

Ridge Regression: Prostate Cancer Data

Data

```
rm(list=ls())
# training and test data
require(ElemStatLearn)
train <- prostate[prostate$train,-10]
test <- prostate[!prostate$train,-10]
# n.ro of observations
n <- nrow(train)
m <- nrow(test)

# design matrix (without the intercept)
X<-as.matrix(
  scale(train[, -9]) # standardized
)
X.star <- as.matrix(
  scale(test[, -9]) # standardized
)
p = ncol(X)
# response
y = scale(train[, "lpsa"]) # standardized
y.star = scale(test[, "lpsa"]) # standardized
```

Exercise

The training and test data are

$$\underset{n \times 1}{\mathbf{y}}, \underset{n \times p}{\mathbf{X}}, \underset{m \times 1}{\mathbf{y}^*}, \underset{m \times p}{\mathbf{X}^*}$$

with $n = 67$, $m = 30$ and $p = 8$ for the Prostate Cancer data set, where the response and each predictor of the design matrix are standardized to have mean 0 and variance 1, separately for the training and test set.

The response variable is `lpsa`, and the predictor variables are `lcavol`, `lweight`, `age`, `lbph`, `svi`, `lcp`, `gleason`, `pgg45`.

1. Compute the ridge estimate

$$\hat{\beta}(\lambda) = (\mathbf{X}^T \mathbf{X} + \lambda \mathbf{I}_{p \times p})^{-1} \mathbf{X}^T \mathbf{y}$$

for $\lambda = 10$.

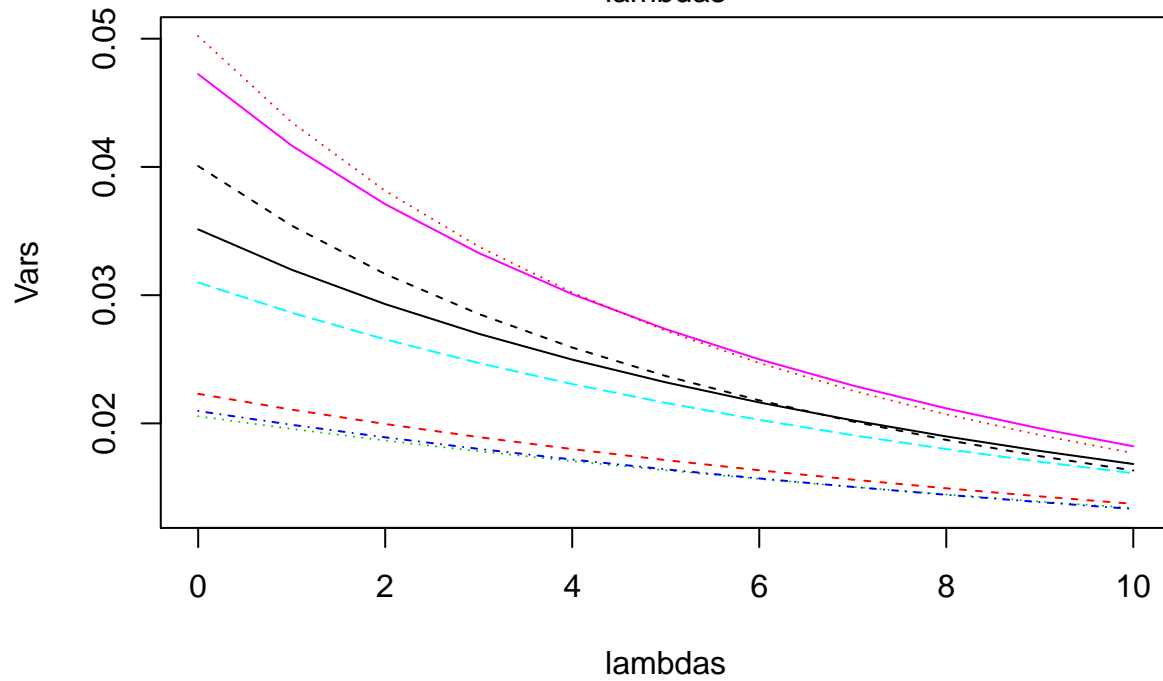
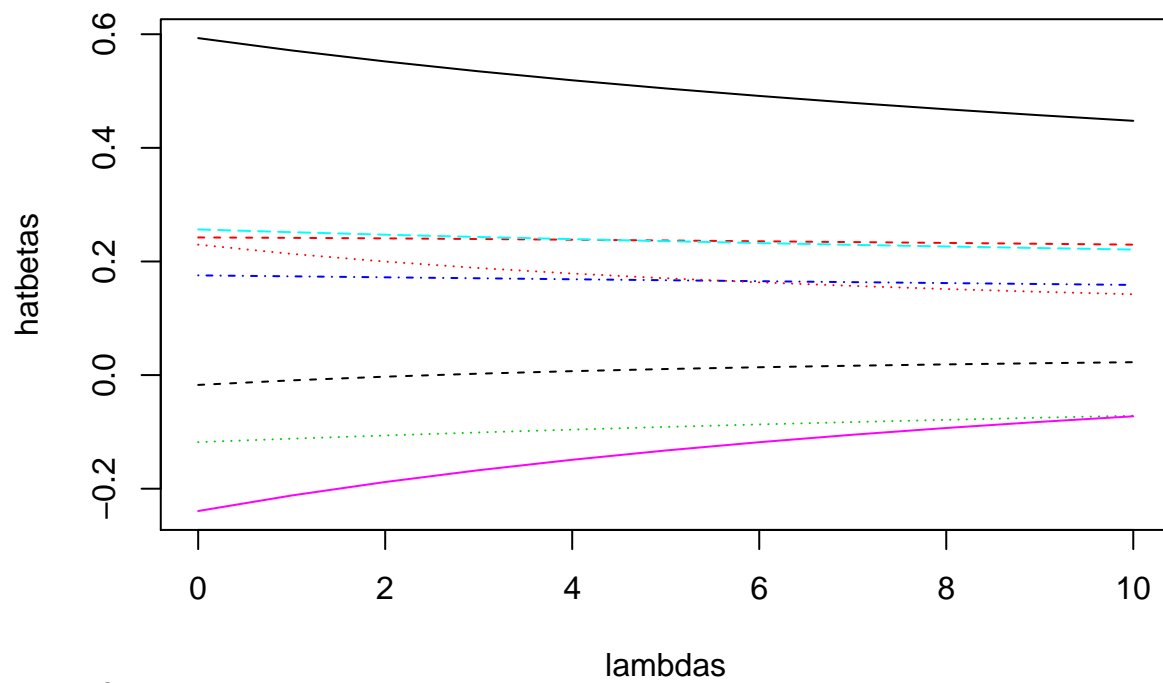
2. For $\lambda = 0, 1, 2, \dots, 9, 10$, plot the solution path of the estimated coefficients $\hat{\beta}_1(\lambda), \dots, \hat{\beta}_p(\lambda)$ and the corresponding variances $\text{Var}(\hat{\beta}_1(\lambda)), \dots, \text{Var}(\hat{\beta}_p(\lambda))$, which are the elements in the diagonal of the variance/covariance matrix

$$\text{Var}(\hat{\beta}(\lambda)) = \sigma^2 \mathbf{W}_\lambda (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{W}_\lambda^T$$

where $\mathbf{W}_\lambda = [\mathbf{I}_{p \times p} + \lambda(\mathbf{Z}^T \mathbf{X})^{-1}]^{-1}$ and is set to $\sigma^2 = 1$.

3. Compare the LOOCV estimate and the test MSE for the linear model and the ridge regression model with $\lambda = 5$.

```
##           [,1]
## lcavol    0.44761782
## lweight   0.22960005
## age       -0.07149012
## lbph      0.15871434
## svi       0.22098408
## lcp       -0.07256130
## gleason   0.02269095
## pgg45     0.14225969
```



```
## L00CV ridge: 0.382395
```

```
## MSE ridge: 0.4565821
## LOOCV lm: 0.3867472
## MSE lm: 0.4747565
```