this filter.

## Usage

```
preAssociationProbeFiltering(data, K = 0.3, percentage = 0.05)
```

## **Arguments**

data A MultiAssayExperiment with a DNA methylation martrix or a DNA methyla-

tion matrix

K Cut off to consider probes as methylated or unmethylated. Default: 0.3

percentage The percentage of samples we should have at least considered as methylated and

unmethylated

## Value

An object with the same class, but with the probes removed.

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#### References

Yao, Lijing, et al. "Inferring regulatory element landscapes and transcription factor networks from cancer methylomes." Genome biology 16.1 (2015): 1. Method section (Linking enhancer probes with methylation changes to target genes with expression changes).

#### **Examples**

```
random.probe <- runif(100, 0, 1)</pre>
bias_1.probe <- runif(100, 0, 0.3)
bias_g.probe <- runif(100, 0.3, 1)
met <- rbind(random.probe,bias_1.probe,bias_g.probe)</pre>
met <- preAssociationProbeFiltering(data = met, K = 0.3, percentage = 0.05)</pre>
met <- rbind(random.probe,random.probe,random.probe)</pre>
met <- preAssociationProbeFiltering(met, K = 0.3, percentage = 0.05)</pre>
data <- ELMER:::getdata("elmer.data.example") # Get data from ELMER.data</pre>
data <- preAssociationProbeFiltering(data, K = 0.3, percentage = 0.05)</pre>
cg24741609 <- runif(100, 0, 1)
cg17468663 <- runif(100, 0, 0.3)
cg14036402 <- runif(100, 0.3, 1)
met <- rbind(cg24741609,cg14036402,cg17468663)
colnames(met) <- paste("sample",1:100)</pre>
exp <- met
rownames(exp) <- c("ENSG00000141510","ENSG00000171862","ENSG00000171863")
sample.info <- S4Vectors::DataFrame(primary = paste("sample",1:100),</pre>
                                      sample.type = rep(c("Normal", "Tumor"),50))
rownames(sample.info) <- colnames(exp)</pre>
mae <- createMAE(exp = exp, met = met, colData = sample.info, genome = "hg38")</pre>
mae <- preAssociationProbeFiltering(mae, K = 0.3, percentage = 0.05)</pre>
```

promoterMeth

promoterMeth to calculate associations of gene expression with DNA methylation at promoter regions

## **Description**

promoterMeth is a function to calculate associations of gene expression with DNA methylation at promoter regions.

## Usage

#### **Arguments**

data A Multi Assay Experiment object with DNA methylation and gene expression

Summarized Experiment objects

sig.pvalue A number specifies significant cutoff for gene silenced by promoter methylation.

Default is 0.01. P value is raw P value without adjustment.

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minSubgroupFrac

A number ranging from 0 to 1 specifying the percentage of samples used to create the groups U (unmethylated) and M (methylated) used to link probes to genes. Default is 0.4 (lowest quintile of all samples will be in the U group and

the highest quintile of all samples in the M group).

upstream Number of bp upstream of TSS to consider as promoter region downstream Number of bp downstream of TSS to consider as promoter region

save A logic. If it is true, the result will be saved.

cores Number of cores to be used in paralellization. Default 1 (no paralellization)

#### **Details**

promoterMeth

#### Value

A data frame contains genes whose expression significantly anti-correlated with promoter methylation.

## **Examples**

```
## Not run:
   data(elmer.data.example.promoter)
   Gene.promoter <- promoterMeth(mae.promoter)
## End(Not run)</pre>
```

render\_report

Build report for TCGA.pipe function

## **Description**

Build HTML report

```
render_report(
   title = "Report",
   mae.file,
   group.col,
   group1,
   group2,
   direction,
   dir.out,
   genome = "hg38",
   mode = "supervised",
   minSubgroupFrac = 0.2,
   minMetdiff = 0.3,
   metfdr = 0.01,
   permu = 10000,
   rawpval = 0.01,
```

52 render\_report

```
pe = 0.01,
nprobes = 10,
lower.OR = 1.1,
out_file = file.path(getwd(), "report.html"),
funcivar = FALSE
)
```

#### **Arguments**

title HTML report title mae.file Absolute path to the mae used in the analysis (.rda or .rds) Group col group.col Group 1 group1 group2 Group 2 direction used in the analysis direction dir.out Absolute path to folder with results. dir.out used in the analysis Genome of reference used in the analysis genome mode mode used in the analysis minSubgroupFrac minSubgroupFrac used in the analysis minMetdiff minMetdiff used in the analysis metfdr used in the analysis metfdr permu used in the analysis permu rawpval used in the analysis rawpval pe pe used in the analysis nprobes nprobes used in the analysis lower.OR lower.OR used in the analysis Output file name (i.e report.html) out\_file

## **Examples**

funcivar

```
## Not run:
render_report(
  group.col = "TN",
  group1 = "Tumor",
  group2 = "Normal",
  dir.out = "~/paper_elmer/Result/BRCA/TN_Tumor_vs_Normal/hypo/",
  direction = "hypo",
  mae.file = "~/paper_elmer/Result/BRCA/BRCA_mae_hg38.rda"
)
## End(Not run)
```

Include funcivar analysis?

scatter 53

scatter scatter

# Description

scatter

## Usage

```
scatter(
  meth,
  exp,
  legend.title = "Legend",
  category = NULL,
  xlab = NULL,
  ylab = NULL,
  ylim = NULL,
  dots.size = 0.9,
  title = NULL,
  correlation = FALSE,
  correlation.text.size = 3,
  color.value = NULL,
  lm_line = FALSE
)
```

## **Arguments**

meth A vector of number.

exp A vector of number or matrix with sample in column and gene in rows.

legend.title Plot legend title

category A vector of sample labels.

xlab A character specify the title of x axis. ylab A character specify the title of y axis.

ylim y-axis limit i.e. c(0,25)

dots.size Control dots size

title A character specify the figure title. correlation Show spearman correlation values

correlation.text.size

Correlation values

color.value A vector specify the color of each category, such as

lm\_line A logic. If it is TRUE, regression line will be added to the graph.

#### Value

A ggplot figure object

54 scatter.plot

scatter.plot	scatter.plot to plot scatter plots between gene expression and DNA methylation.
--------------	---

## Description

scatter.plot is a function to plot various scatter plots between gene expression and DNA methylation. When byPair is specified, scatter plot for individual probe-gene pairs will be generated. When byProbe is specified, scatter plots for one probes with nearby 20 gene pairs will be generated. When byTF is specified, scatter plot for TF expression and average DNA methylation at certain motif sites will be generated.

## Usage

```
scatter.plot(data,
             byPair = list(probe = c(), gene = c()),
             byProbe = list(probe = c(), numFlankingGenes = 20),
             byTF = list(TF = c(), probe = c()),
             category = NULL,
             ylim = NULL,
             dots.size = 0.9,
             correlation = FALSE,
             width = 7,
             height = 6,
             dir.out = "./",
             save = TRUE, ...)
```

## Arg

height

PDF height

Ę	guments		
	data	A multiAssay Experiment with DNA methylation and Gene Expression data. See create MAE function.	
	byPair	A list: byPair =list(probe=c(),gene=c()); probe contains a vector of probes' name and gene contains a vector of gene ID. The length of probe should be the same with length of gene. Output see numFlankingGenes	
	byProbe	A list byProbe =list(probe=c(), geneNum=20); probe contains a vector of probes'name and geneNum specify the number of gene near the probes will ploted. 20 is default for numFlankingGenes Output see detail.	
	byTF	A list byTF =list(TF=c(), probe=c()); TF contains a vector of TF's symbol and probe contains the a vector of probes' name. Output see detail.	
	category	A vector labels subtype of samples or a character which is the column name in the colData(data) in the multiAssayExperiment object. Once specified, samples will label different color. The color can be customized by using color.value.	
	ylim	y-axis limit i.e. $c(0.25)$	
	dots.size	Control dots size	
	correlation	Add pearson correlation values to the plot	
	width	PDF width	

schematic.plot 55

dir.out	A path specify the directory to which the figures will be saved. Current directory
	is default.
save	A logic. If true, figure will be saved to dir.out.
	color.value, lm_line in scatter function

#### **Details**

byPair The output will be scatter plot for individual pairs.

byProbe The output will be scatter plot for the probe and nearby genes.

by TF The output will be scatter plot for the TFs and the average DNA methylation at the probes set specified in by TF list.

#### Value

Scatter plots.

#### Author(s)

```
Lijing Yao (maintainer: lijingya@usc.edu)
```

#### **Examples**

schematic.plot

schematic.plot to plot schematic plots showing the locations of genes and probes.

## **Description**

schematic.plot is a function to plot schematic plots showing the locations of genes and probes.

56 schematic.plot

#### **Arguments**

data	A Multi Assay Experiment object with DNA methylation and gene expression Summarized Experiment objects	
group.col	A column defining the groups of the sample. You can view the available columns using: colnames(MultiAssayExperiment::colData(data)).	
group1	A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2.	
group2	A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2.#' @param byProbe A vector of probe names.	
pair	A data frame with three columns: Probe, Gene ID (Ensemble gene ID) and Pe (empirical p-value). This is the ouput of get.pair function.	
byProbe	A vector of probe names	
byGeneID	A vector of gene ID	
byCoordinate	A list contains chr, start and end. by Coordinate=list(chr=c(), start=c(), end=c()).	
statehub.tracks		
	Relative path to a statehub track.	
dir.out	A path specify the directory for outputs. Default is current directory	
save	A logic. If true, figures will be saved to dir.out.	
	Parameters for GetNearGenes	

#### **Details**

byProbes: When a vector of probes' name are provided, function will produce schematic plots for each individual probes. The schematic plot contains probe, nearby 20 (or the number of gene user specified.) genes and the significantly linked gene to the probe.

byGene: When a vector of gene ID are provided, function will produce schematic plots for each individual genes. The schematic plot contains the gene and all the significantly linked probes.

by Coordinate: When a genomic coordinate is provided, function will produce a schematic plot for this coordinate. The schematic plot contains all genes and significantly linked probes in the range and the significant links.

# Examples

Stat.diff.meth 57

```
pair = pair,
               byGeneID = "ENSG00000009790")
schematic.plot(data,
               group.col = "definition",
               group1 = "Primary solid Tumor",
               group2 = "Solid Tissue Normal",
               pair = pair,
               byCoordinate = list(chr="chr1", start = 209000000, end = 209960000))
## Not run:
   schematic.plot(data,
                  group.col = "definition",
                  group1 = "Primary solid Tumor",
                  group2 = "Solid Tissue Normal",
                  pair = pair,
                  byProbe = "cg19403323",
                  statehub.tracks = "hg38/ENCODE/mcf-7.16mark.segmentation.bed")
## End(Not run)
```

Stat.diff.meth

Stat.diff.meth

#### **Description**

Stat.diff.meth

## Usage

```
Stat.diff.meth(
  meth,
  groups,
  group1,
  group2,
  test = t.test,
  min.samples = 5,
  percentage = 0.2,
  Top.m = NULL
)
```

#### **Arguments**

meth A matrix contain DNA methylation data.

groups A vector of category of samples.
group1 Group 1 label in groups vector
group2 Group 2 label in groups vector

test A function specify which statistic test will be used.

min.samples Minimun number of samples to use in the analysis. Default 5. If you have 10

samples in one group, percentage is 0.2 this will give 2 samples in the lower

quintile, but then 5 will be used.

percentage A number specify the percentage of normal and tumor samples used in the test.

Top.m A logic. If to identify hypomethylated probe Top.m should be FALSE. hyper-

methylated probe is TRUE.

Stat.nonpara

## Value

Statistic test results to identify differentially methylated probes.

	Stat.nonpara	U test (non parameter test) for permutation. This is one probe vs nearby gene which is good for computing each probes for nearby genes.
--	--------------	---

# Description

U test (non parameter test) for permutation. This is one probe vs nearby gene which is good for computing each probes for nearby genes.

# Usage

```
Stat.nonpara(
   Probe,
   NearGenes,
   Top = NULL,
   correlation = "negative",
   unmethy = NULL,
   methy = NULL,
   Meths = Meths,
   Exps = Exps
)
```

# Arguments

Probe	A character of name of Probe in array.
NearGenes	A list of nearby gene for each probe which is output of GetNearGenes function.
Тор	A number determines the percentage of top methylated/unmethylated samples. Only used if unmethy and methy are not set.
correlation	Type of correlation to evaluate (negative or positive). Negative (default) checks if hypomethylated region has a upregulated target gene. Positive checks if region hypermethylated has a upregulated target gene.
unmethy	Index of U (unmethylated) group.
methy	Index of M (methylated) group.
Meths	A matrix contains methylation for each probe (row) and each sample (column).
Exps	A matrix contains Expression for each gene (row) and each sample (column).

## Value

U test results

Stat.nonpara.permu 59

Stat.nonpara.permu

Stat.nonpara.permu

#### **Description**

Stat.nonpara.permu

#### Usage

```
Stat.nonpara.permu(
   Probe,
   Gene,
   Top = 0.2,
   correlation = "negative",
   unmethy = NULL,
   methy = NULL,
   Meths = Meths,
   Exps = Exps
)
```

## **Arguments**

Probe A character of name of Probe in array.

Gene A vector of gene ID.

Top A number determines the percentage of top methylated/unmethylated samples.

Only used if unmethy and methy are not set.

correlation Type of correlation to evaluate (negative or positive). Negative (default) checks

if hypomethylated region has a upregulated target gene. Positive checks if region

hypermethylated has a upregulated target gene.

unmethy Index of U (unmethylated) group. methy Index of M (methylated) group.

Meths A matrix contains methylation for each probe (row) and each sample (column).

Exps A matrix contains Expression for each gene (row) and each sample (column).

## Value

U test results

summarizeTF

Make MR TF binary table

#### **Description**

This function uses ELMER analysis results and summarizes the MR TF identified in each analysis

```
summarizeTF(files = NULL, path = NULL, classification = "family", top = FALSE)
```

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## **Arguments**

files Output of get.pairs function: dataframe or file path

path Directory path with the ELMER results. Files with the following pattern will be

 $selected\ TF.* with.motif.summary.csv.$ 

classification Consider subfamily or family classifications

top Get only the top potential (default) or all potentials

TCGA.pipe

ELMER analysis pipeline for TCGA data.

## **Description**

ELMER analysis pipeline for TCGA data. This pipeline combine every steps of **ELMER** analyses: get.feature.probe, get.diff.meth, get.pair, get.permu, get.enriched.motif and get.TFs. Every steps' results are saved.

## Usage

```
TCGA.pipe(
 disease,
  genome = "hg38",
  analysis = "all",
 wd = getwd(),
 cores = 1,
 mode = "unsupervised",
 Data = NULL,
 diff.dir = "hypo",
 genes = NULL,
 mutant_variant_classification = c("Frame_Shift_Del", "Frame_Shift_Ins",
    "Missense_Mutation", "Nonsense_Mutation", "Splice_Site", "In_Frame_Del",
    "In_Frame_Ins", "Translation_Start_Site", "Nonstop_Mutation"),
 group.col = "TN",
 group1 = "Tumor",
 group2 = "Normal",
)
```

## **Arguments**

disease	TCGA short form disease name such as COAD
genome	Data aligned against which genome of reference. Options: "hg19", "hg38" (default)
analysis	A vector of characters listing the analysis need to be done. Analysis can be "download", "distal.probes", "diffMeth", "pair", "motif", "TF.search". Default is "all" meaning all the analysis will be processed.
wd	A path shows working directory. Default is "./"
cores	A interger which defines number of core to be used in parallel process. Default

is 1: don't use parallel process.

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mode	This option will automatically set the percentage of samples to be used in the analysis. Options: "supervised" (use 100% of samples) or "unsupervised" (use 20% of samples).
Data	A path shows the folder containing DNA methylation, expression and clinic data
diff.dir	A character can be "hypo" or "hyper", showing direction DNA methylation changes. If it is "hypo", get.diff.meth function will identify all significantly hypomethylated CpG sites; If "hyper", get.diff.meth function will identify all significantly hypermethylated CpG sites
genes	List of genes for which mutations will be verified. A column in the MAE with the name of the gene will be created with two groups WT (tumor samples without mutation), MUT (tumor samples w/ mutation), NA (not tumor samples)
mutant_variant	_classification
	List of TCGA variant classification from MAF files to consider a samples mutant. Only used when argument gene is set.
group.col	A column defining the groups of the sample. You can view the available columns using: colnames(MultiAssayExperiment::colData(data)).
group1	A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2.
group2	A group from group.col. ELMER will run group1 vs group2. That means, if direction is hyper, get probes hypermethylated in group 1 compared to group 2.
• • •	A list of parameters for functions: GetNearGenes, get.feature.probe, get.diff.meth, get.pair

## Value

Different analysis results.

# **Examples**

```
data <- ELMER:::getdata("elmer.data.example")</pre>
  TCGA.pipe(
    disease = "LUSC",
    data = data,
    analysis = c("diffMeth","pair", "motif","TF.search"),
    mode = "supervised",
    group.col = "definition",
    group1 = "Primary solid Tumor",
    group2 = "Solid Tissue Normal",
    diff.dir = c("hypo"),
    dir.out = "pipe",
    sig.dif = 0.0001,
    pvalue = 1.0,
    min.incidence = 0,
    lower.OR = 0.0
 )
## Not run:
 \label{linear_continuity} \verb|distal.probe| <- TCGA.pipe(disease = "LUSC", analysis="distal.enhancer", wd="~/")| \\
 TCGA.pipe(disease = "LUSC", analysis = "all", genome = "hg19", cores = 1, permu.size=300, Pe=0.01)
 projects <- TCGAbiolinks:::getGDCprojects()$project_id</pre>
  projects <- gsub("TCGA-","",projects[grep1('^TCGA',projects,per1=TRUE)])</pre>
  for(proj in projects) TCGA.pipe(disease = proj,analysis = "download")
  plyr::alply(sort(projects),1,function(proj) {
       tryCatch({
```

TF.rank.plot

```
print(proj);
         TCGA.pipe(disease = proj,analysis = c("createMAE"))})
       }, .progress = "text")
  plyr::alply(sort(projects),1,function(proj) {
   tryCatch({
      print(proj);
      TCGA.pipe(disease = proj,
                 analysis = c("diffMeth", "pair", "motif", "TF.search"))})
  }, .progress = "text")
  # Evaluation mutation
  TCGA.pipe(disease = "LUSC", analysis = "createMAE", gene = "NFE2L2")
  TCGA.pipe(
    disease = "LUSC",analysis = c("diffMeth","pair", "motif","TF.search"),
    mode = "supervised",
    group.col = "NFE2L2", group1 = "Mutant", group2 = "WT",
    diff.dir = c("hypo"),
    dir.out = "LUSC_NFE2L2_MutvsWT"
## End(Not run)
```

TF.rank.plot

TF.rank.plot to plot the scores (-log10(P value)) which assess the correlation between TF expression and average DNA methylation at motif sites.

## **Description**

TF.rank.plot is a function to plot the scores (-log10(P value)) which assess the correlation between TF expression and average DNA methylation at motif sites. The the motif relevant TF and top3 TFs will be labeled in a different color.

## Usage

```
TF.rank.plot(
  motif.pvalue,
  motif,
  title = NULL,
  TF.label = NULL,
  dir.out = "./",
  save = TRUE,
  cores = 1
)
```

## **Arguments**

motif.pvalue A matrix or a path specifying location of "XXX.with.motif.pvalue.rda" which is output of getTF.

motif A vector of characters specify the motif to plot
title Tite title (the motif will still be added to the title)

TF.rank.plot 63

TF.label	A list shows the label for each motif. If TF.label is not specified, the motif relevant TF and top3 TF will be labeled.	
dir.out	A path specify the directory to which the figures will be saved. Current directory is default.	
save	A logic. If true (default), figure will be saved to dir.out	
cores	A interger which defines the number of cores to be used in parallel process. Default is 1: no parallel process.	

#### Value

A plot shows the score (-log(P value)) of association between TF expression and DNA methylation at sites of a certain motif.

#### Author(s)

Lijing Yao (maintainer: lijingya@usc.edu)

#### **Examples**

```
library(ELMER)
data <- tryCatch(ELMER:::getdata("elmer.data.example"), error = function(e) {</pre>
  message(e)
  data(elmer.data.example, envir = environment())
"cg10097755", "cg09247779", "cg09181054"))
TF <- get.TFs(data,
             enriched.motif,
             group.col = "definition",
             group1 = "Primary solid Tumor",
             group2 = "Solid Tissue Normal",
             TFs = data.frame(
                    external_gene_name=c("TP53","TP63","TP73"),
                    ensembl_gene_id= c("ENSG00000141510",
                                      "ENSG00000073282",
                                      "ENSG00000078900"),
                    stringsAsFactors = FALSE),
            label="hypo")
TF.meth.cor <- get(load("getTF.hypo.TFs.with.motif.pvalue.rda"))</pre>
TF.rank.plot(motif.pvalue=TF.meth.cor,
           motif="P53_HUMAN.H11MO.0.A"
           TF.label=createMotifRelevantTfs("subfamily")["P53_HUMAN.H11MO.0.A"],
           save=TRUE)
TF.rank.plot(motif.pvalue=TF.meth.cor,
           motif="P53_HUMAN.H11MO.0.A",
           save=TRUE)
# Same as above
TF.rank.plot(motif.pvalue=TF.meth.cor,
           motif="P53_HUMAN.H11MO.0.A",
           dir.out = "TFplots",
           TF.label=createMotifRelevantTfs("family")["P53_HUMAN.H11MO.0.A"],
           save=TRUE)
```

64 TFsurvival.plot

TFsurvival.plot	Creates survival plot of based on the expression of a TF

# **Description**

This function will create a survival plot for the samples with higher, midium, low expression of a given transcription factor. By defau;t samples with higher expression are the top 30

## Usage

```
TFsurvival.plot(data, TF, xlim = NULL, percentage = 0.3, save = TRUE)
```

## **Arguments**

data A multi assay Experiment with clinical data in the phenotypic data matrix con-

taining the following columns: vital\_status, days\_to\_last\_follow\_up and days\_to\_death.

Default from GDC and TCGAbiolinks

TF A gene symbol

xlim Limit x axis showed in plot

percentage A number ranges from 0 to 1 specifying the percentage of samples in the higher

and lower expression groups. Default is 0.3

save Save plot as PDF

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