Quick start guide for the ncvreg package

Patrick Breheny

April 17, 2018

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

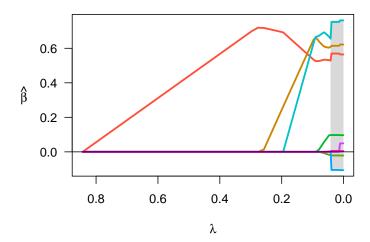
```
> # Linear regression
> data(Prostate)
> head(Prostate$X)
     lcavol lweight age
                              lbph svi
                                             1cp gleason pgg45
1 -0.5798185 2.769459 50 -1.386294
                                    0 -1.386294
2 -0.9942523 3.319626 58 -1.386294
                                    0 -1.386294
                                                       6
                                                             0
3 -0.5108256 2.691243
                      74 -1.386294
                                     0 -1.386294
                                                            20
4 -1.2039728 3.282789 58 -1.386294
                                                             0
                                    0 -1.386294
5 0.7514161 3.432373 62 -1.386294
                                                             0
                                    0 -1.386294
6 -1.0498221 3.228826
                      50 -1.386294
                                     0 -1.386294
                                                             0
> head(Prostate$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(Prostate$X, Prostate$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

```
> plot(fit)
```

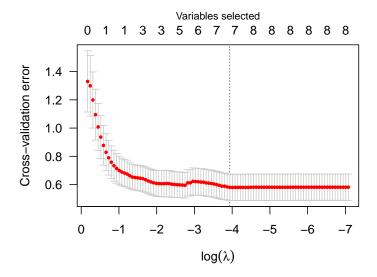


Notice that variables enter the model one at a time, and that at any given value of λ , several coefficients are zero. To see what the coefficients are, we could use the coef function:

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

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



The coefficients corresponding to the value of λ that minimizes the cross-validation error can be obtained via coef:

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

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

1          2     3     4     5     6
0.8304040 0.7650906 0.4262072 0.6230118 1.7449492 0.8449595

> predict(cvfit, type="nvars")
0.01948
          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 logstic regression and Cox proportional hazards regression are also available.