Package 'probsvm'

May 17, 2013

Title probsvm: Class proba	bility estimation for Support Vector Machines	
Date 2013-5-16		
Version 1.00		
Author Chong Zhang, Seung Jun Shin, Junhui Wang, Yichao Wu, Hao Helen Zhang, and Yufeng Liu		
Maintainer Chong Zhang	<pre><chongz@live.unc.edu></chongz@live.unc.edu></pre>	
Depends kernlab		
Description This package p is distributional assum	provides multiclass conditional probability estimation for the SVM, which aption free.	
License GPL-2		
LazyLoad yes		
R topics document	ed:	
r ····r		1
Index		5
predict.probsvm	Prediction function that provides class label prediction and class conditional probability estimation for objects with class "probsvm", returned by the probsvm function.	
		_

Description

Type Package

Prediction of a test data set on the obtained probsvm model. See the description of probsvm for details.

Usage

```
## S3 method for class 'probsvm':
predict(object,new.x=NULL,...)
```

2 probsvm

Arguments

An object returned by a call of probsvm function.

The new predictor matrix. The number and order of predictors in new.x should be the same as those of x, which is used in the call of probsvm function. If not specified, the program uses the x matrix as the prediction object.

Not used.

Value

object The model from a probsvm function.

new.x The predictor matrix used for prediction.

pred.prob The predicted class conditional probability. The class of each column is recorded in the column names.

pred.y Predicted label for new.x.

See Also

probsvm

probsvm Main function that provides models for multiclass conditional probability estimation and label prediction

Description

The function uses x and y to build a multiclass prediction model. The multiclass method can be either one-versus-one (ovo), or one-versus-rest (ovr, default). The probability estimation method for a binary classifier is proposed in Wang et al. (2008), and then we use the method proposed in Wu et al. (2004) to combine the results from binary classifiers in the ovo method. For the ovr method, we rescale those binary probabilities for multiclass problems. The function automatically chooses the best penalty parameter via 5-fold (default) Cross Validations (CV). Linear kernel, polynomial kernel and radial(Gaussian) kernel are available. A solution path is provided by Shin et al. (2012), which boosts the computational speed.

Usage

```
probsvm(x, y, fold=5,
kernel=c("linear", "polynomial", "radial"),
kparam=NULL, Inum=20, type="ovr",
lambdas=2^(-10:10))
```

Arguments

X	The x matrix/data.frame for the training dataset. Columns represent the covariates, and rows represent the instances. There should be no NA/NaN values in x.
У	The labels for the training dataset.
fold	Number of folds in CV. Default 5.

probsvm 3

kernel	Type of kernel used for learning. kernel="linear" for linear kernel (default), kernel="polynomial" for polynomial kernel, and kernel="radial" for radial(gaussian) kernel.
kparam	The parameter for the kernel. For linear kernel, this argument is not needed. In polynomial kernel, it represents the degree of the polynomials, and in radial(gaussian) kernel, it represents the usual sigma value.
Inum	Number of knots on [0,1] to estimate the class conditional probability. The larger Inum is, the more accurate the final result is, yet the more time it takes to compute. Default 20.
type	The type of multiclass method. The option ovo is for the one-versus-one method, and ovr is for the one-versus-rest method (default).
lambdas	The user-specified lambda value vector. Each element should be positive. Default $2^{(-10:10)}$.

Value

All arguments

All arguments are returned.

lambdas The lambda values used for selecting the best model. Sorted in an increasing

order.

best.lambda The best lambda values selected from CV. Used for model prediction.

call The call of probsvm.

Author(s)

Chong Zhang, Seung Jun Shin, Junhui Wang, Yichao Wu, Hao Helen Zhang, and Yufeng Liu

References

Shin, S.J., Y. Wu, and H.H. Zhang (2012). Two-Dimensional Solution Surface for Weighted Support Vector Machines, *Journal of Computational and Graphical Statistics*, in press.

Wang, J., X. Shen, and Y. Liu (2008). Probability estimation for large margin classifiers. *Biometrika* **95(1)**, 149-167.

Wu, T.-F., C.-J. Lin, and R. C. Weng (2004). Probability estimates for multi-class classification by pairwise coupling. *Journal of Machine Learning Research* **5**, 975-1005.

See Also

```
predict.probsvm
```

Examples

```
# iris data #
data(iris)
iris.x=iris[c(1:20,51:70,101:120),-5]
iris.y=iris[c(1:20,51:70,101:120),5]
iris.test=iris[c(21:50,71:100,121:150),-5]
```

4 probsvm

```
a = probsvm(iris.x,iris.y,type="ovo",
Inum=10,fold=2,lambdas=2^seq(-10,10,by=3))
predict(a, iris.test)
```

Index

predict.probsvm, 1, 3probsvm, 2, 2