# Package 'Rgtsvm'

# March 14, 2018

Title The e1071 compatibility SVM package for GPU architecture based on the GTSVM software

Description The e1071 compatibility SVM package is built on the CUDA-enable GPU architec-

silon regression; 4 kernel functions; K-fold cross validation; big matrix handle.

ture with the following features, including: binary classification, multiclass classification and ep-

Imports graphics, grDevices, bit64, tools, methods, stats, utils, class

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Suggests SparseM, Matrix

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attach.bigmatrix

Wrapping a big matrix into a reference class

#### **Description**

This function wraps a big matrix into a reference class in order to avoid multiple variable copying when variables are passed into deep calls in R. It seems to become pointer calling in C language.

# Usage

```
attach.bigmatrix(data)
```

#### **Arguments**

data

matrix object

#### Value

Return a reference class with the name "BigMatrix.refer". It can be used in svm and predict calling in Rgtsvm.

#### See Also

```
load.bigmatrix
```

```
library(mvtnorm);
size=5000;
dimension=100:
covar.mat <- matrix(runif(dimension*dimension), nrow=dimension);</pre>
covar.mat <- t(covar.mat)</pre>
covar.mat <- round( (covar.mat + t(covar.mat))/2, 4);</pre>
zero <- rmvnorm(size, mean=c(1:dimension), sigma= covar.mat);</pre>
one <- rmvnorm(size, mean=c(1:dimension)-5, sigma= covar.mat);</pre>
x <- rbind(zero,one);</pre>
y <- c(rep(0,nrow(zero)),rep(1,nrow(one)));</pre>
i.all <- sample(1:(2*size));</pre>
i.training <- i.all[(1:round(2*size*0.8))];</pre>
i.test <- i.all[-c(1:round(2*size*0.8))];</pre>
bigm.x <- attach.bigmatrix( data = x[ i.training,]);</pre>
model.gpu <- svm(bigm.x,y[ i.training ],type="C-classification");</pre>
y.pred <- predict(model.gpu,x[i.test,]);</pre>
cat("accuracy", sum(y.pred==y[i.test])/length(i.test),"\n");
```

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getGPUcount

Checking the count of GPU device.

## Usage

```
getGPUcount()
```

## Value

The count of GPU device.

#### See Also

```
selectGPUdevice,resetGPUdevice,
```

# **Examples**

```
cat("GPU count:", getGPUcount(), "\n");
```

load.bigmatrix

Loading a big matrix from RData or RDS file

## Description

Creating a big matrix based on the matrix variable in a RData file or a RDS file.

#### Usage

```
load.bigmatrix(file.data, variable = NULL)
```

## **Arguments**

file.data File name, RData file or RDS file.

variable String, variable name in the RData file. If variable is NULL, the data file should

be in RDS format.

## Value

Return a reference class with the name "BigMatrix.refer". It can be used in svm and predict calling in Rgtsvm.

## See Also

```
attach.bigmatrix
```

```
# The example can not be executed!
#
# x0_bm <- load.bigmatrix("X0.RDS")
# x1_bm <- load.bigmatrix("X1.Rdata", "x1")</pre>
```

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load.svmlight

Load SVMlight data file into a sparse matrix.

#### **Description**

Load SVMlight data file into a sparse matrix.

## Usage

```
load.svmlight(filename, .loadbyC = TRUE)
```

## **Arguments**

filename string, SVM light filename.

.loadbyC logical value, indicating whether loading data in C or R.

#### **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

Scatter plot for classification models

## **Description**

A scatter figure is generated for the classification model trained by the svm. The figure only shows two feature columns within training data. Optionally, a filled contour is predicted and added to the class regions.

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#### **Arguments**

Х an object of class gtsvm generated by svm. data data frame with the training data. formula formula object indicating two feature columns used as the X axis and Y axis if more than two feature columns in the training data. fill logical value indicating whether a contour plot for the class regions is added. grid integer value indicating the grid number to generating the the contour if fill = TRUE. slice a list of feature columns that are constant during the predicting for the grid contour if training data has more than two feature columns. all feature columns but two feature columns used in the X axis and Y axis must be held constant to generate a grid contour prediction.. In order to generate a contour, except two feature columns of X axis and Y axis, other feature columns hold constant values for predicting. Default unspecified feature columns use 0 for numeric variables and the first level for factors. Color palette indicating the color of the class data points and support vectors symbolPalette belong to. character or integer code indicating plot character or symbol for support vectors, svSymbol check the available values in pch in points function. character or integer code indicating plot character or symbol for non-support dataSymbol vectors, check the available values in pch in points function.

additional graphics parameters passed to filled.contour or plot.

## Author(s)

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

## See Also

svm

plot.tune

plot.tune	Plot performance of tuning object
•	1 3 3 5 3

## **Description**

Contour or perspective plot to visualize the performace of tuning results. It is limited to two tuning parameters.

# Usage

## **Arguments**

x	an object of class tune generated by the tune.svm
type	string indicating whether a contour plot or a perspective plot is used if two parameters shall be plotted. Ignored if only one tuning parameter in the x.
theta	numeric value indicating angle of azimuthal direction for perspective plot.
col	the color(s) of the surface facets for the perspective plot. Ignored if transparent colors are specified.
main	string indicating main title.
xlab	string or expression indicating titles for the X axes.
ylab	string or expression indicating titles for the Y axes.
swapxy	logical value indicating whether the parameter axes are swapped, only used in case of two parameters.
transform.x	function indicating how to transform the first parameter in $\boldsymbol{X}$ axis. Ignored if NULL.
transform.y	function indicating how to transform the second parameter in $\boldsymbol{Y}$ axis. Ignored if NULL.
transform.z	function indicating how to transform the error measures in $Z$ axis in the perspective plot or color palette in the contour plot. Ignored if NULL.
color.palette	color palette used in the contour plot.
nlevels	integer value indicating number of levels used in the contour plot.
• • •	additional graphics parameters passed to filled.contour or persp.

# Author(s)

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

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#### See Also

```
tune.svm
```

# **Examples**

predict.gtsvm

Predict method for SVMs on CUDA-enabled GPU

## **Description**

This function performs prediction based on a SVM model trained by svm in package Rgtsvm.

#### Usage

## Arguments

object an object of class "gtsvm" returned by svm in Rgtsvm package.

newdata data frame, or matrix, or sparse matrix of the test data. A vector test data must

transform to a n x 1 matrix.

decision.values

logical value indicating whether the decision values of binary classfication or

multiclass classification shall be returned. Only valid for classification.

probability logical value indicating whether class probabilities should be computed and re-

turned.

gpu.id integer value indicating GPU device ID, starting from 0. No GPU device is

changed by default.

verbose logical value indicating whether some alogrithm information is output into the

R console, default is FALSE.

... Unused currently.

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.

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#### Value

A vector of predicted values or labels are returned.

If decision.values is required, the vector has a "decision.values" attribute containing a decision matrix with number of samples in rows and number of classes and in columns.

If probability is required, the vector has a "probabilities" attribute containing a decision matrix with number of samples in rows and number of classes and in columns, only valid for C-classification.

#### Note

The test data can be an object containing the new input data: data frame, or matrix, or a sparse matrix. The sparse matrix can be defined by the class Matrix provided by the Matrix package, or the class matrix.csr provided by the SparseM package, or the class simple\_triplet\_matrix provided by the slam package)

If the training data was scaled by the svm calling, this function shall scale the test data accordingly using scale and center of the training data.

#### Author(s)

```
Zhong Wang (R interface & eps-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/~cotter/projects/gtsvm/
```

#### See Also

svm

predict.load

Load and initialize a training model on CUDA-enabled GPU

## Description

This function loads and initializes a SVM model trained by svm in package *Rgtsvm*. This function works with predict.run and predict.unload to implement the phased prediction with 3 steps: model loading, multiple predicting call, removing model from GPU. It is suitable for the prediction with huge samples using a big training model.

#### Usage

```
predict.load(object, gpu.id=NULL, verbose=FALSE )
```

## **Arguments**

object an object of class "gtsvm" returned by svm in Rgtsvm package.

gpu.id a vector indicating multiple GPU device IDs are used for the prediction in par-

allel.

verbose logical value indicating whether some alogrithm information is output into the

R console, default is FALSE.

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#### Value

A slim model with one external pointer returned by the C/C++ functions is returned. The feature vectors and fitting information are removed in order to reduce the memory usage.

## Author(s)

Zhong Wang (R interface & eps-regression in CUDA) <zw355@cornell.edu>

#### See Also

```
predict.gtsvm
```

predict.run

Run prediction using a loaded model on CUDA-enabled GPU

## **Description**

This function performs the phased prediction using 3 steps, which are model loading, prediction based on model, and model unloading. This prediction step can be called by multiple times.

## Usage

```
predict.run(object, newdata,
    decision.values = FALSE,
    probability = FALSE,
    verbose=FALSE,
    ...,
    na.action = na.omit)
```

## **Arguments**

object an object of class "gtsvm" returned by predict.load in *Rgtsvm* package.

newdata data frame, or matrix, or sparse matrix of the test data. A vector test data must

transform to a n x 1 matrix.

decision.values

logical value indicating whether the decision values of binary classfication or

multiclass classification shall be returned. Only valid for classification.

probability logical value indicating whether class probabilities should be computed and re-

turned.

verbose logical value indicating whether some alogrithm information is output into the

R console, default is FALSE.

... Unused currently.

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.

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#### Value

A vector of predicted values or labels are returned. The structue is same as predict.gtsvm.

If decision.values is required, the vector has a "decision.values" attribute containing a decision matrix with number of samples in rows and number of classes and in columns.

If probability is required, the vector has a "probabilities" attribute containing a decision matrix with number of samples in rows and number of classes and in columns, only valid for C-classification.

#### Note

```
See predict.gtsvm
```

#### Author(s)

Zhong Wang (R interface & eps-regression in CUDA) <zw355@cornel1.edu>

#### See Also

```
predict.load, predict.unload
```

## **Examples**

```
data(iris)
attach(iris)

## classification training
model <- svm(Species ~ ., data = iris)

object <- predict.load(model);
r <- predict.run(object, newdata = iris, decision.values = TRUE )
predict.unload(object);
show(r);</pre>
```

predict.unload

Unload SVM object allocated on CUDA-enabled GPU

## **Description**

This function is cleanup step in the phased prediction which is suitable for the multiple predicting calls using a big model.

```
predict.unload(object, verbose=FALSE )
```

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## **Arguments**

object an object of class "gtsvm" returned by predict.load in *Rgtsvm* package.

verbose logical value indicating whether some alogrithm information is output into the

R console, default is FALSE.

#### Value

None. The procedure will be interrupted if any errors happen.

## Author(s)

Zhong Wang (R interface & eps-regression in CUDA) <zw355@cornell.edu>

#### See Also

```
predict.run, predict.load
```

resetGPUdevice

Reset GPU device.

## Usage

```
resetGPUdevice()
```

## Value

This function returns error code, which 0 indicates success otherwise indicates failure.

## See Also

selectGPUdevice

## **Examples**

resetGPUdevice();

 ${\tt selectGPUdevice}$ 

Selecting GPU device to run training or prediction.

# Description

It is useful to select one GPU devices to run if multiple GPU devices are on the host.

```
selectGPUdevice( gpu.id )
```

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## **Arguments**

gpu.id integer value indicating GPU device ID, starting from 0. No GPU device is changed by default.

#### Value

This function returns error code, which 0 indicates success otherwise indicates failure.

## See Also

```
svm,predict.gtsvm, predict.loadsvm
```

#### **Examples**

```
selectGPUdevice(0);
```

svm

Training a model of SVMs on CUDA-enabled GPU

# Description

svm in the **Rgtsvm** pakcage is used to train a support vector machine by C-classfication or epsilon regression on CUDA-enabled GPU.

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

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```
...,
subset,
na.action = na.omit)
```

# Arguments

guinents	
formula	a formula object describing the training model.
data	an optional data frame containing the variables in the model. By default the variables are taken from the environment which 'svm' is called from.
X	a data matrix, or a vector, or a sparse matrix as training data.
у	a factor (for C-classification) or a numeric vector (for eps-regression) specifying the response for each row of x.
scale	logical value or a logical vector, indicating whether the feature columns are scaled. By default, both x and y variables are scaled to zero mean and unit variance. The center and scale values are returned and can be used for scaling new test data.
type	string indicating the SVM method, only C-classification or eps-regression available, default is C-classification
kernel	the kernel function used in training and predicting, four options are available (attached in details), default is radial.
degree	integer value indicating parameter value used in kernel of type polynomial, default is 3.
gamma	numerical value indicating parameter value needed for all kernels except linear, default is 1/(number of feature vector)).
coef0	numerical value indicating parameter value used in kernels of type polynomial and sigmoid, default is $0$ .
cost	numerical value indicating regularization term in the Lagrange formulation, which is cost of constraints violation, default is 1.
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	numerical value indicating the tolerance of termination criterion for the training algorithm, default is 0.001.
epsilon	numerical value indicating epsilon in the insensitive-loss function for the eps-regression method, default is $0.1$ .
shrinking	logical value indicating whether to use the shrinking-heuristics, default is TRUE.
cross	integer value indicating whether a k-fold cross validation on the training data is performed to assess the quality of the model. Ignored if corss=0.
rough.cross	integer value which is less than cross, indicating how many tests are performed for cross-validation. The function will return partial tests for cross-validation rather than all repeated tests in order to reduce the running time.
fitted	logical value indicating whether the prediction should be performed and returned in the function calling, default is TRUE.
probability	logical value indicating whether the model should allow for probability predictions.
no.change.x	logical value indicating whither the function can change the x parameter. It would save CPU memory if this parameter is assigned to FALSE for the big matrix x.

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gpu.id integer value indicating GPU device ID, starting from 0. No GPU device is changed by default. maxIter integer value indicating maximum iteration, default is the number of feature elements (NROW(x)\*NCOL(x))logical value indicating whether some alogrithm information is output on R converbose sole, default is FALSE. additional parameters for the low level fitting function svm.default subset a vector of index values 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.

#### Details

**Rgtsvm** uses a sparse matrix and regular matrix. The sparse matrix can be defined by the class Matrix provided by the **Matrix** package, or the class matrix.csr provided by the **SparseM** package, or the class simple\_triplet\_matrix provided by the **slam** package)

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

The kernel function has 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:  $tanh(\gamma u'v + coef0)$ 

plot.gtsvm provides a simple visualization method for the classification model trained by svm.

#### Value

This function returns a trained model. It is an object of class "gtsvm" with following properties:

total.nSV the total number of support vectors.

nSV the number of support vectors in each group.

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 multiply the training labels.

rho the negative intercept.

fitted the prediction values if fitted=TRUE

fitted.accuracy

The predicted accuracy for C-classification if fitted=TRUE.

fitted.MSE the mean square error for eps-regression if fitted=TRUE.

fitted.r2 the R square for eps-regression if fitted=TRUE.

residuals the difference between the true values and prediction for eps-regression if

 ${\tt fitted=TRUE}.$ 

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**MSE** the mean square errors at each cross-validation test for eps-regression if cross>0. tot.MSE the mean square error in cross-validation for eps-regression if cross>0. scorrcoeff the R square in cross-validation for eps-regression if cross>0. the accuracies at each cross-validation test for C-classification if cross>0. accuracies the total accuracies in cross-validation for C-classification if cross>0. tot.accuracy logical values if probability is predicted. compprob numerical value indicating (1) the parameter A of the logistic distributions fitprobA ted to the decision values of the binary classifiers (2) the \gamma coefficient of the softmax function for the multi-classification in one-agianst-rest mode. (3) the \zeta parameter of the Laplace distributions for the eps-regression if probability==TRUE. probB

numerical value indicating the parameter B of the logistic distributions fitted to

the decision values of the binary classifiers if probability==TRUE.

#### 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/~cotter/projects/gtsvm/
```

#### 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)
```

```
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)
```

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tune.control

Control parameters for the tune function

## **Description**

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

# Usage

```
tune.control(random = FALSE,
    nrepeat = 1,
    repeat.aggregate = mean,
    sampling = c("cross", "fix", "bootstrap"),
    sampling.aggregate = mean,
    sampling.dispersion = sd,
    cross = 10,
    fix = 2/3,
    nboot = 10,
    boot.size = 9/10,
    best.model = TRUE,
    performances = TRUE,
    rough.cross = 0,
    error.fun = NULL)
```

## Arguments

random FALSE or an integer value, indicating whether or how many parameter combina-

tions are drawn from the parameter space.

nrepeat integer value, specifies how many replication of training and prediction shall be

repeated, default is 1.

repeat.aggregate

function, used to aggregate the replicates results, default is mean.

sampling string value indicating sampling scheme, three sampling methods are available,

default is cross.

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sampling.aggregate

function to aggregate the test results for each parameter combination, default is

mean.

sampling.dispersion

function to disperse the test results for each parameter combination, default is

sd.

cross integer value, used if sampling = "cross", indicating the number of partitions

for cross-validation, default is 10.

fix numeric value, used if sampling = "fix", part of the data used for training in

fixed sampling., default is 2/3

nboot integer value, used if sampling = "boot", indicating number of bootstrap

replications.

boot.size numeric value, used if sampling = "boot", indicating size of the bootstrap

samples.

best.model logical value, indicating whether the best model is returned based on the the best

parameter set on the complete training set).

performances logical value, indicating whether the performance results for all parameter com-

binations are returned.

rough.cross integer value if sampling = "cross", indicating how many training and test-

ing for cross-validation are conducted, it is intended to reduce the computation time by decreasing the cross-validation jobs while maintaining the higher cross

number.

error.fun function for the error measure, 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.

## **Details**

Three options for sampling are available:

1: sampling = "cross", a cross-times cross validation.

2: sampling = "boot", sampling with replacement on the nboot training sets of boot.size (part).

3: sampling = "fix", the dataset is split into a training set of size fix and validation set.

# Value

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

#### Author(s)

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

#### See Also

tune.svm

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tune.svm	Parameter tuning function using grid search

## **Description**

tune.svm() tunes hyperparameters of statistical methods using a grid search over supplied parameter ranges. best.tune() returns the best model detected by tune.svm.

# Usage

```
tune.svm(x, y = NULL, data = NULL, degree = NULL, gamma = NULL,
      coef0 = NULL, cost = NULL, nu = NULL,
      class.weights = NULL, epsilon = NULL,...)
best.tune(...)
```

# **Arguments**

x	formula object or training matrix.
у	vector indicating the response variable only if train.x is not a formula object, otherwise ignore.
data	data frame only if the formula interface is used in train.x, otherwise ignore.
degree	a numeric vector indicating the tuning values for the parameter degree.
gamma	a numeric vector indicating the tuning values for the parameter gamma.
coef0	an numeric vector indicating the tuning values for the parameter coef0.
cost	a numeric vector indicating the tuning values for the parameter cost.
nu	a numeric vector indicating the tuning values for the parameter nu.
class.weights	a numeric vector indicating the tuning values for the parameter class.weights.
epsilon	a numeric vector indicating the tuning values for the parameter epsilon.
	Further parameters passed to the training functions, i.e. svm, tune.

## **Details**

To measure performance, classification error and mean squared error can be used for the classificaition and epsilon regression respectively.

### Value

This function return an object of class tune, including the components:

```
best.parameters
                  a 1 x k data frame, k number of parameters.
best.performance
                  value, 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.
```

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# Author(s)

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

## See Also

```
tune.control, plot.tune
```

```
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, gamma = 2^(-1:1), cost = 2^(2:4))
summary(obj)
plot(obj)</pre>
```

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