

Package ‘PEAm5c’

January 5, 2018

Type Package

Title An integrated R toolkit for plant m5C analysis.

Version 0.1.0

Author Jie Song, Jingjing Zhai amd Chuang Ma

Maintainer Jie Song <q2516581@gmail.com>

Depends R (>= 3.3.2), seqinr (>= 3.3-6), stringr (>= 1.2.0),
randomForest (>= 4.6-12),ggplot2 (>= 2.2.1)

Imports pROC (>= 1.5.4), ROCR (>= 1.0-5),FSelector (>= 0.21),e1071 (>= 1.6-8)

Description PEA5C is an integrated R toolkit designed to identify plants' m5C sites. The toolkit contains functions such as analysis sequence characteristics, m5C prediction from transcriptional group size, and custom build recognition model.

License GPL (>= 2)

Encoding UTF-8

LazyData true

URL <http://bioinfo.nwafu.edu.cn/software>

NeedsCompilation no

R topics documented:

cvgroup	2
extra_model	2
extra_motif_seq	3
FeatureExtract	4
PEA_ml	5
predict_m5c	6
predict_self_model	7
Index	8

cvgroup	<i>Cross validation grouping</i>
---------	----------------------------------

Description

Grouping data sets by cross validation

Usage

```
cvgroup(data, cvnum, seed=1234)
```

Arguments

data	A numeric vector, i can be the dataset number set (such as 1:1000)
cvnum	A numeric value, the number of Cross validation group (k-fold)

Author(s)

Jingjing Zhai, Chuang Ma, Jie Song.

Examples

```
cvgroup(1:100, 5)
```

extra_model	<i>extra model from PEA_ml</i>
-------------	--------------------------------

Description

Propose the desired model from the established result information

Usage

```
extra_model(res, ignum)
```

Arguments

res	a list , result of PEA_ml function
ignum	a numeric value, how many feature what you want

Value

models and selected feature

Examples

```
load(paste0(system.file(package = "PEAm5c"), "/data/samples.Rds"))
aaa <- PEA_ml(pos_sample = pos_sample, neg_sample = neg_sample)
ddd <- extra_model(res = aaa, ignum=150)
ddd
```

extra_motif_seq	<i>Scanning the motifs in the transcript.</i>
-----------------	---

Description

For a given motif, all transcript sequences will be scanned and then according to the size of "up" "down", the output fits the sequence of length.

Usage

```
extra_motif_seq(input_seq_dir, text='c', up=5, end=5)
```

Arguments

input_seq_dir	A path representing the file name of the sequence in FASTA format.
text	A string, which specifies the motif to be searched.
up	A numeric, The length of the upstream sequence required.
end	A numeric, The length of the downstream sequence required.

Value

A list of sequences around motif.

Author(s)

Jie Song, Jingjing Zhai, Chuang Ma

Examples

```
aaa <- extra_motif_seq(input_seq_dir = paste0(system.file(package = "PEAm5c"), "/data/cdna.fa"), up = 5)
aaa
```

FeatureExtract	<i>Feature encoding</i>
----------------	-------------------------

Description

This function contains three feature encoding scheme, binary, k-mer and PseDNC. For binary encoding scheme, a vector of 404 (4*101) features is generated through assigning 'A', 'C', 'G', 'U' and 'N' with (1,0,0,0), (0,1,0,0), (0,0,1,0), (0,0,0,1) and (0,0,0,0), respectively. Here 'N' is a gap used to ensure the fixed features of each sample, if an m6A/non- m6A site occurs near the initiation or termination of the transcript. For K-mer encoding, the composition of short sequence with different lengths was considered to encoding samples. For PseDNC (pseudo dinucleotide composition) encoding, the local and global sequence-order information along the RNA sequence was used for scoring the each sample.

Usage

```
FeatureExtract(RNaseq, lambda = 6, w = 0.9)
```

Arguments

RNaseq	A list containing the FASTA format sequences.
lambda	The lambda parameter for the PseDNC-related features, default is 6.
w	The weighting parameter for PseDNC-related features, default is 0.9.

Value

A matrix with features.

Author(s)

Jie Song, Jingjing Zhai, Chuang Ma

Examples

```
aaa <- extra_motif_seq(input_seq_dir = paste0(system.file(package = "PEAm5c"),"/data/cdna.fa"),up = 5)
aaa <- lapply(aaa, c2s)
bbb <- FeatureExtract(aaa)
bbb[1:10,]
```

PEA_ml	<i>Machine learning-based Randomforest algorithm and information gain rank for sequences</i>
--------	--

Description

First of all, the function separate the positive and negative sample to the independent test set automatically according to setting and training set, concentrated training shall be carried out in accordance with the cross validation in training and assessment, in the results returned to characteristics of the evaluation, prediction score and evaluation as well as the sample group. In this case, the function can define the proportion of positive and negative samples (when negative sample size can be selected to maintain multiple models), and provide two methods of random forest and support vector machine.

Usage

```
PEA_ml(pos_sample,neg_sample,independent_num=100,ig="ALL",
        times = 1,modeltype = "RFC",cvnum = 5,repeatTimes = 1, ntree=200,over_sampling = F)
```

Arguments

pos_sample	A numeric matrix recording the features for positive sample.
neg_sample	A numeric matrix recording the features for negative sample.
independent_num	A numeric value, how many independent samples what you want
ig	A numeric value,vector or "ALL", how many features what you want base the top of information gain rank
modeltype	A character string, machine learning method.
cvnum	An integer value, cross-times cross validation.
repeatTimes	An integer value,If the negative sample is larger than the limit of the positive sample, the number of the negative samples and the number of samples of the positive sample is repeated
over_sampling	F or T, where TRUE represents balance the positive and negative samples according to the ratio based smote simulation
times	A numeric value, where 1 represents balance the positive and negative sample.

Value

A list of result.

The first level is used feature num group.

The second level is cross validation group.

The third level is the detail information including

```

positives.test.score.id
negatives.test.score.id
positives.test.score
negatives.test.score
positives.test
negatives.test
auc_test
auc_test_id

```

Author(s)

Jie Song, Jingjing Zhai, Chuang Ma

Examples

```

load(paste0(system.file(package = "PEAm5c"), "/data/samples.Rds"))
aaa <- PEA_ml(pos_sample = pos_sample, neg_sample = neg_sample)
aaa

```

predict_m5c

Predict m5C sites by PEA-m5C

Description

The m5C methylation level of the detection sequences was evaluated by the proposed m5C random forest models.

Usage

```
predict_m5c(sample_feature)
```

Arguments

`sample_feature` A dataframe or list of Undetected sequence.

Value

A matrix with transcript , its sites, predicted score and m5C level.

Author(s)

Jie Song, Jingjing Zhai, Chuang Ma

Examples

```
aaa <- extra_motif_seq(input_seq_dir = paste0(system.file(package = "PEAm5c"),"/data/cdna.fa"),up = 5)
aaa <- lapply(aaa, c2s)
bbb <- FeatureExtract(aaa)
ccc <- predict_m5c(bbb)
```

predict_self_model	<i>Predict CMRs sites by self models</i>
--------------------	--

Description

The CMRs methylation level of the detection sequences was evaluated by the self models.

Usage

```
predict_self_model(models,sequence_dir,end = 5,up = 5)
```

Arguments

models	A dataframe self models and selected feature.
sequence_dir	A path representing the file name of the sequence in FASTA format.
...	

Examples

```
load(paste0(system.file(package = "PEAm5c"),"/data/samples.Rds"))
aaa <- PEA_ml(pos_sample = pos_sample,neg_sample = neg_sample)
ddd <- extra_model(res = aaa)
ddd
#
eee <- predict_self_model(models = ddd,sequence_dir = paste0(system.file(package = "PEAm5c"),"/data/cdna.fa"))
table(eee[,4])
```

Index

- *Topic **PEA-m5C**
 - predict_m5c, [6](#)
- *Topic **PseDNC**
 - FeatureExtract, [4](#)
- *Topic **\textasciitildekw1**
 - cvgroup, [2](#)
 - extra_model, [2](#)
 - PEA_ml, [5](#)
 - predict_self_model, [7](#)
- *Topic **\textasciitildekw2**
 - cvgroup, [2](#)
 - extra_model, [2](#)
 - PEA_ml, [5](#)
 - predict_self_model, [7](#)
- *Topic **extra motif seq**
 - extra_motif_seq, [3](#)
- *Topic **feature encoding**
 - FeatureExtract, [4](#)

cvgroup, [2](#)

extra_model, [2](#)

extra_motif_seq, [3](#)

FeatureExtract, [4](#)

PEA_ml, [5](#)

predict_m5c, [6](#)

predict_self_model, [7](#)