

# Package ‘cleanUpdTSeq’

April 15, 2019

**Type** Package

**Title** This package classifies putative polyadenylation sites as true or false/internally oligodT primed

**Version** 1.20.1

**Date** 2019-04-09

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**Depends** R (>= 2.15), BiocGenerics (>= 0.1.0), methods, BSgenome, BSgenome.Drerio.UCSC.danRer7, GenomicRanges, seqinr, e1071

**Suggests** BiocStyle, knitr, RUnit

**Description** This package uses the Naive Bayes classifier (from e1071) to assign probability values to putative polyadenylation sites (pA sites) based on training data from zebrafish. This will allow the user to separate true, biologically relevant pA sites from false, oligodT primed pA sites.

**License** GPL-2

**biocViews** Sequencing, SequenceMatching, Genetics, GeneRegulation

**VignetteBuilder** knitr

**git\_url** <https://git.bioconductor.org/packages/cleanUpdTSeq>

**git\_branch** RELEASE\_3\_8

**git\_last\_commit** aa0569e

**git\_last\_commit\_date** 2019-04-09

**Date/Publication** 2019-04-15

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cleanUpdTSeq-package *This package classifies putative polyadenylation sites.*

---

## Description

3'ends of transcripts have generally been poorly annotated. With the advent of deep sequencing, many methods have been developed to identify 3'ends. The majority of these methods use an oligodT primer which can bind to internal adenine-rich sequences, and lead to artifactual identification of polyadenylation sites. Heuristic filtering methods rely on a certain number of As downstream of a putative polyadenylation site to classify the site as true or oligodT primed. This package provides a robust method to classify putative polyadenylation sites using a Naive Bayes classifier.

## Details

Package: cleanUpdTSeq  
 Type: Package  
 Version: 1.0  
 Date: 2013-07-22  
 License: GPL-2

## Author(s)

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## References

1. Meyer, D., et al., e1071: Misc Functions of the Department of Statistics (e1071), TU Wien. 2012.
2. Pages, H., BSgenome: Infrastructure for Biostrings-based genome data packages.
3. Sarah Sheppard, Nathan D. Lawson, and Lihua Julie Zhu. 2013. Accurate identification of polyadenylation sites from 3' end deep sequencing using a naive Bayes classifier. Bioinformatics. Under revision

## Examples

```
#read in a test set
#### first install the package using the following command
#### BiocManager::install("cleanUpdTSeq")
```

```

if (interactive())
{
library(cleanUpdTSeq)
testFile = system.file("extdata", "test.bed", package="cleanUpdTSeq")
testSet = read.table(testFile, sep = "\t", header = TRUE)

#convert the test set to GRanges with upstream and downstream sequence information
peaks = BED2GRangesSeq(testSet,upstream.seq.ind = 7, downstream.seq.ind = 8, withSeq=TRUE)
#build the feature vector for the test set with sequence information
testSet.NaiveBayes = buildFeatureVector(peaks,BSgenomeName = Drerio, upstream = 40,
  downstream = 30, wordSize = 6, alphabet=c("ACGT"),
  sampleType = "unknown",replaceNAdistance = 30,
  method = "NaiveBayes", ZeroBasedIndex = 1, fetchSeq = FALSE)

#convert the test set to GRanges without upstream and downstream sequence information
  peaks = BED2GRangesSeq(testSet,withSeq=FALSE)

#build the feature vector for the test set without sequence information
testSet.NaiveBayes = buildFeatureVector(peaks,BSgenomeName = Drerio, upstream = 40,
  downstream = 30, wordSize = 6, alphabet=c("ACGT"),
  sampleType = "unknown",replaceNAdistance = 30,
  method = "NaiveBayes", ZeroBasedIndex = 1, fetchSeq = TRUE)

#predict the test set
data(data.NaiveBayes)
predictTestSet(data.NaiveBayes$Negative, data.NaiveBayes$Positive, testSet.NaiveBayes,
outputFile = "test-predNaiveBayes.tsv", assignmentCutoff = 0.5)
}

```

---

BED2GRangesSeq

*BED2GRangesSeq*


---

## Description

This function converts an object of data.frame from a bed file with sequence information to GRanges with sequence information.

## Usage

```

BED2GRangesSeq(data.BED, header = FALSE,
  upstream.seq.ind = 7, downstream.seq.ind = 8,
  withSeq)

```

## Arguments

`data.BED` An object of data.frame from a bed file. The data.frame should at least contains 3 required fields: `chrome`, `chromStart`, `chromend`. The fourth field for "name" is suggested for keeping track of the putative polyadenylation site from the input to the output. The sixth field for "strand" is suggested, as this will affect the classification. For this function, the bed data.frame may also contain two additional fields for the sequence upstream and downstream of the putative pA site. If these are not supplied, the sequence may be obtained when the feature vector is built. Please see <http://genome.ucsc.edu/FAQ/FAQformat.html#format1> for more information about the bed file format.

header	header = Boolean TRUE if the first row is the header FALSE if the first row is data
upstream.seq.ind	upstream.seq.ind = int, to delineate the column location containing the sequence upstream of the putative pA site
downstream.seq.ind	downstream.seq.ind = int, to delineate the column location containing the sequence downstream of the putative pA site
withSeq	TRUE indicates that there are upstream and downstream sequences in the file, FALSE indicates that there is no upstream or downstream sequences in the file

**Value**

Returns object of GRanges

**Author(s)**

Sarah Sheppard, Jianhong Ou, Nathan Lawson, Lihua Julie Zhu

**Examples**

```
testFile <- system.file("extdata", "test.bed", package="cleanUpdTSeq")
testSet <- read.table(testFile, sep = "\t", header = TRUE)
peaks <- BED2GRangesSeq(testSet,withSeq=TRUE)
```

---

buildClassifier	<i>get Naive Bayes Classifier</i>
-----------------	-----------------------------------

---

**Description**

Computes the conditional a-posterior probabilities of a categorical class variable given independent predictor variables using the Bayes rule.

**Usage**

```
buildClassifier(Ndata.NaiveBayes, Pdata.NaiveBayes,
               upstream=40L, downstream=30L, wordSize=6L,
               genome=Drerio, alphabet=c("ACGT"))
```

**Arguments**

Ndata.NaiveBayes	This is the negative training data, described further in <a href="#">data.NaiveBayes</a> .
Pdata.NaiveBayes	This is the positive training data, described further in <a href="#">data.NaiveBayes</a> .
upstream	This is the length of upstream sequence to use in the analysis.
downstream	This is the length of downstream sequence to use in the analysis.
wordSize	This is the size of the word to use as a feature for the upstream sequence. wordSize = 6 should always be used.
genome	Name of the genome to use to get sequence. To find out a list of available genomes, please type <code>available.genomes()</code> in R.
alphabet	These are the bases to use, for example DNA bases ACTG.

**Value**

An object of class "naiveBayes".

**Author(s)**

Jianhong Ou

**See Also**

[naiveBayes](#)

**Examples**

```
if (interactive()){
  data(data.NaiveBayes)
  classifier <- buildClassifier(data.NaiveBayes$Negative, data.NaiveBayes$Positive)
}
```

---

buildFeatureVector      *buildFeatureVector*

---

**Description**

This function creates a data frame. Fields include peak name, upstream sequence, downstream sequence, and features to be used in classifying the putative polyadenylation site.

**Usage**

```
buildFeatureVector(peaks, BSgenomeName = Drerio, upstream = 50L,
  downstream = 40L, wordSize = 6L, alphabet = c("ACGT"),
  sampleType = c("TP", "TN", "unknown"), replaceNAdistance = 40L,
  method = c("NaiveBayes", "SVM"), ZeroBasedIndex = 1L, fetchSeq = FALSE)
```

**Arguments**

peaks	An object of GRanges that may contain the upstream and downstream sequence information. This item is created by the function <a href="#">BED2GRangesSeq</a> .
BSgenomeName	Name of the genome to use to get sequence. To find out a list of available genomes, please type <code>available.genomes()</code> in R.
upstream	This is the length of upstream sequence to use in the analysis.
downstream	This is the length of downstream sequence to use in the analysis.
wordSize	This is the size of the word to use as a feature for the upstream sequence. wordSize = 6 should always be used.
alphabet	These are the bases to use, for example DNA bases ACTG.
sampleType	This is the type of sample. For example TP (true positive) or TN (true negative) for training data or unknown for test data.
replaceNAdistance	If there is no A in the downstream sequence, then use this for the average distance of As to the putative polyadenylation site.

method	This is which machine learning method to specify. For this release, method should always be set to NaiveBayes.
ZeroBasedIndex	If the coordinates are set using Zero Based indexing, set this = 1.
fetchSeq	Boolean, for getting upstream and downstream sequence at this step or not.

**Value**

An object of "[featureVector](#)"

**Author(s)**

Sarah Sheppard, Jianhong Ou, Nathan Lawson, Lihua Julie Zhu

**Examples**

```
testFile = system.file("extdata", "test.bed", package="cleanUpdTSeq")
testSet = read.table(testFile, sep = "\t", header = TRUE)

#convert the test set to GRanges with upstream and downstream sequence information
peaks = BED2GRangesSeq(testSet[1:10, ], upstream.seq.ind = 7, downstream.seq.ind = 8, withSeq=TRUE)
#build the feature vector for the test set with sequence information
testSet.NaiveBayes = buildFeatureVector(peaks,BSgenomeName = Drerio, upstream = 40,
  downstream = 30, wordSize = 6, alphabet=c("ACGT"),
  sampleType = "unknown",replaceNAdistance = 30,
  method = "NaiveBayes", ZeroBasedIndex = 1, fetchSeq = FALSE)

#convert the test set to GRanges without upstream and downstream sequence information
peaks = BED2GRangesSeq(testSet[1:10, ],withSeq=FALSE)

#build the feature vector for the test set without sequence information
testSet.NaiveBayes = buildFeatureVector(peaks,BSgenomeName = Drerio, upstream = 40,
  downstream = 30, wordSize = 6, alphabet=c("ACGT"),
  sampleType = "unknown",replaceNAdistance = 30,
  method = "NaiveBayes", ZeroBasedIndex = 1, fetchSeq = TRUE)
```

---

classifier

*An object of class "naiveBayes" generated from data.NaiveBayes*

---

**Description**

An object of class "naiveBayes" generated from data.NaiveBayes

**Usage**

```
data("classifier")
```

**Format**

An object of class "[PASclassifier](#)" including components:

**Details**

```
classifier is generated by data(data.NaiveBayes) classifier <- buildClassifier(data.NaiveBayes$Ndata.NaiveBayes,
data.NaiveBayes$Pdata.NaiveBayes)
```

**Examples**

```
data(classifier)
names(classifier)
```

---

data.NaiveBayes	<i>Training Data</i>
-----------------	----------------------

---

**Description**

This is the negative and positive training data.

**Usage**

```
data(data.NaiveBayes)
```

**Format**

A list with 2 data frame, "Negative" and "Positive". Negative has 9219 observations on the following 4120 variables. And Positive is a data frame with 22770 observations on the following 4120 variables. The format is:

Negative 'data.frame': 9219 obs. of 4120 variables:

Positive 'data.frame': 22770 obs. of 4120 variables:

Both of them have same structure.

y a numeric vector

n.A.Downstream a numeric vector

n.C.Downstream a numeric vector

n.T.Downstream a numeric vector

n.G.Downstream a numeric vector

avg.distanceA2PeakEnd a numeric vector

dimer: **such as AA, AC, AG, AT, CA, ... etc.** a numeric vector

heximer: **such as AAAAAA, ACGTAC, ... etc.** a factor with levels 0 1

upstream.seq a vector of sequence string

downstream.seq a vector of sequence string

**Examples**

```
data(data.NaiveBayes)
head(str(data.NaiveBayes$Negative))
head(str(data.NaiveBayes$Positive))
```

---

```
featureVector-class    Class "featureVector"
```

---

### Description

An object of class "featureVector" represents the output of `buildFeatureVector`

### Objects from the Class

Objects can be created by calls of the form `new("featureVector", data, info)`.

### Slots

**data** An object of data frame. Fields include peak name, upstream sequence, downstream sequence, and features to be used in classifying the putative polyadenylation site

**info** Object of class `modelInfo`

### Methods

`$, $<-` Get or set the slot of `featureVector`

---

```
getDownstreamSequence  getDownstreamSequence
```

---

### Description

This function gets the sequence upstream of a putative pA site (including the site)

### Usage

```
getDownstreamSequence(peaks, downstream = 20, genome)
```

### Arguments

<code>peaks</code>	GRanges containing putative pA sites
<code>downstream</code>	<code>downstream = int</code> . This is the length of the sequence to get.
<code>genome</code>	BSgenomeName

### Value

Returns an object of GRanges with downstream sequences.

### Author(s)

Sarah Sheppard, Jianhong Ou, Nathan Lawson, Lihua Julie Zhu

### Examples

```
testFile <- system.file("extdata", "test.bed", package="cleanUpdTSeq")
testSet <- read.table(testFile, sep="\t", header=TRUE)
peaks <- BED2GRangesSeq(testSet[1:10, ], withSeq=FALSE)
seq = getDownstreamSequence(peaks, downstream=30, genome=Drerio)
```



---

getUpstreamSequence      *Get upstream sequences of the putative pA site*

---

### Description

This function gets the sequence upstream of a putative pA site (including the site)

### Usage

```
getUpstreamSequence(peaks, upstream = 40, genome)
```

### Arguments

peaks	GRanges containing putative pA sites
upstream	upstream = int. This is the length of the sequence to get.
genome	BSgenomeName

### Value

Returns an object of GRanges with upstream sequences.

### Author(s)

Sarah Sheppard, Jianhong Ou, Nathan Lawson, Lihua Julie Zhu

### Examples

```
testFile <- system.file("extdata", "test.bed", package="cleanUpdTSeq")
testSet <- read.table(testFile, sep="\t", header=TRUE)
peaks <- BED2GRangesSeq(testSet[1:10, ], withSeq=FALSE)
seq = getUpstreamSequence(peaks, upstream=40, genome=Drrerio)
```

---

modelInfo-class      *Class "modelInfo"*

---

### Description

An object of class "modelInfo" represents the information of sequence to use in the analysis

### Objects from the Class

Objects can be created by calls of the form `new("modelInfo", upstream, downstream, wordSize, alphabe, genom`

### Slots

**genome** Name of the genome to use to get sequence. To find out a list of available genomes, please type `available.genomes()` in R.

**upstream** This is the length of upstream sequence to use in the analysis.

**downstream** This is the length of downstream sequence to use in the analysis.

**wordSize** This is the size of the word to use as a feature for the upstream sequence. `wordSize = 6` should always be used.

**alphabet** These are the bases to use, for example DNA bases ACTG.

**Methods**

\$, \$<- Get or set the slot of [modelInfo](#)

---

naiveBayes-class	<i>Class "naiveBayes"</i>
------------------	---------------------------

---

**Description**

An object of class "naiveBayes" represents the conditional a-posterior probabilities of a categorical class variable given independent predictor variables using the Bayes rule.

**Objects from the Class**

Objects can be created by calls of the form `new("naiveBayes", apriori, tables,`

**Slots**

`apriori` Class distribution for the dependent variable.

`tables` A list of tables, one for each predictor variable. For each categorical variable a table giving, for each attribute level, the conditional probabilities given the target class. For each numeric variable, a table giving, for each target class, mean and standard deviation of the (sub-)variable.

**Methods**

\$, \$<- Get or set the slot of [naiveBayes](#)

---

PASclassifier-class	<i>Class "PASclassifier"</i>
---------------------	------------------------------

---

**Description**

An object of class "PASclassifier" represents the output of [buildClassifier](#)

**Objects from the Class**

Objects can be created by calls of the form `new("PASclassifier", classifier, info).`

**Slots**

`classifier` Object of class "naiveBayes" The output of [naiveBayes](#). An object of class "naive-Bayes" including components:

`apriori` Class distribution for the dependent variable.

`tables` A list of tables, one for each predictor variable. For each categorical variable a table giving, for each attribute level, the conditional probabilities given the target class. For each numeric variable, a table giving, for each target class, mean and standard deviation of the (sub-)variable.

`info` Object of class [modelInfo](#)

**Methods**

**\$, \$<-** Get or set the slot of [PASClassifier](#)

**Examples**

```
data(classifier)
classifier$info$upstream
classifier$info$wordSize
classifier$info$alphabet
```

---

predictTestSet	<i>predictTestSet</i>
----------------	-----------------------

---

**Description**

This function can be used to predict the probabilities for a set of putative pA sites.

**Usage**

```
predictTestSet(Ndata.NaiveBayes, Pdata.NaiveBayes, testSet.NaiveBayes, classifier=NULL,
  outputFile = "test-predNaiveBayes.tsv", assignmentCutoff = 0.5)
```

**Arguments**

Ndata.NaiveBayes	This is the negative training data, described further in <a href="#">data.NaiveBayes</a> .
Pdata.NaiveBayes	This is the positive training data, described further in <a href="#">data.NaiveBayes</a> .
classifier	An object of class <a href="#">PASClassifier</a> .
testSet.NaiveBayes	This is the test data, a feature vector that has been built for Naive Bayes analysis using the function "buildFeatureVector".
outputFile	This is the name of the file the output will be written to.
assignmentCutoff	This is the cutoff used to assign whether a putative pA is true or false. This can be any floating point number between 0 and 1. For example, assignmentCutoff = 0.5 will assign an putative pA site with prob.1 > 0.5 to the True class (1), and any putative pA site with prob.1 <= 0.5 as False (0).

**Value**

The output is written to a tab separated file containing fields for peak name, the probability of the putative pA site being false (prob.0), the probability of the putative pA site being true (prob.1), the predicted class (0/False or 1/True) depending on the assignment cutoff, and the upstream and downstream sequence used in assessing the putative pA site.

PeakName	This is the name of the putative pA site (originally from the 4th field in the bed file).
prob False/oligodT internally primed	This is the probability that the putative pA site is false. Values range from 0-1, with 1 meaning the site is False/oligodT internally primed.

prob True            This is the probability that the putative pA site is true. Values range from 0-1, with 1 meaning the site is True.

pred.class           This is the predicted class of the putative pA site, based on the assignment cutoff. 0= False/oligoT internally primed, 1 = True

UpstreamSeq         This is the upstream sequence of the putative pA site used in the analysis.

DownstreamSeq      This is the downstream sequence of the putative pA site used in the analysis.

The function also return an invisible matrix including all info as decribed above.

**Author(s)**

Sarah Sheppard, Jianhong Ou, Nathan Lawson, Lihua Julie Zhu

**References**

Sarah Sheppard, Nathan D. Lawson, and Lihua Julie Zhu. 2013. Accurate identification of polyadenylation sites from 3' end deep sequencing using a naive Bayes classifier. *Bioinformatics*. Under revision

**Examples**

```
testFile = system.file("extdata", "test.bed", package="cleanUpdTSeq")
testSet = read.table(testFile, sep = "\t", header = TRUE)

#convert the test set to GRanges without upstream and downstream sequence information
peaks = BED2GRangesSeq(testSet,withSeq=FALSE)

#build the feature vector for the test set without sequence information
testSet.NaiveBayes = buildFeatureVector(peaks,BSgenomeName = Drerio, upstream = 40,
downstream = 30, wordSize = 6, alphabet=c("ACGT"),
sampleType = "unknown",replaceNAdistance = 30,
method = "NaiveBayes", ZeroBasedIndex = 1, fetchSeq = TRUE)

data(data.NaiveBayes)

## sample the test data for code testing, DO NOT do this for real data
## START SAMPLING
samp <- c(1:22, sample(23:4119, 50), 4119, 4120)
Ndata.NaiveBayes <- data.NaiveBayes$Negative[,samp]
Pdata.NaiveBayes <- data.NaiveBayes$Positive[,samp]
testSet.NaiveBayes@data <- testSet.NaiveBayes@data[, samp-1]
## END SAMPLING

predictTestSet(Ndata.NaiveBayes,
               Pdata.NaiveBayes,
               testSet.NaiveBayes,
               outputFile="test-predNaiveBayes.xls",
               assignmentCutoff = 0.5)
```

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