

# Package ‘SVMMaj’

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**Title** SVMMaj algorithm

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**Description** Implements the SVM-Maj algorithm to train data with Support Vector Machine, this algorithm uses two efficient updates, one for linear kernel and one for the nonlinear kernel.

**Imports** graphics, stats, MASS, methods

**Depends** R (>= 2.9.0), kernlab

**License** GPL-2

**LazyData** Yes

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## R topics documented:

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AusCredit

*Australian Credit Approval Dataset*

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## Description

This file concerns credit card applications of 690 households.

## Usage

AusCredit  
AusCredit.tr  
AusCredit.te

## Format

This data set has been split into two components for the convenience of the model training.

data.frame-object  $X$  consists of with 6 numerical and 8 categorical attributes. The labels have been changed for the convenience of the statistical algorithms. For example, attribute 4 originally had 3 labels p,g,gg and these have been changed to labels 1,2,3.

Factor  $y$  indicates whether the application has been Accepted or Rejected

The training set AusCredit.tr contains a randomly selected set of 400 subjects, and AusCredit.te contains the remaining 290 subjects. AusCredit contains all 690 objects.

## Details

All attribute names and values have been changed to meaningless symbols to protect confidentiality of the data.

This dataset is interesting because there is a good mix of attributes – continuous, nominal with small numbers of values, and nominal with larger numbers of values. There are also a few missing values.

## Source

Chih-Chung Chang and Chih-Jen Lin, LIBSVM : a library for support vector machines, 2001.  
Software available at <http://www.csie.ntu.edu.tw/~cjlin/libsvm>.

## Examples

```
attach(AusCredit)
summary(X)
summary(y)
detach(AusCredit)
```

---

diabetes	<i>Pima Indians Diabetes Data Set</i>
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**Description**

From National Institute of Diabetes and Digestive and Kidney Diseases.

**Usage**

diabetes  
diabetes.tr  
diabetes.te

**Format**

X is a data frame of 768 female patients with 8 attributes.

no.pregnant	number of pregnancies.
glucose	plasma glucose concentration in an oral glucose tolerance test
blood.press	diastolic blood pressure (mm Hg)
triceps.thick	triceps skin fold thickness (mm)
insulin	2-Hour serum insulin (mu U/ml)
BMI	body mass index (weight in kg/(height in m) <sup>2</sup> )
pedigree	diabetes pedigree function
age	age in years

y contains the class labels: Yes or No, for diabetic according to WHO criteria.

The training set diabetes.tr contains a randomly selected set of 600 subjects, and diabetes.te contains the remaining 168 subjects. diabetes contains all 768 objects.

**Details**

Several constraints were placed on the selection of these instances from a larger database. In particular, all patients here are females at least 21 years old of Pima Indian heritage.

**Source**

Chih-Chung Chang and Chih-Jen Lin, LIBSVM : a library for support vector machines, 2001. Software available at <http://www.csie.ntu.edu.tw/~cjlin/libsvm>.

**References**

Smith, J.W., Everhart, J.E., Dickson, W.C., Knowler, W.C., & Johannes, R.S. (1988). Using the ADAP learning algorithm to forecast the onset of diabetes mellitus. In Proceedings of the *Symposium on Computer Applications and Medical Care* (pp. 261–265). IEEE Computer Society Press.

**Examples**

```
attach(diabetes)
summary(X)
summary(y)
```

hinge

*Hinge error function of SVM-Maj***Description**

This function creates a function to compute the hinge error, given its predicted value  $q$  and its class  $y$ , according to the loss term of the Support Vector machine loss function.

**Usage**

```
getHinge(hinge = 'quadratic', hingeK = 3, eps = 1e-8)
## S3 method for class 'hinge'
plot(x, y=1, z=NULL,...)
```

**Arguments**

hinge	Hinge error function to be used, possible values are 'absolute', 'quadratic' and 'huber'
hingeK	The parameter of the huber hinge (only if hinge = 'huber').
eps	Specifies the maximum steepness of the quadratic majorization function $m(q) = a*q^2 - 2*b*q + c$ , where $a \leq .25 * eps^{-1}$ .
x	The hinge object returned from getHinge.
y	Specifies the class (-1 or 1) to be plotted for the hinge error.
z	If specified, the majorization function with the supporting point $z$ will also be plotted.
...	Other arguments passed to plot method.

**Value**

The hinge error function with arguments  $q$  and  $y$  to compute the hinge error. The function returns a list with the parameters of the majorization function SVM-Maj ( $a$ ,  $b$  and  $c$ ) and the loss error of each object (loss).

**Author(s)**

Hok San Yip, Patrick J.F. Groenen, Georgi Nalbantov

**References**

P.J.F. Groenen, G. Nalbantov and J.C. Bioch (2008) *SVM-Maj: a majorization approach to linear support vector machines with different hinge errors*.

**See Also**[svmmaj](#)**Examples**

```

hingefunction <- getHinge()
## plot hinge function value and, if specified, the majorization function at z
plot(hingefunction,z=3)
## generate loss function value
loss <- hingefunction(q = -10:10 ,y = 1)$loss
loss

```

isb

*I-spline basis of each column of a given matrix***Description**

Create a I-spline basis for an array. \ will equally distribute the knots over the value range using quantiles.

**Usage**

```
isb(x, spline.knots = 0 , knots=NULL, spline.degree = 1)
```

**Arguments**

x	The predictor variable, which will be transformed into I-spline basis.
spline.knots	Number of inner knots to use. isb will equally distribute the knots over the value range using quantiles. spline.knots will only be used if knots is not given.
knots	An array consisting all knots (boundary knots as well as the interior knots) to be used to create the spline basis.
spline.degree	The polynomial degree of the spline basis.

**Value**

Returns the I-spline with the used spline settings as attribute. The spline settings attribute can transform the same attribute of any other objects using the same knots.

**Author(s)**

Hok San Yip, Patrick J.F. Groenen, Georgi Nalbantov

## References

P.J.F. Groenen, G. Nalbantov and J.C. Bioch (2008) *SVM-Maj: a majorization approach to linear support vector machines with different hinge errors*. J.O. Ramsay (1988) *Monotone regression splines in action*. Statistical Science, 3(4):425-461

## See Also

[svmmaj](#)

## Examples

```
## create I-spline basis for the first 50 observations
x <- iris$Sepal.Length
B1 <- isb(x[1:50],spline.knots=4,spline.degree=3)
## extracting the spline transformation settings
spline.param <- attr(B1,'splineInterval')
## use the same settings to apply to the next 50 observations
B2 <- isb(x[-(1:50)],spline.degree=3,knots=spline.param)
```

---

normalize

*Normalize/standardize the columns of a matrix*

---

## Description

Standardize the columns of an attribute matrix *X* to zscores, to the range  $[0 \ 1]$  or a prespecified scale.

## Usage

```
normalize(x,standardize = 'zscore')
```

## Arguments

<i>x</i>	An attribute variable which will be scaled.
<i>standardize</i>	Either a string value denoting a predefined scaling, or a list with values <i>a</i> and <i>b</i> corresponding with the numeric centering and scaling, that is, using the function $x * \text{standardize}\$b - \text{standardize}\$a$ .

## Value

The standardized matrix. The numeric centering and scalings used are returned as attribute "standardize".

## Author(s)

Hok San Yip, Patrick J.F. Groenen, Georgi Nalbantov

## References

P.J.F. Groenen, G. Nalbantov and J.C. Bioch (2008) *SVM-Maj: a majorization approach to linear support vector machines with different hinge errors*.

## See Also

[svmmaj](#)

## Examples

```
## standardize the first 50 objects to zscores
x <- iris$Sepal.Length
x1 <- normalize(x[1:50],standardize='zscore')
## use the same settings to apply to the next 100 observations
x2 <- normalize(x[-(1:50)],standardize=attr(x1,'standardization'))
```

---

predict.svmmaj

*Out-of-Sample Prediction from Unseen Data.*

---

## Description

This function predicts the predicted value (including intercept), given a previous trained model which has been returned by [svmmaj](#).

## Usage

```
## S3 method for class 'svmmaj'
predict(object, X.new , y=NULL, show.plot=FALSE,...)
```

## Arguments

object	Model which has been trained beforehand using <a href="#">svmmaj</a> .
X.new	Attribute matrix of the objects to be predicted, which has the same number of attributes as the untransformed attribute matrix in model.
y	The actual class labels (only if show.plot==TRUE).
show.plot	If show.plot=TRUE, it plots the density of the predicted value for both class labels, if y is not specified, the density of all objects will be plotted.
...	Arguments to be passed to methods.

## Value

The predicted value (including intercept) of class `q.svmmaj`, with attributes:

y	The observed class labels of each object.
yhat	he predicted class labels of each object.
classes	The class labels.

**Author(s)**

Hok San Yip, Patrick J.F. Groenen, Georgi Nalbantov

**References**

P.J.F. Groenen, G. Nalbantov and J.C. Bioch (2008) *SVM-Maj: a majorization approach to linear support vector machines with different hinge errors*.

**See Also**

[svmmaj](#)

**Examples**

```
attach(AusCredit)

## model training
model <- svmmaj(X[1:400,],y[1:400],hinge='quadratic', lambda=1)
## model prediction
q4 <- predict(model,X[-(1:400),],y[-(1:400)],show.plot=TRUE)
q4
```

---

svmmaj

---

*SVM-Maj Algorithm*


---

**Description**

SVM-Maj is an algorithm to compute a support vector machine (SVM) solution. In its most simple form, it aims at finding hyperplane that optimally separates two given classes. This objective is equivalent to finding a linear combination of  $k$  predictor variables to predict the two classes for  $n$  observations. SVM-Maj minimizes the standard support vector machine (SVM) loss function. The algorithm uses three efficient updates for three different situations: primal method which is efficient in the case of  $n > k$ , the decomposition method, used when the matrix of predictor variables is not of full rank, and a dual method, that is efficient when  $n < k$ . Apart from the standard absolute hinge error, SVM-Maj can also handle the quadratic and the Huber hinge.

**Usage**

```
svmmaj(X,y,lambda=1,...)
## Default S3 method:
svmmaj(X, y, lambda=1, weights.obs = 1, weights.var= 1, scale = 'none',
spline.knots=0L, spline.degree=1L, kernel=vanilladot, kernel.sigma=1, kernel.degree=1,
kernel.scale=1, kernel.offset=0, hinge='absolute', hinge.k=5, convergence=1e-8,
print.step=FALSE, initial.point=NULL, increase.step = 20, eps=1e-8,
check.positive=TRUE,na.action=na.omit,...)

plotWeights(object,plotdim=c(3,3))
## S3 method for class 'svmmaj'
plot(x,titletext=NULL,...)
```



**Arguments**

<code>X</code>	A data frame (or object coercible by <a href="#">as.data.frame</a> to a data frame) consisting the attributes, the class of each attribute can be either numeric, logical or factor.
<code>y</code>	A factor (or object coercible by <a href="#">factor</a> to a factor) consisting the class labels.
<code>lambda</code>	Regularization parameter of the penalty term.
<code>weights.obs</code>	Vector of length <code>n</code> with the nonnegative weight for the residual of each object (with length <code>n</code> ). If the length is 2, then it specifies the weight per class.
<code>weights.var</code>	Vector of length <code>k</code> with weights for each attribute.
<code>scale</code>	Specifies whether the columns of attribute matrix <code>X</code> needs to be standardized into zscores or to the interval <code>[0 1]</code> . Possible values are: <code>none</code> , <code>zscore</code> and <code>interval</code> . Moreover, the standardization parameters can be given instead.
<code>spline.knots</code>	Number of internal knots of the spline basis. When the number of knots exceeds the number of (categorical) values of an explanatory variable, the duplicate knots will be removed using <code>link[base]{unique}</code> . <code>spline.knots = 0</code> and <code>spline.degree=1</code> in case of no splines.
<code>spline.degree</code>	The polynomial degree of the splines, for no splines: <code>spline.knots = 0</code> and <code>spline.degree=1</code> .
<code>kernel</code>	Specifies which kernel function to be used (see <a href="#">dots</a> of package <b>kernlab</b> ). Default kernel is the linear kernel .
<code>kernel.sigma</code>	The kernel parameter sigma for the RBF kernel (see <a href="#">rbfdot</a> ). Default value is 1.
<code>kernel.degree</code>	The degree parameter of the polynomial kernel (see <a href="#">polydot</a> ).
<code>kernel.scale</code>	The scale parameter of the polynomial kernel (see <a href="#">polydot</a> ).
<code>kernel.offset</code>	The offset used in a polynomial kernel ( <a href="#">polydot</a> ).
<code>hinge</code>	Specifies which hinge term to be used. Possible values are: <code>absolute</code> , <code>quadratic</code> , <code>huber</code> .
<code>hinge.k</code>	The parameter of the huber hinge, if the huber hinge is used.
<code>convergence</code>	Specifies the convergence criterion of the algorithm. Default is <code>1e-08</code> .
<code>print.step</code>	<code>print.step=TRUE</code> shows the progress of the iteration.
<code>initial.point</code>	Initial solution.
<code>increase.step</code>	The iteration number from which relaxed update will be used.
<code>eps</code>	The relaxation of the majorization function for absolute hinge: $.25 * \text{eps}^{-1}$ is the maximum steepness of the majorization function.
<code>check.positive</code>	Specifies whether a check has to be made for positive input values.
<code>na.action</code>	Generic function for handling NA values.
<code>object</code>	The model returned from <code>svmmaj</code> .
<code>x</code>	The model returned from <code>svmmaj</code> used to plot the distribution of the objects per class using <code>plot.svmmaj</code> .
<code>titletext</code>	An optional title for <code>plot.svmmaj</code> .
<code>plotdim</code>	A vector of the form <code>c(nr, nc)</code> . Subsequent figures will be drawn in an <code>nr</code> -by- <code>nc</code> array on the device.
<code>...</code>	Other arguments passed to methods.

**Value**

Returns a `svmmaj`-class, of which the methods `plot`, `plotWeights`, `summary` and `predict` can be applied. (see also [predict.svmmaj](#)) The returning object consist of the following values:

<code>call</code>	The function specifications which has been called.
<code>lambda</code>	The regularization parameter of the penalty term which has been used.
<code>loss</code>	The corresponding loss function value of the final solution.
<code>iteration</code>	Number of iterations needed to evaluate the algorithm.
<code>X</code>	The attribute matrix of $\dim(X) = c(n,k)$ .
<code>y</code>	The vector of length <code>n</code> with the actual class labels. These labels can be numeric <code>[0 1]</code> or two strings.
<code>classes</code>	A vector of length <code>n</code> with the predicted class labels of each object, derived from <code>q.tilde</code>
<code>Xtrans</code>	The attribute matrix <code>X</code> after standardization and (if specified) spline transformation.
<code>norm.param</code>	The applied normalization parameters (see <a href="#">normalize</a> ).
<code>splineInterval</code>	The spline knots which has been used (see <a href="#">isb</a> ).
<code>splineLength</code>	Denotes the number of spline basis of each explanatory variable in <code>X</code> .
<code>method</code>	The decomposition matrices used in estimating the model.
<code>hinge</code>	The hinge function which has been used (see <a href="#">hinge</a> ).
<code>beta</code>	If identified, the beta parameters for the linear combination (only available for linear kernel).
<code>q</code>	A vector of length <code>n</code> with predicted values of each object including the intercept.
<code>nSV</code>	Number of support vectors.

`plotWeights` shows, one graph per attribute, the weights of all attributes. The type of graph depends on the type of the attribute: the spline line of the corresponding attribute in case a spline has been used, a bar plot for categorical and logical values, and a linear line for all other type of the attribute values. This function cannot be used in a model with a non-linear kernel.

**Author(s)**

Hok San Yip, Patrick J.F. Groenen, Georgi Nalbantov

**References**

P.J.F. Groenen, G. Nalbantov and J.C. Bioch (2008) *SVM-Maj: a majorization approach to linear support vector machines with different hinge errors*.

**See Also**

[dots](#) for the computations of the kernels. [predict.svmmaj](#) [normalize](#) [isb](#) [hinge](#)

## Examples

```
attach(diabetes)

## using default settings
model1 <- model <- svmmaj(X,y,hinge='quadratic', lambda=1)
summary(model1)

weights.obs = list(positive=2,negative=1)
## using radial basis kernel
model2 <- svmmaj(X,y,hinge='quadratic', lambda=1, weights.obs=weights.obs,
                 standardize='interval',kernel=rbfdot,
                 kernel.sigma=1)
summary(model2)
## I-spline basis
model3 <- svmmaj(X,y,,weights.obs=weights.obs,spline.knots=3,spline.degree=2)
plotWeights(model3,plotdim=c(2,4))
```

---

svmmajcrossval	<i>k-fold Cross-Validation of SVM-Maj</i>
----------------	---

---

## Description

This function performs a gridsearch of k-fold crossvalidations using SVM-Maj and returns the combination of input values which has the best forecasting performance.

## Usage

```
svmmajcrossval(X,y,search.grid=list(lambda=2^seq(5,-5,length.out=19)),...,convergence=1e-4,
               weights.obs=1, check.positive = TRUE, print.progress=FALSE, ngroup = 5, groups = NULL)
```

## Arguments

X	A data frame (or object coercible by <a href="#">as.data.frame</a> to a data frame) consisting the attributes.
y	A factor (or object coercible by <a href="#">factor</a> to a factor) consisting the class labels.
search.grid	A list with for each factor the range of values to search for.
...	Other arguments to be passed through svmmaj.
convergence	Specifies the convergence criterion for svmmaj. Default is 1e-08.
weights.obs	Weights for the classes.
check.positive	Specifies whether a check should be performed for positive lambda and weights.obs.
print.progress	=TRUE shows the progress of the cross-validation.
ngroup	The number of groups to be divided into.
groups	A predetermined group division for performing the cross validation.

**Value**

<code>missclass.opt</code>	The minimum (weighted) missclassification rate found in out-of-sample training along the search grid.
<code>param.opt</code>	The level of the factors which gives the minimum loss term value.
<code>param.grid</code>	The matrix of all gridpoints which has been performed during the cross-validation, with its corresponding weighted out-of-sample missclassification rate.

**Author(s)**

Hok San Yip, Patrick J.F. Groenen, Georgi Nalbantov

**References**

P.J.F. Groenen, G. Nalbantov and J.C. Bioch (2008) *SVM-Maj: a majorization approach to linear support vector machines with different hinge errors*.

**See Also**

[svmmaj](#)

**Examples**

```
require(MASS)
Xt <- subset(Pima.tr,select==type)
yt <- Pima.tr$type

## setting grid range
search.grid= list(lambda=10^seq(1,-1) ,kernel.sigma=2^seq(0,2) )

## performing gridsearch with k-fold crossvalidation
results<-svmmajcrossval(Xt,yt,search.grid = search.grid,
standardize = 'interval',kernel = rbfdot,hinge='quadratic')
results
```

---

voting

*Congressional Voting Records Data Set*


---

**Description**

1984 United States Congressional Voting Records; Classify as Republican or Democrat.

**Usage**

```
voting
voting.tr
voting.te
```

**Format**

X is a data frame with 434 congress members and 16 attributes: 16 key votes identified by the Congressional Quarterly Almanac (CQA). All attributes are binary values, with 1= yes and 0= no.

X1	handicapped-infants
X2	water-project-cost-sharing
X3	adoption-of-the-budget-resolution
X4	physician-fee-freeze
X5	el-salvador-aid
X6	religious-groups-in-schools
X7	anti-satellite-test-ban
X8	aid-to-nicaraguan-contras
X9	mx-missile
X10	immigration
X11	synfuels-corporation-cutback
X12	education-spending
X13	superfund-right-to-sue
X14	crime
X15	duty-free-exports
X16	export-administration-act-south-africa

y consists factors which denotes whether the congress memeber is a Republican or a Democrat.

The training set `voting.tr` contains a randomly selected set of 300 subjects, and `voting.te` contains the remaining 134 subjects. `voting` contains all 434 objects.

**Details**

This data set includes votes for each of the U.S. House of Representatives Congressmen on the 16 key votes identified by the CQA. The CQA lists nine different types of votes: voted for, paired for, and announced for (these three simplified to yea), voted against, paired against, and announced against (these three simplified to nay), voted present, voted present to avoid conflict of interest, and did not vote or otherwise make a position known (these three simplified to an unknown disposition).

**Source**

Chih-Chung Chang and Chih-Jen Lin, LIBSVM : a library for support vector machines, 2001. Software available at <http://www.csie.ntu.edu.tw/~cjlin/libsvm>.

**Examples**

```
attach(voting)
summary(X)
summary(y)
```

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