

short 10-15 min

Stat Learning

e 1) The table below shows simulation results for bootstrapping OLS (reg) and lasso and ridge regression (RR) with 10-fold CV when $\beta = (1, 1, 0, 0)^T$. The β_i columns give coverage = the proportion of CIs that contained β_i and the average length of the CI. The test is for $H_0 : (\beta_3, \beta_4)^T = \mathbf{0}$ and H_0 is true. The "coverage" is the proportion of times the prediction region method bootstrap test failed to reject H_0 . OLS used 1000 runs while 100 runs were used for lasso and ridge regression. Since 100 runs were used, a cov in $[0.89, 1]$ is reasonable for a nominal value of 0.95. If the coverage for both methods ≥ 0.89 , the method with the shorter average CI length was more precise. (If one method had coverage ≥ 0.89 and the other had coverage < 0.89 , we will say the method with coverage ≥ 0.89 was more precise.) The results for the lasso test were omitted since sometimes S_T^* was singular. (Lengths for the test column are not comparable unless the statistics have the same asymptotic distribution.)

3.5

table 3.5

and RR

Table 1: Bootstrapping lasso, $n = 100, \psi = 0.9, p = 4, B = 250$

| | | β_1 | β_2 | β_3 | β_4 | test |
|-------|-----|-----------|-----------|-----------|------------------|-------|
| reg | cov | 0.942 | 0.951 | 0.949 | 0.943 | 0.943 |
| | len | 0.658 | 5.447 | 5.444 | 5.438 | 2.490 |
| RR | cov | 0.97 | 0.02 | 0.11 | 0.992 | 0.05 |
| | len | 0.681 | 0.329 | 0.334 | 0.334 | 2.546 |
| reg | cov | 0.947 | 0.955 | 0.950 | 0.951 | 0.952 |
| | len | 0.658 | 5.511 | 5.497 | 5.500 | 2.491 |
| lasso | cov | 0.93 | 0.91 | 0.92 | 0.99 | |
| | len | 0.698 | 3.765 | 3.922 | 3.803 | |

0.10

a) For β_3 and β_4 which method, ridge regression or the OLS full model, was better?

OLS

(RR cov is too low)

b) For β_3 and β_4 which method, lasso or the OLS full model, was more precise?

lasso

(3.8 < 5.5)

3.13 e 2) Suppose the MLR model $\mathbf{Y} = \mathbf{X}\beta + e$, and the regression method fits $\mathbf{Z} = \mathbf{W}\eta + e$. Suppose $\hat{Z} = 245.63$ and $\bar{Y} = 105.37$. What is \hat{Y} ?

10679

$$\hat{Y} = \hat{Z} + \bar{Y} = \boxed{351.00}$$

3.14 2) 3) To get a large sample 90% PI for a future value Y_f of the response variable, find a large sample 90% PI for a future residual and add \hat{Y}_f to the endpoints of the of that PI. Suppose forward selection is used and the large sample 90% PI for a future residual is $[-778.28, 1336.44]$. What is the large sample 90% PI for Y_f if $\hat{\beta}_{I_{min}} = (241.545, 1.001)^T$ used a constant and the predictor *mmen* with corresponding $\mathbf{x}_{I_{min},f} = (1, 75000)^T$?

more data

$$\hat{Y}_f = 241.545 + 1.001(75000) = \underline{75316.545}$$

So 95% PI = $[-778.28 + \hat{Y}_f, 1336.44 + \hat{Y}_f]$

$$= \boxed{[74538.265, 76652.985]}$$

106
710

25