

# Package ‘fastAdaboost’

August 29, 2016

**Type** Package

**Title** a Fast Implementation of Adaboost

**Description** Implements Adaboost based on C++ backend code.

This is blazingly fast and especially useful for large, in memory data sets. The package uses decision trees as weak classifiers. Once the classifiers have been trained, they can be used to predict new data. Currently, we support only binary classification tasks. The package implements the Adaboost.M1 algorithm and the real Adaboost(SAMME.R) algorithm.

**Version** 1.0.0

**Date** 2016-02-23

**Author** Sourav Chatterjee [aut, cre]

**Maintainer** Sourav Chatterjee <souravc83@gmail.com>

**License** MIT + file LICENSE

**URL** <https://github.com/souravc83/fastAdaboost>

**BugReports** <https://github.com/souravc83/fastAdaboost/issues>

**Depends** R (>= 3.1.2)

**Imports** Rcpp, rpart

**Suggests** testthat, knitr, MASS

**LazyData** yes

**LinkingTo** Rcpp (>= 0.12.0)

**RoxygenNote** 5.0.1

**NeedsCompilation** yes

**Repository** CRAN

**Date/Publication** 2016-02-28 09:59:32

## R topics documented:

adaboost . . . . .	2
fastAdaboost . . . . .	3
get_tree . . . . .	4
predict.adaboost . . . . .	4
predict.real_adaboost . . . . .	5
print.adaboost . . . . .	6
print.real_adaboost . . . . .	7
real_adaboost . . . . .	8

<b>Index</b>	<b>10</b>
--------------	-----------

---

adaboost	<i>Adaboost.M1 algorithm</i>
----------	------------------------------

---

### Description

Implements Freund and Schapire's Adaboost.M1 algorithm

### Usage

```
adaboost(formula, data, nIter, ...)
```

### Arguments

formula	Formula for models
data	Input dataframe
nIter	no. of classifiers
...	other optional arguments, not implemented now

### Details

This implements the Adaboost.M1 algorithm for a binary classification task. The target variable must be a factor with exactly two levels. The final classifier is a linear combination of weak decision tree classifiers.

### Value

object of class adaboost

### References

Freund, Y. and Schapire, R.E. (1996):“Experiments with a new boosting algorithm” . *In Proceedings of the Thirteenth International Conference on Machine Learning*, pp. 148–156, Morgan Kaufmann.

**See Also**

[real\\_adaboost](#), [predict.adaboost](#)

**Examples**

```
fakedata <- data.frame( X=c(rnorm(100,0,1),rnorm(100,1,1)), Y=c(rep(0,100),rep(1,100) ) )
fakedata$Y <- factor(fakedata$Y)
test_adaboost <- adaboost(Y~X, data=fakedata,10)
```

---

fastAdaboost

*fastAdaboost: fast adaboost implementation for R*

---

**Description**

fastAdaboost provides a blazingly fast implementation of both discrete and real adaboost algorithms, based on a C++ backend. The goal of the package is to provide fast performance for large in-memory data sets.

**Author(s)**

Sourav Chatterjee

**References**

Freund, Y. and Schapire, R.E. (1996):“Experiments with a new boosting algorithm” . *In Proceedings of the Thirteenth International Conference on Machine Learning*, pp. 148–156, Morgan Kaufmann.

Zhu, Ji, et al. “Multi-class adaboost” *Ann Arbor* 1001.48109 (2006): 1612.

**Examples**

```
fakedata <- data.frame( X=c(rnorm(100,0,1),rnorm(100,1,1)), Y=c(rep(0,100),rep(1,100) ) )
fakedata$Y <- factor(fakedata$Y)
test_adaboost <- adaboost(Y~X, fakedata, 10)
pred <- predict( test_adaboost,newdata=fakedata)
print(pred$error)
```

---

get_tree	<i>Fetches a decision tree</i>
----------	--------------------------------

---

**Description**

returns a single weak decision tree classifier which is part of the strong classifier

**Usage**

```
get_tree(object, tree_num)
```

**Arguments**

object	object of class adaboost
tree_num	integer describing the tree to get

**Details**

returns an individual tree from the adaboost object This can provide the user with some clarity on the individual building blocks of the strong classifier

**Value**

object of class rpart

**See Also**

[adaboost](#)

**Examples**

```
fakedata <- data.frame( X=c(rnorm(100,0,1),rnorm(100,1,1)), Y=c(rep(0,100),rep(1,100) ) )
fakedata$Y <- factor(fakedata$Y)
test_adaboost <- adaboost(Y~X, fakedata, 10)
tree <- get_tree(test_adaboost,5)
```

---

predict.adaboost	<i>predict method for adaboost objects</i>
------------------	--

---

**Description**

predictions for model corresponding to adaboost.m1 algorithm

**Usage**

```
## S3 method for class 'adaboost'
predict(object, newdata, ...)
```

**Arguments**

object	an object of class adaboost
newdata	dataframe on which we are looking to predict
...	arguments passed to predict.default

**Details**

makes predictions for an adaboost object on a new dataset. The target variable is not required for the prediction to work. However, the user must ensure that the test data has the same columns which were used as inputs to fit the original model. The error component of the prediction object(as in `pred$error`) can be used to get the error of the test set if the test data is labeled.

**Value**

predicted object, which is a list with the following components

formula	the formula used.
votes	total weighted votes achieved by each class
class	the class predicted by the classifier
prob	a matrix with predicted probability of each class for each observation
error	The error on the test data if labeled, otherwise NA

**See Also**

[adaboost](#)

**Examples**

```
fakedata <- data.frame( X=c(rnorm(100,0,1),rnorm(100,1,1)), Y=c(rep(0,100),rep(1,100) ) )
fakedata$Y <- factor(fakedata$Y)
test_adaboost <- adaboost(Y~X, fakedata, 10)
pred <- predict( test_adaboost,newdata=fakedata)
print(pred$error)
print( table(pred$class,fakedata$Y) )
```

---

`predict.real_adaboost` *predict method for real\_adaboost objects*

---

**Description**

predictions for model corresponding to real\_adaboost algorithm

**Usage**

```
## S3 method for class 'real_adaboost'
predict(object, newdata, ...)
```

**Arguments**

object	an object of class real_adaboost
newdata	dataframe on which we are looking to predict
...	arguments passed to predict.default

**Details**

makes predictions for an adaboost object on a new dataset using the real\_adaboost algorithm. The target variable is not required for the prediction to work. However, the user must ensure that the test data has the same columns which were used as inputs to fit the original model. The error component of the prediction object(as in pred\$error) can be used to get the error of the test set if the test data is labeled.

**Value**

predicted object, which is a list with the following components

formula	the formula used.
votes	total weighted votes achieved by each class
class	the class predicted by the classifier
prob	a matrix with predicted probability of each class for each observation
error	The error on the test data if labeled, otherwise NA

**See Also**

[real\\_adaboost](#)

**Examples**

```
fakedata <- data.frame( X=c(rnorm(100,0,1),rnorm(100,1,1)), Y=c(rep(0,100),rep(1,100) ) )
fakedata$Y <- factor(fakedata$Y)
test_real_adaboost <- real_adaboost(Y~X, fakedata, 10)
pred <- predict(test_real_adaboost,newdata=fakedata)
print(pred$error)
print( table(pred$class,fakedata$Y) )
```

---

print.adaboost

*Print adaboost.m1 model summary*

---

**Description**

S3 method to print an adaboost object

### Usage

```
## S3 method for class 'adaboost'  
print(x, ...)
```

### Arguments

x                    object of class adaboost  
...                   arguments passed to print.default

### Details

Displays basic information on the model, such as function call, dependent variable, the number of trees, and weights assigned to each tree

### Value

None

### See Also

[print.real\\_adaboost](#)

### Examples

```
fakedata <- data.frame( X=c(rnorm(100,0,1),rnorm(100,1,1)), Y=c(rep(0,100),rep(1,100) ) )  
fakedata$Y <- factor(fakedata$Y)  
test_adaboost <- adaboost(Y~X, fakedata, 10)  
print(test_adaboost)
```

---

print.real\_adaboost    *Print real adaboost model summary*

---

### Description

S3 method to print a real\_adaboost object

### Usage

```
## S3 method for class 'real_adaboost'  
print(x, ...)
```

### Arguments

x                    object of class real\_adaboost  
...                   arguments passed to print.default

**Details**

Displays basic information on the model, such as function call, dependent variable and the number of trees

**Value**

None

**See Also**

[print.adaboost](#)

**Examples**

```
fakedata <- data.frame( X=c(rnorm(100,0,1),rnorm(100,1,1)), Y=c(rep(0,100),rep(1,100) ) )
fakedata$Y <- factor(fakedata$Y)
test_real_adaboost<- real_adaboost(Y~X, fakedata, 10)
print(test_real_adaboost)
```

---

real_adaboost	<i>Real Adaboost algorithm</i>
---------------	--------------------------------

---

**Description**

Implements Zhu et al's real adaboost or SAMME.R algorithm

**Usage**

```
real_adaboost(formula, data, nIter, ...)
```

**Arguments**

formula	Formula for models
data	Input dataframe
nIter	no. of classifiers
...	other optional arguments, not implemented now

**Details**

This implements the real adaboost algorithm for a binary classification task. The target variable must be a factor with exactly two levels. The final classifier is a linear combination of weak decision tree classifiers. Real adaboost uses the class probabilities of the weak classifiers to iteratively update example weights. It has been found to have lower generalization errors than adaboost.m1 for the same number of iterations.

**Value**

object of class real\_adaboost



## References

Zhu, Ji, et al. "Multi-class adaboost" *Ann Arbor* 1001.48109 (2006): 1612.

## See Also

[adaboost](#), [predict.real\\_adaboost](#)

## Examples

```
fakedata <- data.frame( X=c(rnorm(100,0,1),rnorm(100,1,1)), Y=c(rep(0,100),rep(1,100) ) )
fakedata$Y <- factor(fakedata$Y)
test_adaboost <- real_adaboost(Y~X, data=fakedata,10)
```

# Index

`adaboost`, [2](#), [4](#), [5](#), [9](#)

`fastAdaboost`, [3](#)

`fastAdaboost-package (fastAdaboost)`, [3](#)

`get_tree`, [4](#)

`predict.adaboost`, [3](#), [4](#)

`predict.real_adaboost`, [5](#), [9](#)

`print.adaboost`, [6](#), [8](#)

`print.real_adaboost`, [7](#), [7](#)

`real_adaboost`, [3](#), [6](#), [8](#)