# Package 'EnMCB'

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```
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     Blocks using Ensemble Models
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Illumina Human Methylation 450 kanno

# Description

anno\_matrix

IlluminaHumanMethylation450kanno

# Usage

data(anno\_matrix)

# **Format**

IlluminaHumanMethylation450kanno.ilmn12.hg19 annotation file. This data have several columns

as.data.frame.ridgemat 3

```
as.data.frame.ridgemat
```

data frame ridge matrix

# **Description**

data frame ridge matrix

# Usage

```
## S3 method for class 'ridgemat'
as.data.frame(x, ...)
```

# **Arguments**

x data vector

... other parameters pass to as.data.frame.model.matrix()

as.ridgemat

ridge matrix

# Description

as.matrix attempts to turn its argument

# Usage

```
as.ridgemat(x)
```

# **Arguments**

Х

data vector

CompareMCB

Compare multiple methylation correlated blocks lists

# Description

This function is used to find the Methylation correlated blocks that differentially expressed between groups. This function calculates attractors of all the MCBs among the groups and find the attractor MCBs.

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#### Usage

```
CompareMCB(
   MCBs,
   method = c("attractors")[1],
   p_value = 0.05,
   min_CpGs = 5,
   platform = "Illumina Methylation 450K"
)
```

# Arguments

MCBs Methylation correlated blocks list.

method method used for calculation of differential expression,

should be one of "attractors", "t-test". Defualt is "attractors".

p\_value p value threshold for the test.

min\_CpGs threshold for minimum CpGs must included in the individual MCBs.

platform This parameter indicates the platform used to produce the methlyation profile.

#### **Details**

Currently, only illumina 450k platform is supported, the methylation profile need to convert into matrix format.

### Value

Object of class list with elements:

MCBsites Character set contains all CpG sites in MCBs.
MCBinformation Matrix contains the information of results.

### Author(s)

Xin Yu

# References

Xin Yu, De-Xin Kong, EnMCB: an R/bioconductor package for predicting disease progression based on methylation correlated blocks using ensemble models, Bioinformatics, 2021, btab415

# **Examples**

```
data('demo_data',package = "EnMCB")
```

create\_demo 5

create\_demo

create demo matrix

# Description

Demo matrix for methylation matrix.

# Usage

```
create_demo(model = c("all", "short")[1])
```

# **Arguments**

model

Two options, 'all' or 'short' for creating full dataset or very brief demo.

#### Value

This function will generate a demo data.

# Author(s)

Xin Yu

# **Examples**

```
demo_set<-create_demo()</pre>
```

demo\_data

Expression matrix of demo dataset.

# **Description**

A Expression matrix containing the 10020 CpGs beta value of 455 samples in TCGA lung Adeno-carcinoma dataset. This will call from create\_demo() function.

### Usage

```
data(demo_data)
```

# **Format**

ExpressionSet:

**rownames** rownames of 10020 CpG features **colnames** colnames of 455 samples

realdata Real data matrix for demo.

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demo\_MCBinformation MCB information.

#### **Description**

A dataset containing the number and other attributes of 94 MCBs; This results was created by the identification function IdentifyMCB. This data used for metricMCB function.

#### Usage

```
data(demo_MCBinformation)
```

#### **Format**

A data frame with 94 rows and 8 variables:

MCB\_no MCB code

start Start point of this MCB in the chromosome.

**end** End point of this MCB in the chromosome.

**CpGs** All the CpGs probe names in the MCB.

location Start, end point and the chromosome number of this MCB.

**chromosomes** the chromosome number of this MCB.

length the length of bps of this MCB in the chromosome.

**CpGs\_num** number of CpG probes of this MCB.

demo\_survival\_data Survival data of demo dataset.

# **Description**

A Surv containing survival value of 455 samples in TCGA lung Adenocarcinoma dataset.

# Usage

```
data(demo_survival_data)
```

#### **Format**

Surv data created by Surv() function in survival package. This data have two unnamed arguments, they will match time and event.

DiffMCB 7

DiffMCB Differential expressed methylation correlated blocks	
--	--

### **Description**

This function is used to find the Methylation correlated blocks that differentially expressed between groups based on the attractor framework. This function calculates attractors of all the MCBs among the groups and find the attractor MCBs.

# Usage

```
DiffMCB(
  methylation_matrix,
  class_vector,
  mcb_matrix = NULL,
  min.cpgsize = 5,
  pVals_num = 0.05,
  base_method = c("Fstat", "Tstat", "eBayes")[1],
  sec_method = c("ttest", "kstest")[1],
  ...
)
```

### **Arguments**

```
methylation_matrix
                   methylation profile matrix.
class_vector
                   class vectors that indicated the groups.
mcb_matrix
                   dataframe or matrix results returned by IdentifyMCB function.
                   threshold for minimum CpGs must included in the individual MCBs.
min.cpgsize
                   p value threshold for the test.
pVals_num
base_method
                   base method used for calculation of differentially methylated regions, should be
                   one of 'Fstat', 'Tstat', 'eBayes'. Defualt is Fstat.
                   secondly method in attractor framework, should be one of 'kstest', 'ttest'. De-
sec_method
                   fualt is ttest.
                   other parameters pass to the function.
. . .
```

# Details

Currently, only illumina 450k platform is supported.

If you want to use other platform, please provide the annotation file with CpG's chromosome and loci.

The methylation profile need to convert into matrix format.

### Value

Object of class list with elements:

global Character set contains statistical value for all CpG sites in MCBs.

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tab Matrix contains the information of results.

# Author(s)

Xin Yu

#### References

Xin Yu, De-Xin Kong, EnMCB: an R/bioconductor package for predicting disease progression based on methylation correlated blocks using ensemble models, Bioinformatics, 2021, btab415

#### **Examples**

draw\_survival\_curve

draw survival curve

### **Description**

Draw a survival curve based on survminer package. This is a wrapper function of ggsurvplot.

# Usage

```
draw_survival_curve(
  exp,
  living_days,
  living_events,
  write_name,
  title_name = "",
  threshold = NA,
  file = FALSE
)
```

#### Arguments

exp expression level for variable.

living\_days The survival time (days) for each individual.

living\_events The survival event for each individual, 0 indicates alive and 1 indicates death.

Other choices are TRUE/FALSE (TRUE = death) or 1/2 (2=death). For interval censored data, the status indicator is 0=right censored, 1=event at time, 2=left

censored, 3=interval censored.

write\_name The name for pdf file which contains the result figure.

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title\_name The title for the result figure.

threshold Threshold used to indicate the high risk or low risk.

file If True, function will automatic generate a result pdf, otherwise it will return a

ggplot object. Default is FALSE.

#### Value

This function will generate a pdf file with 300dpi which compare survival curves using the Kaplan-Meier (KM) test.

#### Author(s)

Xin Yu

# **Examples**

```
data(demo_survival_data)
library(survival)
demo_set<-create_demo()
draw_survival_curve(demo_set[1,],
    living_days = demo_survival_data[,1],
    living_events =demo_survival_data[,2],
    write_name = "demo_data" )</pre>
```

ensemble\_model

Trainging stacking ensemble model for Methylation Correlation Block

# **Description**

Method for training a stacking ensemble model for Methylation Correlation Block.

# Usage

```
ensemble_model(single_res,training_set,Surv_training,testing_set,
Surv_testing,ensemble_type)
```

# Arguments

single_res	Methylation Correlation Block information returned by the IndentifyMCB function.
training_set	methylation matrix used for training the model in the analysis.
Surv_training	Survival function contain the survival information for training.
testing_set	methylation matrix used for testing the model in the analysis.
Surv_testing	Survival function contain the survival information for testing.
ensemble_type	Secondary model use for ensemble, one of "Cox", "C-index" and "feature weighted linear regression". "feature weighted linear regression" only uses two meta-features namely kurtosis and S.D.

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#### Value

Object of class list with elements (XXX repesents the model you choose):

cox Model object for the cox model at first level.

svm Model object for the svm model at first level.

enet Model object for the enet model at first level.

mboost Model object for the mboost model at first level.

stacking Model object for the stacking model.

#### Author(s)

Xin Yu

#### References

Xin Yu et al. 2019 Predicting disease progression in lung adenocarcinoma patients based on methylation correlated blocks using ensemble machine learning classifiers (under review)

### **Examples**

```
#import datasets
library(survival)
data(demo_survival_data)
datamatrix<-create_demo()
data(demo_MCBinformation)
#select MCB with at least 3 CpGs.
demo_MCBinformation<-demo_MCBinformation[demo_MCBinformation[,"CpGs_num"]>2,]
trainingset<-colnames(datamatrix) %in% sample(colnames(datamatrix),0.6*length(colnames(datamatrix)))
select_single_one=1
em<-ensemble_model(t(demo_MCBinformation[select_single_one,]),
    training_set=datamatrix[,trainingset],
    Surv_training=demo_survival_data[trainingset])</pre>
```

 $\begin{tabular}{ll} ensemble\_prediction & \it fitting function using stacking ensemble model for Methylation Correlation Block \\ \end{tabular}$ 

# **Description**

predict is a generic function for predictions from the results of stacking ensemble model fitting functions. The function invokes particular methods which is the ensemble model described in the reference.

```
ensemble_prediction(ensemble_model, prediction_data, multiple_results = FALSE)
```

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#### **Arguments**

#### Value

Object of numeric class double

#### References

Xin Yu et al. 2019 Predicting disease progression in lung adenocarcinoma patients based on methylation correlated blocks using ensemble machine learning classifiers (under review)

# **Examples**

```
library(survival)
#import datasets
data(demo_survival_data)
datamatrix<-create_demo()</pre>
data(demo_MCBinformation)
#select MCB with at least 3 CpGs.
{\tt demo\_MCBinformation[-demo\_MCBinformation[,"CpGs\_num"]>2,]}
trainingset<-colnames(datamatrix) %in% sample(colnames(datamatrix), 0.6*length(colnames(datamatrix)))</pre>
testingset<-!trainingset
#select one MCB
select_single_one=1
em<-ensemble_model(t(demo_MCBinformation[select_single_one,]),</pre>
    training_set=datamatrix[,trainingset],
    Surv_training=demo_survival_data[trainingset])
em_prediction_results<-ensemble_prediction(ensemble_model = em,</pre>
prediction_data = datamatrix[,testingset])
```

fast\_roc\_calculation Fast calculation of AUC for ROC using parallel strategy

# **Description**

This function is used to create time-dependent ROC curve from censored survival data using the Kaplan-Meier (KM) or Nearest Neighbor Estimation (NNE) method of Heagerty, Lumley and Pepe, 2000

```
fast_roc_calculation(test_matrix, y_surv, predict_time = 5, roc_method = "NNE")
```

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#### **Arguments**

y\_surv Survival information created by Surv function in survival package.

roc\_method Method for fitting joint distribution of (marker,t), either of KM or NNE, the

default method is NNE.

#### Value

This will retrun a numeric vector contains AUC results for each row in test\_matrix.

#### Author(s)

Xin Yu

### **Examples**

```
data(demo_survival_data)
data('demo_data',package = "EnMCB")
demo_set<-demo_data$realdata
res<-fast_roc_calculation(demo_set[1:2,],demo_survival_data)</pre>
```

IdentifyMCB

Identification of methylation correlated blocks

# Description

This function is used to partition the genome into blocks of tightly co-methylated CpG sites, Methylation correlated blocks. This function calculates Pearson correlation coefficients between the beta values of any two CpGs < CorrelationThreshold was used to identify boundaries between any two

adjacent markers indicating uncorrelated methylation. Markers not separated by a boundary were combined into MCB. Pearson correlation coefficients between two adjacent CpGs were calculated.

```
IdentifyMCB(
  MethylationProfile,
  method = c("pearson", "spearman", "kendall")[1],
  CorrelationThreshold = 0.8,
  PositionGap = 1000,
  platform = "Illumina Methylation 450K",
  verbose = T
)
```

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#### **Arguments**

 ${\it Methylation Profile}$ 

Methylation matrix is used in the analysis.

method used for calculation of correlation,

should be one of "pearson", "spearman", "kendall". Defualt is "pearson".

CorrelationThreshold

coef correlation threshold is used for define boundaries.

PositionGap CpG Gap between any two CpGs positioned CpG sites less than 1000 bp (de-

fault) will be calculated.

platform This parameter indicates the platform used to produce the methlyation profile.

You can use your own annotation file.

verbose True as default, which will print the block information for each chromosome.

# **Details**

Currently, only illumina 450k platform is supported, the methylation profile need to convert into matrix format.

#### Value

Object of class list with elements:

MCBsites Character set contains all CpG sites in MCBs.
MCBinformation Matrix contains the information of results.

### Author(s)

Xin Yu

### References

Xin Yu, De-Xin Kong, EnMCB: an R/bioconductor package for predicting disease progression based on methylation correlated blocks using ensemble models, Bioinformatics, 2021, btab415

### **Examples**

```
data('demo_data',package = "EnMCB")
#import the demo TCGA data with 10000+ CpGs site and 455 samples
#remove # to run
res<-IdentifyMCB(demo_data$realdata)
demo_MCBinformation<-res$MCBinformation</pre>
```

### **Description**

This function is used to partition the genome into blocks of tightly co-methylated CpG sites,

Methylation correlated blocks parallelly. This function calculates Pearson correlation coefficients between

the beta values of any two CpGs < CorrelationThreshold was used to identify boundaries between any two

adjacent markers indicating uncorrelated methylation. Markers not separated by a boundary were combined into MCB.

Pearson correlation coefficients between two adjacent CpGs were calculated.

# Usage

```
IdentifyMCB_parallel(
   MethylationProfile,
   method = c("pearson", "spearman", "kendall")[1],
   CorrelationThreshold = 0.8,
   PositionGap = 1000,
   platform = "Illumina Methylation 450K",
   verbose = T
)
```

#### **Arguments**

MethylationProfile

Methylation matrix is used in the analysis.

method used for calculation of correlation,

should be one of "pearson", "spearman", "kendall". Defualt is "pearson".

CorrelationThreshold

coef correlation threshold is used for define boundaries.

PositionGap CpG Gap between any two CpGs positioned CpG sites less than 1000 bp (de-

fault) will be calculated.

platform This parameter indicates the platform used to produce the methlyation profile.

You can use your own annotation file.

verbose True as default, which will print the block information for each chromosome.

#### **Details**

Currently, only illumina 450k platform is supported, the methylation profile need to convert into matrix format.

# Value

Object of class list with elements:

MCBsites Character set contains all CpG sites in MCBs.
MCBinformation Matrix contains the information of results.

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#### Author(s)

Xin Yu

### References

Xin Yu, De-Xin Kong, EnMCB: an R/bioconductor package for predicting disease progression based on methylation correlated blocks using ensemble models, Bioinformatics, 2021, btab415

### **Examples**

```
data('demo_data',package = "EnMCB")

#import the demo TCGA data with 10000+ CpGs site and 455 samples
#remove # to run
res<-IdentifyMCB_parallel(demo_data$realdata)
demo_MCBinformation<-res$MCBinformation</pre>
```

metricMCB

Calculation of the metric matrix for Methylation Correlation Block

# **Description**

To enable quantitative analysis of the methylation patterns

within individual Methylation Correlation Blocks across many samples, a single metric to define the methylated pattern of multiple CpG sites within each block.

Compound scores which calculated all CpGs within individual Methylation Correlation Blocks by linear, SVM or elastic-net model

Predict values were used as the compound methylation values of Methylation Correlation Blocks.

# Usage

```
metricMCB(MCBset,training_set,Surv,testing_set,
Surv.new,Method,predict_time,ci,silent,alpha,n_mstop,n_nu,theta)
```

#### **Arguments**

McBset Methylation Correlation Block information returned by the IndentifyMCB func-

tion.

training\_set methylation matrix used for training the model in the analysis.

Surv

Survival function contain the survival information for training.

testing\_set methylation matrix used in the analysis. This can be missing then training set

itself will be used as testing set.

Surv. new Survival function contain the survival information for testing.

Method model used to calculate the compound values for multiple Methylation correla-

tion blocks.

Options include "svm" "cox" "mboost" and "enet". The default option is SVM

method.

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<pre>predict_time</pre>	time point of the ROC curve used in the AUC calculations, default is 5 years.
ci	if True, the confidence intervals for AUC under area under the receiver operating characteristic curve will be calculated. This will be time consuming. default is False.
silent	True indicates that processing information and progress bar will be shown.
alpha	The elasticnet mixing parameter, with $0 \le \text{alpha} \le 1$ . alpha=1 is the lasso penalty, and alpha=0 the ridge penalty. It works only when "enet" Method is selected.
n_mstop	an integer giving the number of initial boosting iterations. If mstop = 0, the offset model is returned.  It works only when "mboost" Method is selected.
n_nu	a double (between 0 and 1) defining the step size or shrinkage parameter in mboost model.  It works only when "mboost" Method is selected.
theta	penalty used in the penalized coxph model, which is theta/2 time sum of squared coefficients. default is 1.

#### Value

Object of class list with elements (XXX will be replaced with the model name you choose):

It works only when "cox" Method is selected.

MCB\_XXX\_matrix\_training
MCB\_XXX\_matrix\_training
MCB\_XXX\_matrix\_test\_set

XXX\_auc\_results
Dest\_XXX\_model
Model object for the model with best AUC.

Maximum\_auc

Model object for the whole generated models.

## Author(s)

Xin Yu

# References

Xin Yu et al. 2019 Predicting disease progression in lung adenocarcinoma patients based on methylation correlated blocks using ensemble machine learning classifiers (under review)

### **Examples**

metricMCB.cv 17

```
testing_set = datamatrix[,testingset],
Surv.new = demo_survival_data[testingset],
Method = "cox"
)
```

metricMCB.cv

Calculation of model AUC for Methylation Correlation Blocks using cross validation

# Description

To enable quantitative analysis of the methylation patterns within individual Methylation Correlation Blocks across many samples, a single metric to define the methylated pattern of multiple CpG sites within each block. Compound scores which calculated all CpGs within individual Methylation Correlation Blocks by SVM model were used as the compound methylation values of Methylation Correlation Blocks.

# Usage

```
metricMCB.cv(MCBset,data_set,Surv,nfold,
Method,predict_time,alpha,n_mstop,n_nu,theta,silent)
```

# **Arguments**

MCBset	Methylation Correlation Block information returned by the IndentifyMCB function.
data_set	methylation matrix used for training the model in the analysis.
Surv	Survival function contain the survival information for training.
nfold	fold used in the cross validation precedure.
Method	model used to calculate the compound values for multiple Methylation correlation blocks. Options include "svm", "cox", "mboost", and "enet". The default option is SVM method.
<pre>predict_time</pre>	time point of the ROC curve used in the AUC calculations, default is 3 years.
alpha	The elasticnet mixing parameter, with $0 \le \text{alpha} \le 1$ . alpha=1 is the lasso penalty, and alpha=0 the ridge penalty. It works only when "enet" Method is selected.
n_mstop	an integer giving the number of initial boosting iterations. If $mstop = 0$ , the offset model is returned. It works only when "mboost" Method is selected.
n_nu	a double (between 0 and 1) defining the step size or shrinkage parameter in mboost model. It works only when "mboost" Method is selected.
theta	penalty used in the penalized coxph model, which is theta/2 time sum of squared coefficients. default is 1. It works only when "cox" Method is selected.
silent	Ture indicates that processing information and progress bar will be shown.

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#### Value

Object of class list with elements (XXX will be replaced with the model name you choose):

MCB\_matrix Prediction results of model. auc\_results AUC results for each model.

### Author(s)

Xin Yu

#### References

Xin Yu et al. 2019 Predicting disease progression in lung adenocarcinoma patients based on methylation correlated blocks using ensemble machine learning classifiers (under review)

# **Examples**

multi\_coxph

multivariate survival analysis using coxph

# Description

multivariate survival analysis using coxph

Default is true.

# Usage

```
multi_coxph(dataframe, y_surv, digits = 4, asnumeric = TRUE)
```

# Arguments

dataframe	Clinic data and covariates ready to be tested. Note that Rows are samples and columns are variables.
y_surv	Survival function contain survival data, usually are obtained form Surv() function in survival package.
digits	Integer indicating the number of decimal places.
asnumeric	indicator that the data will be (True) / not (False) transformed into numeric.

#### Value

Object of class matrix with results.

#### Author(s)

Xin Yu

# **Examples**

```
data(demo_survival_data)
data('demo_data',package = "EnMCB")
demo_set<-demo_data$realdata
res<-multi_coxph(t(demo_set),demo_survival_data)</pre>
```

```
predict.mcb.coxph.penal
```

predict coxph penal using MCB

# **Description**

Compute fitted values and regression terms for a model fitted by coxph

# Usage

```
## S3 method for class 'mcb.coxph.penal'
predict(object, newdata, ...)
```

# **Arguments**

object the results of a coxph fit.

newdata Optional new data at which to do predictions. If absent predictions are for the

data frame used in the original fit. When coxph has been called with a formula argument created in another context, i.e., coxph has been called within another function and the formula was passed as an argument to that function, there can

be problems finding the data set. See the note below.

... other parameters pass to predict.coxph

#### Value

prediction values of regression.

#### Author(s)

Xin Yu

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```
pre_process_methylation
```

Preprocess the Beta value matrix

# **Description**

This process is optional for the pipeline. This function pre-process the Beta matrix and transform the Beta value into M value.

# Usage

```
pre_process_methylation(met,Mvalue,constant_offset,remove_na,remove_percentage)
```

# **Arguments**

met methylation matrix for CpGs. Rows are the CpG names, columns are samples.

Mvalue Boolean value, TRUE for the M transformation.

constant\_offset

the constant offset used in the M transformation formula.

remove\_na Boolean value, if TRUE ,CpGs with NA values will be removed.

remove\_percentage

If precentage of NA value exceed the threshold(percentage), the whole CpG probe will be removed. Otherwise, the NA values are replaced with rowmeans.

#### Value

Object of class matrix.

# **Examples**

```
demo_set<-create_demo()
pre_process_methylation(demo_set,Mvalue=FALSE)</pre>
```

univ\_coxph

univariate and multivariate survival analysis using coxph

# Description

univariate and multivariate survival analysis using coxph

```
univ_coxph(dataframe, y_surv, digits = 4, asnumeric = TRUE)
```

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# **Arguments**

dataframe Clinic data and covariates ready to be tested. Rows are variables and columns

are samples.

y\_surv Survival function contain survival data, usually are obtained form Surv() func-

tion in survival package.

digits Integer indicating the number of decimal places.

asnumeric indicator that the data will be (True) / not (False) transformed into numeric.

Default is true.

# Value

Object of class matrix with results.

# Author(s)

Xin Yu

# **Examples**

```
data(demo_survival_data)
data('demo_data',package = "EnMCB")
demo_set<-demo_data$realdata
res<-univ_coxph(demo_set,demo_survival_data)</pre>
```

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