# Package: dcsvm (via r-universe)

January 11, 2025

Type Package

**Title** Density Convoluted Support Vector Machines

Version 0.0.1

Date 2025-01-08

Description Implements an efficient algorithm for solving sparse-penalized support vector machines with kernel density convolution. This package is designed for high-dimensional classification tasks, supporting lasso (L1) and elastic-net penalties for sparse feature selection and providing options for tuning kernel bandwidth and penalty weights. The 'dcsvm' is applicable to fields such as bioinformatics, image analysis, and text classification, where high-dimensional data commonly arise. Learn more about the methodology and algorithm at Wang, Zhou, Gu, and Zou (2023) <doi:10.1109/TIT.2022.3222767>.

**Depends** Matrix

Imports grDevices, graphics, methods, stats

License GPL-2

**NeedsCompilation** yes

**Encoding** UTF-8

RoxygenNote 7.3.2

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**Date/Publication** 2025-01-10 21:20:02 UTC

Repository https://boxiang-wang.r-universe.dev

RemoteUrl https://github.com/cran/dcsvm

RemoteRef HEAD

**RemoteSha** 5a439b70d0dbc4ef00bb398fa774552a0c49b049

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# Description

This package provides tools to perform density-convoluted support vector machine (DCSVM) modeling for high-dimensional data classification.

## **Details**

This package implements the density-convoluted SVM for high-dimensional classification.

Package: dcsvm Type: Package Version: 0.0.1 Date: 2025-01-08 License: GPL-2

The package dcsvm contains five main functions:

- dcsvm
- cv.dcsvm
- coef.dcsvm
- plot.dcsvm
- plot.cv.dcsvm

#### Author(s)

Boxiang Wang, Le Zhou, Yuwen Gu, and Hui Zou Maintainer: Boxiang Wang <boxiang-wang@uiowa.edu> coef.cv.dcsvm 3

#### References

Wang, B., Zhou, L., Gu, Y., and Zou, H. (2023) Density-Convoluted Support Vector Machines for High-Dimensional Classification, IEEE Transactions on Information Theory, Vol. 69(4), 2523-2536,

coef.cv.dcsvm

Compute Coefficients from a "cv.dcsvm" Object

# Description

Computes the coefficients at specified lambda values for a cv.dcsvm object.

## Usage

```
## S3 method for class 'cv.dcsvm'
coef(object, s = c("lambda.1se", "lambda.min"), ...)
```

#### **Arguments**

object	A fitted cv. dcsvm object, obtained by conducting cross-validation on the sparse density-convoluted SVM model.
S	Value(s) of the L1 tuning parameter lambda for computing coefficients. Default is "lambda.1se", the largest lambda value achieving a cross-validation error within one standard error of the minimum. Alternatively, "lambda.min" corresponds to the lambda incurring the least cross-validation error. s can also be numeric, specifying the value(s) to use.
	Other arguments that can be passed to dcsvm.

# Details

Compute Coefficients from a "cv.dcsvm" Object

Computes coefficients at chosen values of lambda from the cv.dcsvm object.

This function computes the coefficients for lambda values suggested by cross-validation.

#### Value

The returned object depends on the choice of s and any additional arguments passed to the dcsvm method.

#### See Also

```
cv.dcsvm and predict.cv.dcsvm methods.
```

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## **Examples**

```
data(colon) colonx[ ,1:100] # Use only the first 100 columns for this example set.seed(1) cv <- cv.dcsvm(colonx, colon, lam2=1, nfolds=5) c1 <- coef(cv, s="lambda.1se")
```

coef.dcsvm

Compute Coefficients for Sparse Density-Convoluted SVM

## **Description**

Computes the coefficients or indices of nonzero coefficients at specified lambda values from a fitted dcsym model.

## Usage

```
## S3 method for class 'dcsvm'
coef(object, s = NULL, type = c("coefficients", "nonzero"), ...)
```

## **Arguments**

object	A fitted dcsvm object.
S	Value(s) of the L1 tuning parameter lambda for computing coefficients. Default is the entire lambda sequence obtained by dcsvm.
type	"coefficients" or "nonzero"? "coefficients" computes the coefficients at given values for s; "nonzero" returns a list of the indices of the nonzero coefficients for each value of s. Default is "coefficients".
	Not used. Other arguments to predict.

# **Details**

Compute Coefficients for Sparse Density-Convoluted SVM

Computes the coefficients or returns the indices of nonzero coefficients at chosen values of lambda from a fitted dcsvm object.

s is the vector of lambda values at which predictions are requested. If s is not in the lambda sequence used for fitting the model, the coef function uses linear interpolation. The new values are interpolated using a fraction of coefficients from both left and right lambda indices.

## Value

Either the coefficients at the requested values of lambda, or a list of the indices of the nonzero coefficients for each lambda.

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#### See Also

```
predict.dcsvm
```

### **Examples**

```
data(colon)
fit <- dcsvm(colon$x, colon$y, lam2=1)
c1 <- coef(fit, type="coefficients", s=c(0.1, 0.005))
c2 <- coef(fit, type="nonzero")</pre>
```

colon

Simplified Gene Expression Data from Alon et al. (1999)

## Description

This dataset contains 62 colon tissue samples with 2000 gene expression levels. Among these samples, 40 are tumor tissues (coded as 1) and 22 are normal tissues (coded as -1).

## Usage

```
data(colon)
```

#### **Details**

Simplified Gene Expression Data from Alon et al. (1999)

Gene expression data (2000 genes for 62 samples) from a DNA microarray experiment of colon tissue samples (Alon et al., 1999).

## Value

A list with the following elements:

A matrix of 62 rows and 2000 columns representing the gene expression levels of 62 colon tissue samples. Each row corresponds to a sample, and each column corresponds to a gene.

y A numeric vector of length 62 representing the tissue type (1 for tumor; -1 for normal).

## Source

The data were introduced in Alon et al. (1999).

#### References

Alon, U., Barkai, N., Notterman, D.A., Gish, K., Ybarra, S., Mack, D., and Levine, A.J. (1999). "Broad patterns of gene expression revealed by clustering analysis of tumor and normal colon tissues probed by oligonucleotide arrays," *Proceedings of the National Academy of Sciences*, **96**(12), 6745–6750.

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## **Examples**

```
# Load the dcsvm library
library(dcsvm)

# Load the dataset
data(colon)

# Check the dimensions of the data
dim(colon$x)

# Count the number of samples in each class
sum(colon$y == -1)
sum(colon$y == 1)
```

cv.dcsvm

Cross-Validation for Sparse Density-Convoluted SVM

# Description

Performs cross-validation for the sparse density-convoluted SVM to estimate the optimal tuning parameter lambda.

## Usage

```
cv.dcsvm(x, y, lambda = NULL, hval = 1,
  pred.loss = c("misclass", "loss"), nfolds = 5, foldid, ...)
```

## **Arguments**

X	A matrix of predictors, i.e., the x matrix used in dcsvm.
у	A vector of binary class labels, i.e., the y used in dcsvm.
lambda	Default is NULL, and the sequence generated by dcsvm is used. User can also provide a new lambda sequence for cross-validation.
hval	The bandwidth parameter for kernel smoothing. Default is 1.
pred.loss	"misclass" for classification error, "loss" for the density-convoluted SVM loss.
nfolds	The number of folds. Default is 5. The allowable range is from 3 to the sample size. Larger nfolds increases computational time.
foldid	An optional vector with values between 1 and nfold, representing the fold indices for each observation. If supplied, nfolds can be missing.
	Other arguments that can be passed to dcsvm.

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## **Details**

Cross-Validation for Sparse Density-Convoluted SVM

Conducts a k-fold cross-validation for dcsvm and returns the suggested values of the L1 parameter lambda.

This function runs dcsvm on the sparse density-convoluted SVM by excluding each fold in turn, then computes the mean cross-validation error and standard deviation. It is adapted from the cv functions in the gcdnet and glmnet packages.

#### Value

A cv. dcsvm object is returned, which includes the cross-validation fit:

lambda	The lambda sequence used in dcsvm.
CVM	A vector of length length(lambda) for the mean cross-validated error.
cvsd	A vector of length length(lambda) for estimates of standard error of cvm.
cvupper	The upper curve: cvm + cvsd.
cvlower	The lower curve: cvm - cvsd.
nzero	Number of non-zero coefficients at each lambda.
name	"Mis-classification error", for plotting purposes.
dcsvm.fit	A fitted dcsvm object using the full data.
lambda.min	The lambda incurring the minimum cross-validation error cvm.
lambda.1se	The largest value of lambda such that error is within one standard error of the minimum.
cv.min	The minimum cross-validation error.
cv.1se	The cross-validation error associated with lambda.1se.

#### See Also

dcsvm, plot.cv.dcsvm, predict.cv.dcsvm, and coef.cv.dcsvm methods.

```
data(colon)
colon$x <- colon$x[ ,1:100] # Use only the first 100 columns for this example
n <- nrow(colon$x)
set.seed(1)
id <- sample(n, trunc(n / 3))
cvfit <- cv.dcsvm(colon$x[-id, ], colon$y[-id], lam2=1, nfolds=5)
plot(cvfit)
predict(cvfit, newx=colon$x[id, ], s="lambda.min")</pre>
```

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dcsvm

Density-Convoluted Support Vector Machine

## **Description**

Fits the density-convoluted support vector machine (DCSVM) through kernel density convolutions.

# Usage

```
dcsvm(
 Х,
 у,
 nlambda = 100,
 lambda.factor = ifelse(nobs < nvars, 0.01, 1e-04),</pre>
  lambda = NULL,
  lam2 = 0,
  kern = c("gaussian", "uniform", "epanechnikov"),
 hval = 1,
 pf = rep(1, nvars),
 pf2 = rep(1, nvars),
 exclude,
 dfmax = nvars + 1,
 pmax = min(dfmax * 1.2, nvars),
 standardize = TRUE,
 eps = 1e-08,
 maxit = 1e+06,
 istrong = TRUE
)
```

# Arguments

X	A numeric matrix with $N$ rows and $p$ columns representing predictors. Each row corresponds to an observation, and each column corresponds to a variable.
У	A numeric vector of length $N$ representing binary responses. Elements must be either -1 or 1.
nlambda	Number of lambda values in the sequence. Default is 100.
lambda.factor	Ratio of the smallest to the largest lambda in the sequence: lambda.factor = min(lambda) / max(lambda). The default value is $0.0001$ if $N >= p$ or $0.01$ if $N < p$ . Takes no effect if a lambda sequence is specified.
lambda	An optional user-specified sequence of lambda values. If lambda = NULL (default), the sequence is computed based on nlambda and lambda factor. The program automatically sorts user-defined lambda sequences in decreasing order.
lam2	Users may tune $\lambda_2$ , which controls the L2 regularization strength. Default is 0 (lasso).
kern	Type of kernel method for smoothing. Options are "gaussian", "uniform", and "epanechnikov". Default is "epanechnikov".

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hval	The bandwidth parameter for kernel smoothing. Default is 1.
pf	A numeric vector of length $p$ representing the L1 penalty weights for each coefficient. A common choice is $(\beta+1/n)^{-1}$ , where $n$ is the sample size and $\beta$ is obtained from L1 DCSVM or enet DCSVM. Default is 1 for all predictors.
pf2	A numeric vector of length $p$ representing the L2 penalty weights for each coefficient. A value of 0 indicates no L2 shrinkage. Default is 1 for all predictors.
exclude	Indices of predictors to exclude from the model. Equivalent to assigning an infinite penalty factor. Default is none.
dfmax	Maximum number of nonzero coefficients allowed in the model. Default is $p+1$ . Useful for large $p$ when a partial path is acceptable.
pmax	Maximum number of variables allowed to ever be nonzero during the computation. Default is $min(dfmax * 1.2, p)$ .
standardize	Logical indicating whether predictors should be standardized to unit variance. Default is TRUE. Note that predictors are always centered.
eps	Convergence threshold. The algorithm stops when $4\max_j(\beta_j^{new}-\beta_j^{old})^2$ is less than eps. Default is 1e-8.
maxit	Maximum number of iterations allowed. Default is 1e6. Consider increasing maxit if the algorithm does not converge.
istrong	Logical indicating whether to use the strong rule for faster computation. Default

#### Value

An object of class dcsvm containing the following components:

b0 Intercept values for each lambda.

is TRUE.

beta Sparse matrix of coefficients for each lambda. Use as.matrix() to convert.

df Number of nonzero coefficients for each lambda.

dim Dimensions of the coefficient matrix.

lambda Sequence of lambda values used.

npasses Total number of iterations across all lambda values.

jerr Warnings and errors. 0 if no errors.

call The matched call.

#### See Also

```
\verb"print.dcsvm", \verb"predict.dcsvm", \verb"coef.dcsvm", \verb"plot.dcsvm", \verb"and cv.dcsvm".
```

```
# Load the data
data(colon)
# Fit the elastic-net penalized DCSVM with lambda2 to be 1
fit <- dcsvm(colon$x, colon$y, lam2 = 1)
print(fit)</pre>
```

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```
# Coefficients at some lambda value c1 <- coef(fit, s = 0.005) # Make predictions predict(fit, newx = colonx[1:10, ], s = c(0.01, 0.005))
```

plot.cv.dcsvm

Plot the Cross-Validation Curve of Sparse Density-Convoluted SVM

### **Description**

Depicts the cross-validation curves for the sparse density-convoluted SVM.

## Usage

```
## S3 method for class 'cv.dcsvm'
plot(x, sign.lambda, ...)
```

# Arguments

```
x A fitted cv.dcsvm object.

sign.lambda Specifies whether to plot against log(lambda) (default) or its negative if sign.lambda = -1.

Other graphical parameters to plot.
```

#### Details

Plot the Cross-Validation Curve of Sparse Density-Convoluted SVM

Plots the cross-validation curve against a function of lambda values, including upper and lower standard deviation curves.

This function visualizes the cross-validation curves for a cv.dcsvm object, which plots the relationship between lambda values and cross-validation error.

### Value

No return value, only called for plots.

#### See Also

```
cv.dcsvm.
```

```
data(colon)
colon$x <- colon$x[ ,1:100] # Use only the first 100 columns for this example
set.seed(1)
cv <- cv.dcsvm(colon$x, colon$y, lam2=1, nfolds=5)
plot(cv)</pre>
```

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plot.dcsvm	Plot Coefficients for Sparse Density-Convoluted SVM

# Description

Plots the solution paths as a coefficient profile plot for a fitted dcsvm model.

# Usage

```
## S3 method for class 'dcsvm'
plot(x, xvar = c("norm", "lambda"), color = FALSE, label = FALSE, ...)
```

# Arguments

X	A fitted dcsvm model.
xvar	Specifies the X-axis. If xvar == "norm", plots against the L1-norm of the coefficients; if xvar == "lambda", plots against the log-lambda sequence.
color	If TRUE, plots the curves with rainbow colors; otherwise, with gray colors (default).
label	If TRUE, labels the curves with variable sequence numbers. Default is FALSE.
	Other graphical parameters to plot.

#### **Details**

Plot Coefficients for Sparse Density-Convoluted SVM

Plots the solution paths for a fitted dcsvm object.

This function generates a coefficient profile plot showing the solution paths of the sparse density-convoluted SVM.

#### Value

No return value, only called for plots.

## See Also

```
print.dcsvm, predict.dcsvm, coef.dcsvm, plot.dcsvm, and cv.dcsvm.
```

```
data(colon)
fit <- dcsvm(colon$x, colon$y)
oldpar <- par(mfrow = c(1,3)) #changes par() and stores original par()
# Plots against the L1-norm of the coefficients
plot(fit)
# Plots against the log-lambda sequence
plot(fit, xvar="lambda", label=TRUE)
# Plots with colors</pre>
```

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```
plot(fit, color=TRUE)
# Reset to user's option
par(oldpar)
```

predict.cv.dcsvm

Make Predictions from a "cv.dcsvm" Object

#### **Description**

Predicts class labels for new data based on the cross-validated lambda values from a cv.dcsvm object.

# Usage

```
## S3 method for class 'cv.dcsvm'
predict(object, newx, s = c("lambda.1se", "lambda.min"), ...)
```

# Arguments

object	A fitted cv.dcsvm object.
newx	A matrix of new values for x at which predictions are to be made. Must be a matrix. See documentation for predict.dcsvm.
S	Value(s) of the L1 tuning parameter lambda for making predictions. Default is s = "lambda.1se" saved in the cv.dcsvm object. An alternative choice is s = "lambda.min". s can also be numeric, representing the specific value(s) to use.
	Not used. Other arguments to predict.

## Details

Make Predictions from a "cv.dcsvm" Object

This function predicts the class labels of new observations using the sparse density-convoluted SVM at the lambda values suggested by cv.dcsvm.

This function uses the cross-validation results to make predictions. It is adapted from the predict.cv function in the glmnet and gcdnet packages.

#### Value

Predicted class labels or fitted values, depending on the choice of s and any arguments passed to the dcsvm method.

#### See Also

```
cv.dcsvm, and coef.cv.dcsvm methods.
```

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## **Examples**

```
data(colon)
colon$x <- colon$x[ , 1:100] # Use only the first 100 columns for this example
set.seed(1)
cv <- cv.dcsvm(colon$x, colon$y, lam2=1, nfolds=5)
predict(cv$dcsvm.fit, newx=colon$x[2:5, ],
    s=cv$lambda.1se, type="class")</pre>
```

predict.dcsvm

Make Predictions for Sparse Density-Convoluted SVM

## **Description**

Predicts binary class labels or fitted values for a dcsvm model using new data.

#### **Usage**

```
## S3 method for class 'dcsvm'
predict(object, newx, s = NULL, type = c("class", "link"), ...)
```

## **Arguments**

object	A fitted dcsvm object.
newx	A matrix of new values for x at which predictions are to be made. Note that newx must be a matrix; vectors or other formats are not accepted.
S	Value(s) of the L1 tuning parameter lambda for computing coefficients. Default is the entire lambda sequence obtained by dcsvm.
type	"class" or "link"? "class" produces the predicted binary class labels, while "link" returns the fitted values. Default is "class".
	Not used. Other arguments to predict.

#### **Details**

Make Predictions for Sparse Density-Convoluted SVM

This function predicts the binary class labels or the fitted values of a dcsvm object.

s represents the new lambda values for making predictions. If s is not part of the original lambda sequence generated by dcsvm, predict.dcsvm uses linear interpolation to compute predictions by combining adjacent lambda values in the original sequence. This functionality is adapted from the predict methods in the glmnet and gcdnet packages.

### Value

Returns either the predicted class labels or the fitted values, depending on the choice of type.

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## See Also

```
coef.dcsvm
```

## **Examples**

```
data(colon)
fit <- dcsvm(colon$x, colon$y, lam2=1)
print(predict(fit, type="class", newx=colon$x[2:5, ]))</pre>
```

print.dcsvm

Print a DCSVM Object

## **Description**

Prints a summary of the dcsvm object, showing the solution paths.

### Usage

```
## S3 method for class 'dcsvm'
print(x, digits = max(3, getOption("digits") - 3), ...)
```

# **Arguments**

```
    A fitted dcsvm object.
    digits Specifies the significant digits to use in the output. Default is max(3, getOption("digits") - 3).
    Additional arguments to print.
```

#### **Details**

Print a DCSVM Object

Print a summary of the dcsvm solution paths.

This function prints a two-column matrix with columns Df and Lambda. The Df column shows the number of nonzero coefficients, and the Lambda column displays the corresponding lambda value. It is adapted from the print function in the gcdnet and glmnet packages.

## Value

A two-column matrix with one column showing the number of nonzero coefficients and the other column showing the lambda values.

### See Also

```
print.dcsvm, predict.dcsvm, coef.dcsvm, plot.dcsvm, and cv.dcsvm.
```

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```
data(colon)
fit <- dcsvm(colon$x, colon$y)
print(fit)</pre>
```

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