

# Package: svmpath (via r-universe)

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**Title** The SVM Path Algorithm

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**Description** Computes the entire regularization path for the two-class svm classifier with essentially the same cost as a single SVM fit.

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**Depends** kernlab

**License** GPL-2

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## Contents

balanced.overlap . . . . .	2
plot.svmpath . . . . .	2
predict.svmpath . . . . .	4
print.svmpath . . . . .	5
radial.kernel . . . . .	6
summary.svmpath . . . . .	7
svmpath . . . . .	8

<b>Index</b>	<b>11</b>
--------------	-----------

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balanced.overlap      *simple examples for svmpath*

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### Description

Datasets for illustrating the svmpath function, that can be plotted while its running

### Usage

```
data(svmpath)
```

### Format

In each case a list with a component `x` (t column matrix) and a component `y` (vector of +1/-1 values) "Balanced" refers to whether the number of +1s is the same as the -1s. "Overlap" indicates whether the classes are linearly separable. `mixture.data` is a balanced dataset with 100 observations in each class. The others are smaller with between 10-12 obs total.

### References

The paper <http://www-stat.stanford.edu/~hastie/Papers/svmpath.pdf>, as well as the talk <http://www-stat.stanford.edu/~hastie/TALKS/svmpathtalk.pdf>.

### Examples

```
data(svmpath)
attach(balanced.overlap)
svmpath(x,y,trace=TRUE,plot=TRUE)
detach(2)
```

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plot.svmpath      *plot the svm solution at a step along the path*

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### Description

produces a plot of the svm solution along the path, and optionally indicates support points

### Usage

```
## S3 method for class 'svmpath'
plot(x, step, Size = 60, elbow.show = TRUE, support.show = TRUE, ...)
```

**Arguments**

<code>x</code>	the <code>svmpath</code> object
<code>step</code>	which step to plot; default is the last step. Use <code>summary</code> to see how many steps
<code>Size</code>	If the solution is non-linear, this is the gridsize for <code>countour</code>
<code>elbow.show</code>	Should the points on the elbow be indicated
<code>support.show</code>	Should the support points be indicated
<code>...</code>	additional arguments to plot, allowing one to change, for example, "main", "xlab" etc

**Details**

A two-dimensional plot is produced of the SVM solution. Makes sense only if `X` is two-dimensional. If not, the first two dimensions will be used

**Value**

A list is returned silently, with the ingredients of the plot

**Author(s)**

Trevor Hastie

**References**

The paper <http://www-stat.stanford.edu/~hastie/Papers/svmpath.pdf>, as well as the talk <http://www-stat.stanford.edu/~hastie/TALKS/svmpathtalk.pdf>.

**See Also**

`coef.svmpath`, `svmpath`, `predict.svmpath`, `print.svmpath`, `summary.svmpath`

**Examples**

```
data(svmpath)
attach(balanced.overlap)
fit <- svmpath(x,y,trace=TRUE,plot=FALSE)
plot(fit,step=2)
detach(2)
```

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predict.svmpath      *Make predictions from a "svmpath" object*

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### Description

Provide a value for lambda, and produce the fitted lagrange alpha values. Provide values for x, and get fitted function values or class labels.

### Usage

```
## S3 method for class 'svmpath'
predict(object, newx, lambda, type = c("function", "class",
"alpha", "margin"),...)
```

### Arguments

object	fitted svmpath object
newx	values of x at which prediction are wanted. This is a matrix with observations per row
lambda	the value of the regularization parameter. Note that lambda is equivalent to 1/C for the usual parametrization of a SVM
type	type of prediction, with default "function". For type="alpha" or type="margin" the newx argument is not required
...	Generic compatibility

### Details

This implementation of the SVM uses a parameterization that is slightly different but equivalent to the usual (Vapnik) SVM. Here  $\lambda = 1/C$ . The Lagrange multipliers are related via  $\alpha_i^* = \alpha_i/\lambda$ , where  $\alpha_i^*$  is the usual multiplier, and  $\alpha_i$  our multiplier. Note that if alpha=0, that observation is right of the elbow; alpha=1, left of the elbow;  $0 < \alpha < 1$  on the elbow. The latter two cases are all support points.

### Value

In each case, the desired prediction.

### Author(s)

Trevor Hastie

### References

The paper <http://www-stat.stanford.edu/~hastie/Papers/svmpath.pdf>, as well as the talk <http://www-stat.stanford.edu/~hastie/TALKS/svmpathtalk.pdf>.

**See Also**

coef.svmpath, svmpath

**Examples**

```
data(svmpath)
attach(balanced.overlap)
fit <- svmpath(x,y,trace=TRUE,plot=TRUE)
predict(fit, lambda=1,type="alpha")
predict(fit, x, lambda=.9)
detach(2)
```

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print.svmpath	<i>Print a summary of the SVM path</i>
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**Description**

print a summary of the fitted svmpath object

**Usage**

```
## S3 method for class 'svmpath'
print(x, digits, maxsteps, ...)
```

**Arguments**

x	object to be printed
digits	number of significant digits (default 6)
maxsteps	the number of steps to print; default all
...	additional arguments to the generic print function

**Value**

For each step taken by the algorithm, one or more lines are printed. The step is described in terms of the observation number involved, a coded version of what happened, such as "L->E" meaning "from the Left set" to the "Elbow". Initially all the sets are empty. It gives the margin (sum of the  $\xi_i$ ), the size of the elbow, and the training error.

**Author(s)**

Trevor Hastie

**References**

The paper <http://www-stat.stanford.edu/~hastie/Papers/svmpath.pdf>, as well as the talk <http://www-stat.stanford.edu/~hastie/TALKS/svmpathtalk.pdf>.

**See Also**

coef.svmpath, svmpath, predict.svmpath

**Examples**

```
data(svmpath)
attach(balanced.overlap)
fit <- svmpath(x,y,trace=TRUE,plot=TRUE)
print(fit)
detach(2)
```

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radial.kernel	<i>compute the kernel matrix for svmpath</i>
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**Description**

compute the kernel matrix for svmpath

**Usage**

```
radial.kernel(x, y=x, param.kernel = 1/p, ...)
poly.kernel(x, y=x, param.kernel = 1, ...)
```

**Arguments**

x	an n x p matrix of features
y	an m x p matrix of features (if omitted, it defaults to x)
param.kernel	the parameter(s) for the kernel. For this radial kernel, the parameter is known in the fields as "gamma". For the polynomial kernel, it is the "degree"
...	unused

**Details**

For the radial kernel, this computes the function  $\exp(-\gamma\|x - y\|^2)$  for each pair of rows x,y from the input matrices. Here g is param.kernel. For the polynomial kernel, it computes  $(xy^T + 1)^d$ , where d is param.kernel.

**Value**

An n x m matrix.

**Author(s)**

Trevor Hastie

## References

The paper <http://www-stat.stanford.edu/~hastie/Papers/svmpath.pdf>, as well as the talk <http://www-stat.stanford.edu/~hastie/TALKS/svmpathtalk.pdf>.

## See Also

svmpath

## Examples

```
data(svmpath)
attach(balanced.overlap)
fit<-svmpath(x,y,kernel=radial.kernel)
detach(2)
```

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summary.svmpath	<i>produce a summary of an svmpath object</i>
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## Description

printing an svmpath object can produce a lot of lines. The summary methods gives a more concise description by picking out a subset of the steps

## Usage

```
## S3 method for class 'svmpath'
summary(object, nsteps = 5, digits = 6, ...)
```

## Arguments

object	the svmpath object
nsteps	usually omitted, but can be changed to get longer summaries
digits	number of significant digits
...	additional arguments to the generic summary function

## Details

Uses the pretty function to extract the approximately the desired number of steps. Always includes the first and last step.

## Value

returns a dataframe with the steps, value of lambda, training error, size of elbow, number of support points, and the sum of the overlaps

## Author(s)

Trevor Hastie

## References

The paper <http://www-stat.stanford.edu/~hastie/Papers/svmpath.pdf>, as well as the talk <http://www-stat.stanford.edu/~hastie/TALKS/svmpathtalk.pdf>.

## See Also

coef.svmpath, svmpath, predict.svmpath, print.svmpath

## Examples

```
data(svmpath)
attach(balanced.overlap)
fit <- svmpath(x,y,trace=TRUE,plot=TRUE)
summary(fit)
detach(2)
```

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svmpath

*Fit the entire regularization path for a 2-class SVM*

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## Description

The SVM has a regularization or cost parameter  $C$ , which controls the amount by which points overlap their soft margins. Typically either a default large value for  $C$  is chosen (allowing minimal overlap), or else a few values are compared using a validation set. This algorithm computes the entire regularization path (i.e. for all possible values of  $C$  for which the solution changes), with a cost a small ( $\sim 3$ ) multiple of the cost of fitting a single model.

## Usage

```
svmpath(x, y, K, kernel.function = poly.kernel, param.kernel = 1, trace,
        plot.it, eps = 1e-10, Nmoves = 3 * n, digits = 6, lambda.min = 1e-04, ridge=0, ...)
```

## Arguments

<code>x</code>	the data matrix ( $n \times p$ ) with $n$ rows (observations) on $p$ variables (columns)
<code>y</code>	The "-1,+1" valued response variable.
<code>K</code>	a $n \times n$ kernel matrix, with default value $K = \text{kernel.function}(x, x)$
<code>kernel.function</code>	This is a user-defined function. Provided are <code>poly.kernel</code> (the default, with parameter set to default to a linear kernel) and <code>radial.kernel</code>
<code>param.kernel</code>	parameter(s) of the kernels
<code>trace</code>	if TRUE, a progress report is printed as the algorithm runs; default is FALSE
<code>plot.it</code>	a flag indicating whether a plot should be produced (default FALSE; only usable with $p=2$ )
<code>eps</code>	a small machine number which is used to identify minimal step sizes

<code>Nmoves</code>	the maximum number of moves
<code>digits</code>	the number of digits in the printout
<code>lambda.min</code>	The smallest value of $\lambda = 1/C$ ; default is $\lambda=10e-4$ , or $C=10000$
<code>ridge</code>	Sometimes the algorithm encounters singularities; in this case a small value of <code>ridge</code> , around $1e-12$ , can help. Default is <code>ridge=0</code>
<code>...</code>	additional arguments to some of the functions called by <code>svmpath</code> . One such argument that can be passed is <code>ridge</code> (default is $1e-10$ ). This is used to produce "stable" solutions to linear equations.

### Details

The algorithm used in `svmpath()` is described in detail in "The Entire Regularization Path for the Support Vector Machine" by Hastie, Rosset, Tibshirani and Zhu (2004). It exploits the fact that the "hinge" loss-function is piecewise linear, and the penalty term is quadratic. This means that in the dual space, the lagrange multipliers will be piecewise linear (c.f. lars).

### Value

a "svmpath" object is returned, for which there are `print`, `summary`, `coef` and `predict` methods.

### Warning

Currently the algorithm can get into machine errors if `epsilon` is too small, or if `lambda.min` is too small. Increasing either from their defaults should make the problems go away, by terminating the algorithm slightly early.

### Note

This implementation of the algorithm does not use updating to solve the "elbow" linear equations. This is possible, since the elbow changes by a small number of points at a time. Future version of the software will do this. The author has encountered numerical problems with early attempts at this.

### Author(s)

Trevor Hastie

### References

The paper <http://www-stat.stanford.edu/~hastie/Papers/svmpath.pdf>, as well as the talk <http://www-stat.stanford.edu/~hastie/TALKS/svmpathtalk.pdf>.

### See Also

`print`, `coef`, `summary`, `predict`, and `FilmPath`

**Examples**

```
data(svmpath)
attach(unbalanced.separated)
svmpath(x,y,trace=TRUE,plot=TRUE)
detach(2)
## Not run: svmpath(x,y,kernel=radial.kernel,param.kernel=.8)
```

# Index

## \* datasets

balanced.overlap, 2

## \* regression

plot.svm<sub>path</sub>, 2

predict.svm<sub>path</sub>, 4

print.svm<sub>path</sub>, 5

radial.kernel, 6

summary.svm<sub>path</sub>, 7

svm<sub>path</sub>, 8

balanced.overlap, 2

balanced.separated (balanced.overlap), 2

mixture.data (balanced.overlap), 2

plot.svm<sub>path</sub>, 2

poly.kernel (radial.kernel), 6

predict.svm<sub>path</sub>, 4

print.svm<sub>path</sub>, 5

radial.kernel, 6

summary.svm<sub>path</sub>, 7

svm<sub>path</sub>, 8

unbalanced.separated

(balanced.overlap), 2