

# Quick start guide for the **ncvreg** package

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This guide is intended to briefly demonstrate the basic usage of **ncvreg**. For more details, see the documentation for individual functions, as well as the references.

**ncvreg** comes with a few example data sets; we'll look at **prostate**, which has 8 features and one continuous response, **prostate\$lp<sub>psa</sub>**, the PSA levels (on the log scale) from men about to undergo radical prostatectomy. The data is available as a data frame; we will turn it into a design matrix **X** and response vector **y** for the purpose of analysis

```
> ## Linear regression
> data(prostate)
> X <- as.matrix(prostate[,1:8])
> y <- prostate$lppsa
> head(X)
```

	lcavol	lweight	age	lbph	svi	lcp	gleason	p <sub>gg45</sub>
1	-0.5798185	2.769459	50	-1.386294	0	-1.386294	6	0
2	-0.9942523	3.319626	58	-1.386294	0	-1.386294	6	0
3	-0.5108256	2.691243	74	-1.386294	0	-1.386294	7	20
4	-1.2039728	3.282789	58	-1.386294	0	-1.386294	6	0
5	0.7514161	3.432373	62	-1.386294	0	-1.386294	6	0
6	-1.0498221	3.228826	50	-1.386294	0	-1.386294	6	0

```
> head(y)
```

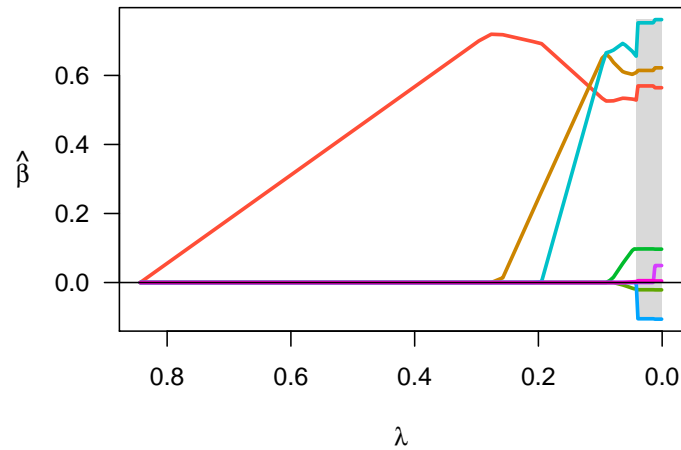
[1]	-0.4307829	-0.1625189	-0.1625189	-0.1625189	0.3715636	0.7654678
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To fit a penalized regression model to this data:

```
> fit <- ncvreg(X, y)
```

The default penalty here is the minimax concave penalty (MCP), but SCAD and lasso penalties are also available. This produces a path of coefficients, which we can plot with

```
> plot(fit)
```



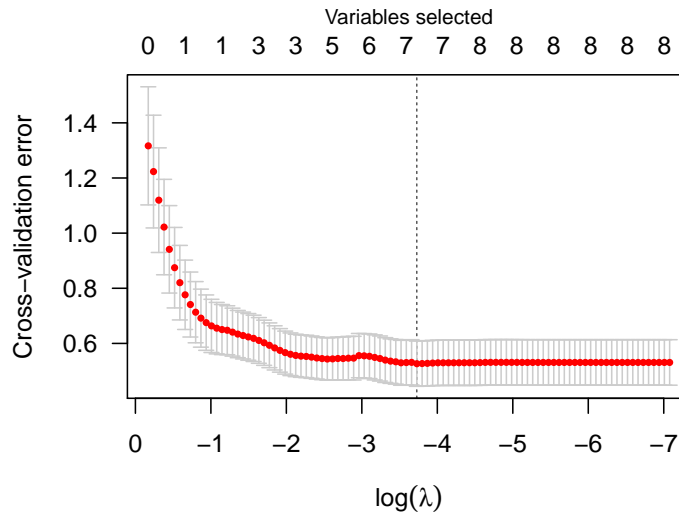
Notice that variables enter the model one at a time, and that at any given value of  $\lambda$ , several coefficients are zero. To see what the coefficients are, we could use the `coef` function:

```
> coef(fit, lambda=0.1)
```

(Intercept)	lcavol	lweight	age	lbph	svi
-0.6973059	0.5387509	0.6382717	0.0000000	0.0000000	0.6102800
lcp	gleason	pgg45			
0.0000000	0.0000000	0.0000000			

Typically, one would carry out cross-validation for the purposes of assessing the predictive accuracy of the model at various values of  $\lambda$ :

```
> cvfit <- cv.ncvreg(X, y)
> plot(cvfit)
```



The coefficients corresponding to the value of  $\lambda$  that minimizes the cross-validation error can be obtained via `coef`:

```
> coef(cvfit)

(Intercept)      lcavol      lweight      age      lbph
0.494155488  0.569546029  0.614419592 -0.020913469  0.097352619
      svi      lcp      gleason      pgg45
0.752398445 -0.104959575  0.000000000  0.005324463
```

Predicted values can be obtained via `predict`, which has a number of options:

```
> predict(cvfit, X=head(X))

      1      2      3      4      5      6
0.8304041 0.7650906 0.4262073 0.6230117 1.7449492 0.8449595

> predict(cvfit, type="nvars")

0.02402
7
```

Note that the original fit (to the full data set) is returned as `cvfit$fit`; it is not necessary to call both `ncvreg` and `cv.ncvreg` to analyze a data set. Methods for logistic regression and Cox proportional hazards regression are also available.