# Package 'Rgtsvm'

## March 19, 2017

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Load SVMlight data file into a sparse matrix.

load.svmlight(filename)

Usage

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#### **Arguments**

filename SVM light filename.

#### **Details**

The file must be symlight format.(http://symlight.joachims.org/)

#### Value

A sparse matrix is returned if the file is loaded or downloaded successfully.

#### Author(s)

```
Zhong Wang (Rinterface) <zw355@cornell.edu>
```

#### **Examples**

```
mat <-load.svmlight("http://www.csie.ntu.edu.tw/~cjlin/libsvmtools/datasets/multiclass/glass.scale")
str(mat);</pre>
```

plot.gtsvm

Plot SVM Objects

#### **Description**

Generates a scatter plot of the input data of a svm fit for classification models by highlighting the classes and support vectors. Optionally, draws a filled contour plot of the class regions.

#### Usage

```
## S3 method for class 'svm'
plot(x, data, formula, fill = TRUE, grid = 50, slice = list(),
symbolPalette = palette(), svSymbol = "x", dataSymbol = "o", ...)
```

#### **Arguments**

X	An object of class svm
data	data to visualize. Should be the same used for fitting.
formula	formula selecting the visualized two dimensions. Only needed if more than two input variables are used.
fill	switch indicating whether a contour plot for the class regions should be added.
grid	granularity for the contour plot.
slice	a list of named values for the dimensions held constant (only needed if more than two variables are used). The defaults for unspecified dimensions are 0 (for numeric variables) and the first level (for factors). Factor levels can either be specified as factors or character vectors of length 1.
symbolPalette	Color palette used for the class the data points and support vectors belong to.
svSymbol	Symbol used for support vectors.
dataSymbol	Symbol used for data points (other than support vectors).
	additional graphics parameters passed to filled.contour and plot.

plot.tune 3

#### Author(s)

```
David Meyer
<David.Meyer@R-project.org>
```

#### See Also

svm

#### **Examples**

plot.tune

Plot Tuning Object

#### **Description**

Visualizes the results of parameter tuning.

#### Usage

#### **Arguments**

x	an object of class tune
type	choose whether a contour plot or a perspective plot is used if two parameters are to be visualized. Ignored if only one parameter has been tuned.
theta	angle of azimuthal direction.
col	the color(s) of the surface facets. Transparent colors are ignored.

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main main title

xlab, ylab titles for the axes. N.B. These must be character strings; expressions are not accepted. Numbers will be coerced to character strings.

swapxy if TRUE, the parameter axes are swaped (only used in case of two parameters).

transform.x, transform.y, transform.z
functions to transform the parameters (x and y) and the error measures (z). Ignored if NULL.

color.palette color palette used in contour plot.

nlevels number of levels used in contour plot.

... Further graphics parameters.

#### Author(s)

```
David Meyer (based on C/C++-code by Chih-Chung Chang and Chih-Jen Lin) <David.Meyer@R-project.org>
```

#### See Also

tune

#### **Examples**

predict.gtsvm

Predict Method for Support Vector Machines

#### **Description**

This function predicts values based upon a model trained by svm in package Rgtsvm.

#### Usage

```
## S3 method for class 'gtsvm'
predict(object, newdata,
    decision.values = FALSE,
    probability = FALSE,
    verbose=FALSE,
    ...,
    na.action = na.omit)
```

#### **Arguments**

object Object of class "gtsvm", created by svm in *Rgtsvm* package.

newdata An object containing the new input data: either a matrix or a sparse matrix (ob-

ject of class Matrix provided by the **Matrix** package, or of class matrix.csr provided by the **SparseM** package, or of class simple\_triplet\_matrix provided by the **slam** package). A vector will be transformed to a n x 1 matrix.

decision.values

Logical controlling whether the decision values of all binary classifiers com-

puted in multiclass classification shall be computed and returned.

probability Logical indicating whether class probabilities should be computed and returned.

Only possible if the model was fitted with the probability option enabled.

verbose logical value indicating whether some alogrithm information(default:FALSE)

... Currently not used.

na.action A function to specify the action to be taken if 'NA's are found. The default

action is na.omit, which leads to rejection of cases with missing values on any required variable. An alternative is na.fail, which causes an error if NA cases

are found. (NOTE: If given, this argument must be named.)

#### Value

A vector of predicted values (for classification: a vector of labels, for density estimation: a logical vector). If score is TRUE, the vector gets a "decision.values" attribute containing a n x c matrix (n number of predicted values, c number of classifiers) of all c binary classifiers' decision values.

#### Note

If the training set was scaled by svm in *Rgtsvm*, the new data is scaled accordingly using scale and center of the training data.

## Author(s)

Zhong Wang (R interface & epe-regression in CUDA) <zw355@cornell.edu> David Meyer (R interface in e1071) <David.Meyer@R-project.org>

Andrew Cotter, Nathan Srebro ,Joseph Keshet (C/C++ code in CUDA) http://ttic.uchicago.edu/textasciitildecotter/proje

#### See Also

svm

svm Training a model of Support Vector Machines by GPU

### **Description**

svm in *Rgtsvm* pakcage is used to train a support vector machine by the C-classfication and epsilon regression method. A formula interface is provided.

#### **Usage**

```
## S3 method for class 'formula'
svm(formula, data = NULL, ..., na.action = na.omit, scale = TRUE);
## Default S3 method:
svm(x,
                  = NULL,
                  = TRUE,
      scale
                  = "C-classification",
      type
                  = "radial",
      kernel
      degree
                  = 3,
      gamma
                  = 0.05,
                  = 0,
      coef0
                  = 1,
      cost
      class.weights= NULL,
      tolerance = 0.001.
                  = 0.1,
      epsilon
      shrinking = TRUE,
      cross
                  = 0,
      probability = FALSE,
                  = TRUE,
      fitted
      rough.cross = 0,
      no.change.x = TRUE,
      verbose
                  = FALSE,
      subset,
      na.action = na.omit)
```

#### **Arguments**

Х

scale

formula a symbolic description of the model to be fit.

an optional data frame containing the variables in the model. By default the data

variables are taken from the environment which 'svm' is called from.

a data matrix, a vector, or a sparse matrix (object of class Matrix provided by the Matrix package, or of class matrix.csr provided by the SparseM package, or of class simple\_triplet\_matrix provided by the slam package).

a response vector with one label for each row/component of x. Can be either a У

factor (for classification tasks) or a numeric vector (for regression).

A logical vector indicating the variables to be scaled. If scale is of length 1, the value is recycled as many times as needed. Per default, data are scaled internally (both x and y variables) to zero mean and unit variance. The center and scale

values are returned and used for later predictions.

only C-classification or eps-regression available. type

kernel the kernel used in training and predicting. You might consider changing some

of the following parameters, depending on the kernel type.

linear: u'v

polynomial:  $(\gamma u'v + coef0)^{degree}$ radial basis:  $e^{(-\gamma |u-v|^2)}$ **sigmoid:**  $tanh(\gamma u'v + coef0)$ 

parameter needed for kernel of type polynomial (default: 3)

gamma parameter needed for all kernels except linear (default: 1/(data dimension)) coef0 parameter needed for kernels of type polynomial and sigmoid (default: 0) cost of constraints violation (default: 1)—it is the 'C'-constant of the regularcost ization term in the Lagrange formulation. class.weights a named vector of weights for the different classes, used for asymmetric class sizes. Not all factor levels have to be supplied (default weight: 1). All components have to be named. tolerance tolerance of termination criterion (default: 0.001) epsilon epsilon in the insensitive-loss function (default: 0.1) shrinking option whether to use the shrinking-heuristics (default: TRUE) cross if a integer value k>0 is specified, a k-fold cross validation on the training data is performed to assess the quality of the model: the accuracy rate for classification and the Mean Squared Error for regression logical indicating whether the fitted values should be computed and included in fitted the model or not (default: TRUE) probability logical indicating whether the model should allow for probability predictions. Number which is less than cross, indicating how many tests are performed for rough.cross cross-validation. The function will return partial tests for cross-validation rather than all repeated tests in order to reduce the running time.

a big matrix, it would be save emeory to use this parameter FALSE. logical value indicating whether some alogrithm information(default:FALSE)

Logical value indicating whther the function can change the x parameter. If x is

... additional parameters for the low level fitting function svm.default

subset An index vector specifying the cases to be used in the training sample. (NOTE:

If given, this argument must be named.)

na.action A function to specify the action to be taken if NAs are found. The default action is

na.omit, which leads to rejection of cases with missing values on any required variable. An alternative is na.fail, which causes an error if NA cases are found.

(NOTE: If given, this argument must be named.)

#### **Details**

no.change.x

verbose

degree

Rgtsvm internally uses a sparse matrix and regular matrix.

If the predictor variables include factors, the formula interface must be used to get a correct model matrix

plot.gtsvm allows a simple graphical visualization of classification models.

#### Value

An object of class "gtsvm" containing the fitted model, including:

SV The resulting support vectors (possibly scaled).

index The index of the resulting support vectors in the data matrix. Note that this index

refers to the preprocessed data (after the possible effect of na.omit and subset)

coefs The corresponding coefficients times the training labels.

#### Author(s)

```
Zhong Wang (R interface & epe-regression in CUDA) <zw355@cornell.edu>
David Meyer (R interface in e1071) <David.Meyer@R-project.org>
Andrew Cotter, Nathan Srebro ,Joseph Keshet (C/C++ code in CUDA) http://ttic.uchicago.edu/textasciitildecotter/proje
```

#### References

- Andrew Cotter, Nathan Srebro, Joseph Keshet. "A GPU-Tailored Approach for Training Kernelized SVMs". 17th ACM SIGKDD Conference on Knowledge Discovery and Data Mining. 2011.
- Chang, Chih-Chung and Lin, Chih-Jen: LIBSVM: a library for Support Vector Machines http://www.csie.ntu.edu.tw/~cjlin/libsvm

#### See Also

```
predict.gtsvm plot.gtsvm matrix.csr (in package SparseM)
```

#### **Examples**

```
data(iris)
attach(iris)
## classification mode
# default with factor response:
model <- svm(Species ~ ., data = iris)</pre>
# alternatively the traditional interface:
x <- subset(iris, select = -Species)</pre>
y <- Species
model <- svm(x, y)
print(model)
summary(model)
# test with train data
pred <- predict(model, x)</pre>
# (same as:)
pred <- fitted(model)</pre>
# Check accuracy:
table(pred, y)
# compute decision values and probabilities:
pred <- predict(model, x, decision.values = TRUE)</pre>
attr(pred, "decision.values")[1:4,]
# visualize (classes by color, SV by crosses):
plot(cmdscale(dist(iris[,-5])),
     col = as.integer(iris[,5]),
     pch = c("o","+")[1:150]
```

tune 9

tune	Parameter Tuning of Functions Using Grid Search

#### **Description**

This generic function tunes hyperparameters of statistical methods using a grid search over supplied parameter ranges.

## Usage

```
tune.svm(method, train.x, train.y = NULL, data = list(), validation.x =
    NULL, validation.y = NULL, ranges = NULL, predict.func = predict,
    tunecontrol = tune.control(), ...)
best.tune(...)
```

#### **Arguments**

method	either the function to be tuned, or a character string naming such a function.
train.x	either a formula or a matrix of predictors.
train.y	the response variable if $train.x$ is a predictor matrix. Ignored if $train.x$ is a formula.
data	data, if a formula interface is used. Ignored, if predictor matrix and response are supplied directly.
validation.x	an optional validation set. Depending on whether a formula interface is used or not, the response can be included in validation.x or separately specified using validation.y.
validation.y	if no formula interface is used, the response of the (optional) validation set.
ranges	a named list of parameter vectors spanning the sampling space. The vectors will usually be created by seq.
predict.func	optional predict function, if the standard predict behavior is inadequate.
tunecontrol	object of class "tune.control", as created by the function tune.control(). If omitted, tune.control() gives the defaults.
	Further parameters passed to the training functions.

#### **Details**

As performance measure, the classification error is used for classification, and the mean squared error for regression. It is possible to specify only one parameter combination (i.e., vectors of length 1) to obtain an error estimation of the specified type (bootstrap, cross-classification, etc.) on the given data set. For convenience, there are several tune.foo() wrappers defined, e.g., for nnet(), randomForest(), rpart(), svm(), and knn().

Cross-validation randomizes the data set before building the splits which—once created—remain constant during the training process. The splits can be recovered through the train.ind component of the returned object.

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#### Value

```
For tune, an object of class tune, including the components:
```

best.parameters

a 1 x k data frame, k number of parameters.

best.performance

best achieved performance.

performances if requested, a data frame of all parameter combinations along with the corre-

sponding performance results.

train.ind list of index vectors used for splits into training and validation sets.

best.model if requested, the model trained on the complete training data using the best pa-

rameter combination.

best.tune() returns the best model detected by tune.

#### Author(s)

```
David Meyer <David.Meyer@R-project.org>
```

#### See Also

```
tune.control, plot.tune, tune.svm, tune.wrapper
```

#### **Examples**

```
data(iris)
## tune `svm' for classification with RBF-kernel (default in svm),
## using one split for training/validation set
obj <- tune(svm, Species~., data = iris,
            ranges = list(gamma = 2^{-1:1}), cost = 2^{(2:4)}),
            tunecontrol = tune.control(sampling = "fix")
           )
## alternatively:
## obj <- tune.svm(Species^{\sim}., data = iris, gamma = 2^{\sim}(-1:1), cost = 2^{\sim}(2:4))
summary(obj)
plot(obj)
## tune `knn' using a convenience function; this time with the
## conventional interface and bootstrap sampling:
x <- iris[,-5]
y <- iris[,5]
obj2 \leftarrow tune.knn(x, y, k = 1:5, tunecontrol = tune.control(sampling = "boot"))
summary(obj2)
plot(obj2)
## tune `rpart' for regression, using 10-fold cross validation (default)
data(mtcars)
obj3 <- tune.rpart(mpg~., data = mtcars, minsplit = c(5,10,15))
summary(obj3)
plot(obj3)
```

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```
## simple error estimation for lm using 10-fold cross validation tune(lm, mpg^{\sim}., data = mtcars)
```

tune.control

Control Parameters for the Tune Function

#### **Description**

Creates an object of class tune.control to be used with the tune function, containing various control parameters.

#### Usage

#### **Arguments**

random if an integer value is specified, random parameter vectors are drawn from the

parameter space.

nrepeat specifies how often training shall be repeated.

repeat.aggregate

function for aggregating the repeated training results.

sampling

cross

sampling scheme. If sampling = "cross", a cross-times cross validation is performed. If sampling = "boot", nboot training sets of size boot.size (part) are sampled (with replacement) from the supplied data. If sampling = "fix", a single split into training/validation set is used, the training set containing a fix part of the supplied data. Note that a separate validation set can be supplied via validation.x and validation.y. It is only used for sampling = "boot" and sampling = "fix"; in the latter case, fix is set to 1.

sampling.aggregate,sampling.dispersion

functions for aggregating the training results on the generated training samples (default: mean and standard deviation).

number of partitions for cross-validation.

fix part of the data used for training in fixed sampling.

nboot number of bootstrap replications.

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boot.size size of the bootstrap samples.

best.model if TRUE, the best model is trained and returned (the best parameter set is used for

training on the complete training set).

performances if TRUE, the performance results for all parameter combinations are returned.

rough.cross integer.

error.fun function returning the error measure to be minimized. It takes two arguments:

a vector of true values and a vector of predicted values. If NULL, the misclassification error is used for categorical predictions and the mean squared error for

numeric predictions.

#### Value

An object of class "tune.control" containing all the above parameters (either the defaults or the user specified values).

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#### See Also

tune

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