## 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.

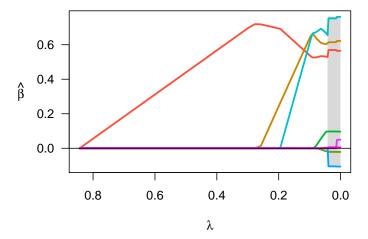
ncvreg comes with a few example data sets; we'll look at prostate, which has 8 features and one continuous response, prostate\$1psa, the PSA levels (on the log scale) from men about to undergo radical prostatectomy. The data is available as a data frame; we will turn it into a design matrix X and response vector y for the purpose of analysis

```
> ## Linear regression
> data(prostate)
> X <- as.matrix(prostate[,1:8])</pre>
> y <- prostate$lpsa</pre>
> head(X)
      lcavol lweight age
                               lbph svi
                                              1cp gleason pgg45
1 -0.5798185 2.769459 50 -1.386294
                                     0 -1.386294
2 -0.9942523 3.319626 58 -1.386294
                                    0 -1.386294
                                                               0
                                                        7
3 -0.5108256 2.691243 74 -1.386294
                                    0 -1.386294
                                                              20
4 -1.2039728 3.282789 58 -1.386294
                                    0 -1.386294
                                                              0
                                                        6
5 0.7514161 3.432373 62 -1.386294
                                      0 -1.386294
                                                               0
6 -1.0498221 3.228826 50 -1.386294
                                      0 -1.386294
> head(y)
[1] -0.4307829 -0.1625189 -0.1625189 -0.1625189 0.3715636 0.7654678
```

To fit a penalized regression model to this data:

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

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

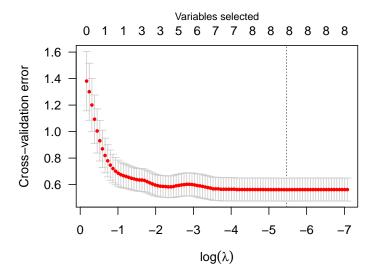


Notice that variables enter the model one at a time, and that at any given value of  $\lambda$ , several coefficients are zero. To see what the coefficients are, we could use the coeff function:

```
> coef(fit, lambda=0.1)
(Intercept)
                 lcavol
                             lweight
                                                          lbph
                                                                       svi
                                              age
 -0.6973059
              0.5387509
                           0.6382717
                                        0.000000
                                                    0.0000000
                                                                 0.6102800
        lcp
                 gleason
                               pgg45
  0.0000000
              0.0000000
                           0.000000
```

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:

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

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 logstic regression and Cox proportional hazards regression are also available.