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In this question, we will predict the number of applications received (Apps) using the other variables in the College data set (ISLR package).

```
library(ISLR)
library(leaps)
names(College)
```

```
## [1] "Private"      "Apps"          "Accept"         "Enroll"         "Top10perc"
## [6] "Top25perc"    "F.Undergrad"   "P.Undergrad"    "Outstate"        "Room.Board"
## [11] "Books"        "Personal"      "PhD"            "Terminal"        "S.F.Ratio"
## [16] "perc.alumni"  "Expend"        "Grad.Rate"
```

```
dim(College)
```

```
## [1] 777 18
```

```
bss_fit <- regsubsets(Apps~., data = College, nvmax=17)
bss_summary = summary(bss_fit)
```

(a) Perform best subset selection to the data. What is the best model obtained according to Cp, BIC and adjusted R2? Show some plots to provide evidence for your answer, and report the coefficients of the best model.

```
bss_summary$rsq
```

```
## [1] 0.8900990 0.9157839 0.9183356 0.9212640 0.9237599 0.9247464 0.9257649
## [8] 0.9268725 0.9276780 0.9283103 0.9288011 0.9289945 0.9291223 0.9291632
## [15] 0.9291878 0.9291885 0.9291887
```

```
par(mfrow = c(2,2))
```

#1

```
plot(bss_summary$cp ,xlab = " Number of Variables ",ylab="Cp", type="l")
points(which.min (bss_summary$cp), bss_summary$cp [which.min
(bss_summary$cp)], col ="red",cex =2, pch =20)
```

#2

```
plot(bss_summary$bic ,xlab=" Number of Variables ",ylab=" BIC", type="l")
points (which.min (bss_summary$bic ), bss_summary$bic [which.min
(bss_summary$bic )], col =" red",cex =2, pch =20)
```

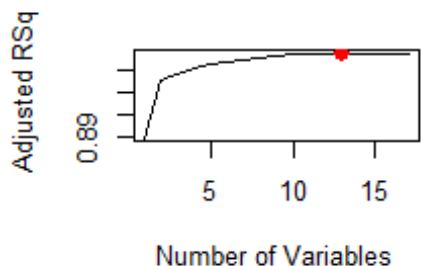
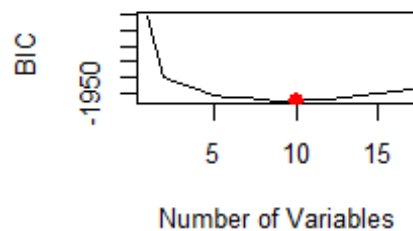
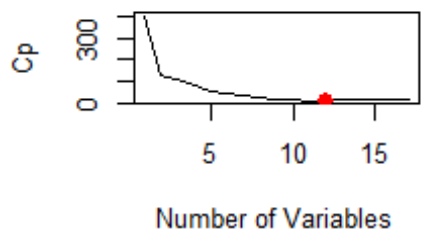
#3

```
plot(bss_summary$adjr2 ,xlab = " Number of Variables ", ylab=" Adjusted
RSq",type="l")
```

```
points (which.max(bss_summary$adjr2),
bss_summary$adjr2[which.max(bss_summary$adjr2)], col ="red",cex =2, pch =20)
```

```
coef(bss_fit , which.min (bss_summary$bic ))
```

```
## (Intercept) PrivateYes Accept Enroll Top10perc
## -100.51668243 -575.07060789 1.58421887 -0.56220848 49.13908916
## Top25perc Outstate Room.Board PhD Expend
## -13.86531103 -0.09466457 0.16373674 -10.01608705 0.07273776
## Grad.Rate
## 7.33268904
```



(b) Repeat (a) using forward stepwise selection and backwards stepwise selection. How does your answer compare to the results in (a)?

#Forward stepwise selection

```
for_ss <- regsubsets(Apps~., data = College, nvmax=17, method = "forward")
for_ss_summary = summary(for_ss)
```

```
par(mfrow =c(2,2))
```

#1

```
plot(for_ss_summary$cp ,xlab =" Number of Variables ",ylab="Cp", type="l")
points(which.min (for_ss_summary$cp), for_ss_summary$cp [which.min
(for_ss_summary$cp)], col ="red",cex =2, pch =20)
```

#2

```

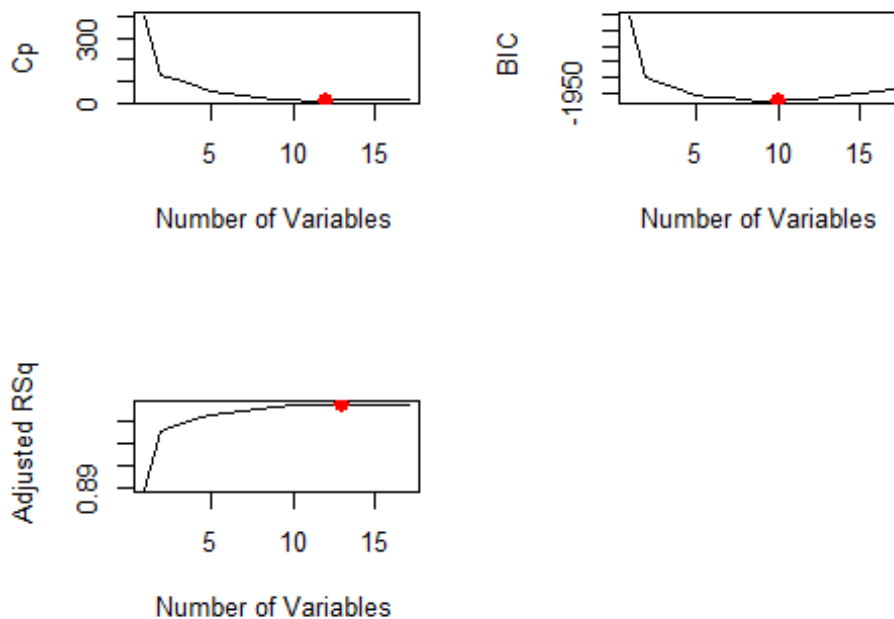
plot(for_ss_summary$bic ,xlab=" Number of Variables ",ylab=" BIC", type="l")
points (which.min (for_ss_summary$bic ), for_ss_summary$bic [which.min
(for_ss_summary$bic )], col =" red",cex =2, pch =20)

#3
plot(for_ss_summary$adjr2 ,xlab =" Number of Variables ", ylab=" Adjusted
RSq",type="l")
points (which.max(for_ss_summary$adjr2),
for_ss_summary$adjr2[which.max(for_ss_summary$adjr2)], col ="red",cex =2, pch
=20)

coef(for_ss , which.min (for_ss_summary$bic ))

## (Intercept) PrivateYes Accept Enroll Top10perc
## -100.51668243 -575.07060789 1.58421887 -0.56220848 49.13908916
## Top25perc Outstate Room.Board PhD Expend
## -13.86531103 -0.09466457 0.16373674 -10.01608705 0.07273776
## Grad.Rate
## 7.33268904

```



#Backward stepwise selection

```

back_ss <- regsubsets(Apps~., data = College, nvmax=17, method = "backward")
back_ss_summary = summary(back_ss)

par(mfrow =c(2,2))

```

#1

```
plot(back_ss_summary$cp ,xlab = " Number of Variables ",ylab="Cp", type="l")
points(which.min (back_ss_summary$cp), back_ss_summary$cp [which.min
(back_ss_summary$cp)], col ="red",cex =2, pch =20)
```

#2

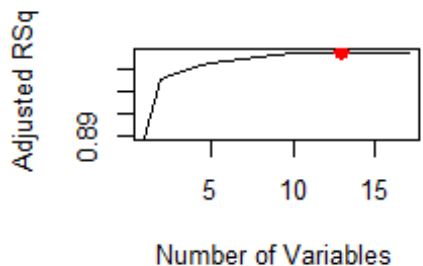
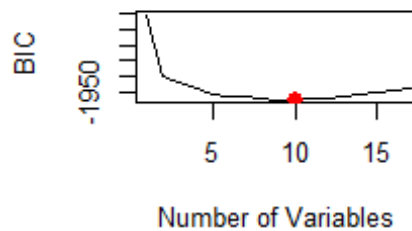
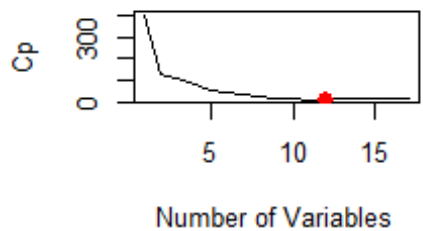
```
plot(back_ss_summary$bic ,xlab=" Number of Variables ",ylab=" BIC", type="l")
points (which.min (back_ss_summary$bic ), back_ss_summary$bic [which.min
(back_ss_summary$bic )], col =" red",cex =2, pch =20)
```

#3

```
plot(back_ss_summary$adjr2 ,xlab = " Number of Variables ", ylab=" Adjusted
RSq",type="l")
points (which.max(back_ss_summary$adjr2),
back_ss_summary$adjr2[which.max(back_ss_summary$adjr2)], col ="red",cex =2,
pch =20)
```

```
coef(back_ss , which.min (back_ss_summary$bic ))
```

```
## (Intercept) PrivateYes Accept Enroll Top10perc
## -100.51668243 -575.07060789 1.58421887 -0.56220848 49.13908916
## Top25perc Outstate Room.Board PhD Expend
## -13.86531103 -0.09466457 0.16373674 -10.01608705 0.07273776
## Grad.Rate
## 7.33268904
```



(c) Fit a lasso model on the data. Use cross-validation to select the optimal value of λ . Create plots of the cross-validation error as a function of λ . Report the resulting coefficient estimates.

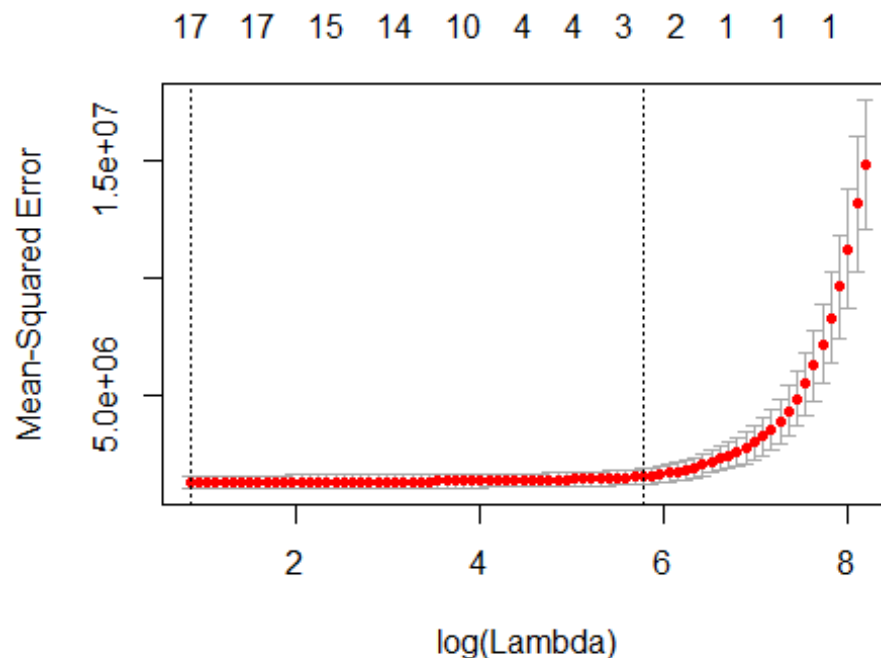
```
library(glmnet)

## Loading required package: Matrix
## Loading required package: foreach
## Loaded glmnet 2.0-16

x = model.matrix(Apps~.,College)[-1]
y = College$Apps
grid = 10^seq(10,-2, length =100)
lasso.mod = glmnet(x,y,alpha =1, lambda =grid)
dim(coef(lasso.mod))

## [1] 18 100

cv.out = cv.glmnet(x,y,alpha =1)
plot(cv.out)
```



```
bestlam =cv.out$lambda.min
bestlam

## [1] 2.3456
```

```
predict(lasso.mod,s=bestlam,type="coefficients")[1:18,]
```

```
## (Intercept) PrivateYes Accept Enroll Top10perc
## -471.91198713 -491.00658329 1.56900019 -0.75108809 47.98058884
## Top25perc F.Undergrad P.Undergrad Outstate Room.Board
## -12.73328855 0.04062555 0.04399959 -0.08298682 0.14925171
## Books Personal PhD Terminal S.F.Ratio
## 0.01455346 0.02878410 -8.37861131 -3.25247875 14.44887626
## perc.alumni Expend Grad.Rate
## -0.05217050 0.07703557 8.25050995
```

(d) Fit a ridge regression model on the data. Use cross-validation to select the optimal value of lambda. Create plots of the cross-validation error as a function of lambda. Report the resulting coefficient estimates.

```
library(glmnet)
x=model.matrix(Apps~.,College)[,-1]
y=College$Apps
grid =10^seq (10,-2, length =100)
ridge.mod = glmnet(x,y,alpha =0, lambda =grid)
coef(ridge.mod)
```

```
## 18 x 100 sparse Matrix of class "dgCMatrix"
```

```
## [[ suppressing 100 column names 's0', 's1', 's2' ... ]]
```

```
##
## (Intercept) 3.001624e+03 3.001619e+03 3.001613e+03 3.001605e+03
## PrivateYes -1.451161e-03 -1.918346e-03 -2.535939e-03 -3.352359e-03
## Accept 5.761598e-07 7.616485e-07 1.006854e-06 1.331001e-06
## Enroll 1.364210e-06 1.803403e-06 2.383992e-06 3.151494e-06
## Top10perc 2.875188e-05 3.800824e-05 5.024462e-05 6.642037e-05
## Top25perc 2.657756e-05 3.513393e-05 4.644496e-05 6.139744e-05
## F.Undergrad 2.513585e-07 3.322809e-07 4.392555e-07 5.806695e-07
## P.Undergrad 3.915797e-07 5.176450e-07 6.842956e-07 9.045976e-07
## Outstate 1.866318e-08 2.467155e-08 3.261432e-08 4.311419e-08
## Room.Board 2.251252e-07 2.976019e-07 3.934120e-07 5.200671e-07
## Books 1.201804e-06 1.588712e-06 2.100182e-06 2.776315e-06
## Personal 3.951405e-07 5.223521e-07 6.905181e-07 9.128232e-07
## PhD 3.581707e-05 4.734801e-05 6.259123e-05 8.274182e-05
## Terminal 3.756761e-05 4.966213e-05 6.565035e-05 8.678579e-05
## S.F.Ratio 3.616432e-05 4.780710e-05 6.319812e-05 8.354411e-05
## perc.alumni -1.089892e-05 -1.440773e-05 -1.904617e-05 -2.517790e-05
## Expend 7.441500e-08 9.837219e-08 1.300422e-07 1.719079e-07
## Grad.Rate 1.278832e-05 1.690539e-05 2.234792e-05 2.954262e-05
##
## (Intercept) 3.001594e+03 3.001580e+03 3.001561e+03 3.001536e+03
## PrivateYes -4.431617e-03 -5.858329e-03 -7.744351e-03 -1.023755e-02
## Accept 1.759503e-06 2.325957e-06 3.074774e-06 4.064662e-06
## Enroll 4.166086e-06 5.507313e-06 7.280330e-06 9.624145e-06
## Top10perc 8.780372e-05 1.160712e-04 1.534389e-04 2.028367e-04
```

## Top25perc	8.116369e-05	1.072934e-04	1.418353e-04	1.874973e-04
## F.Undergrad	7.676100e-07	1.014734e-06	1.341415e-06	1.773268e-06
## P.Undergrad	1.195823e-06	1.580805e-06	2.089726e-06	2.762485e-06
## Outstate	5.699438e-08	7.534314e-08	9.959906e-08	1.316639e-07
## Room.Board	6.874975e-07	9.088300e-07	1.201418e-06	1.588200e-06
## Books	3.670119e-06	4.851673e-06	6.413610e-06	8.478387e-06
## Personal	1.206697e-06	1.595179e-06	2.108726e-06	2.787601e-06
## PhD	1.093796e-04	1.445931e-04	1.911431e-04	2.526790e-04
## Terminal	1.147255e-04	1.516600e-04	2.004851e-04	2.650284e-04
## S.F.Ratio	1.104402e-04	1.459952e-04	1.929967e-04	2.551294e-04
## perc.alumni	-3.328367e-05	-4.399901e-05	-5.816401e-05	-7.688925e-05
## Expend	2.272519e-07	3.004131e-07	3.971276e-07	5.249778e-07
## Grad.Rate	3.905356e-05	5.162644e-05	6.824699e-05	9.021829e-05
##				
## (Intercept)	3.001503e+03	3.001460e+03	3.001402e+03	3.001326e+03
## PrivateYes	-1.353340e-02	-1.789027e-02	-2.364975e-02	-3.126334e-02
## Accept	5.373230e-06	7.103069e-06	9.389796e-06	1.241268e-05
## Enroll	1.272251e-05	1.681834e-05	2.223273e-05	2.939015e-05
## Top10perc	2.681372e-04	3.544599e-04	4.685722e-04	6.194198e-04
## Top25perc	2.478594e-04	3.276538e-04	4.331360e-04	5.725749e-04
## F.Undergrad	2.344147e-06	3.098810e-06	4.096420e-06	5.415182e-06
## P.Undergrad	3.651827e-06	4.827473e-06	6.381587e-06	8.435998e-06
## Outstate	1.740515e-07	2.300851e-07	3.041578e-07	4.020766e-07
## Room.Board	2.099501e-06	2.775406e-06	3.668904e-06	4.850042e-06
## Books	1.120788e-05	1.481607e-05	1.958583e-05	2.589105e-05
## Personal	3.685027e-06	4.871357e-06	6.439591e-06	8.512663e-06
## PhD	3.340251e-04	4.415587e-04	5.837096e-04	7.716211e-04
## Terminal	3.503502e-04	4.631391e-04	6.122372e-04	8.093319e-04
## S.F.Ratio	3.372647e-04	4.458416e-04	5.893720e-04	7.791076e-04
## perc.alumni	-1.016428e-04	-1.343654e-04	-1.776223e-04	-2.348049e-04
## Expend	6.939872e-07	9.174057e-07	1.212748e-06	1.603168e-06
## Grad.Rate	1.192629e-04	1.576579e-04	2.084133e-04	2.755082e-04
##				
## (Intercept)	3.001226e+03	3.001093e+03	3.000917e+03	3.000685e+03
## PrivateYes	-4.132789e-02	-5.463231e-02	-7.221943e-02	-9.546761e-02
## Accept	1.640870e-05	2.169111e-05	2.867397e-05	3.790460e-05
## Enroll	3.885167e-05	5.135896e-05	6.789237e-05	8.974770e-05
## Top10perc	8.188276e-04	1.082426e-03	1.430876e-03	1.891486e-03
## Top25perc	7.569010e-04	1.000562e-03	1.322656e-03	1.748423e-03
## F.Undergrad	7.158475e-06	9.462948e-06	1.250923e-05	1.653605e-05
## P.Undergrad	1.115175e-05	1.474169e-05	1.948720e-05	2.576015e-05
## Outstate	5.315179e-07	7.026289e-07	9.288227e-07	1.227829e-06
## Room.Board	6.411413e-06	8.475410e-06	1.120382e-05	1.481048e-05
## Books	3.422599e-05	4.524395e-05	5.980845e-05	7.906082e-05
## Personal	1.125307e-05	1.487560e-05	1.966414e-05	2.599393e-05
## PhD	1.020022e-03	1.348383e-03	1.782437e-03	2.356195e-03
## Terminal	1.069872e-03	1.414279e-03	1.869542e-03	2.471334e-03
## S.F.Ratio	1.029921e-03	1.361472e-03	1.799746e-03	2.379089e-03
## perc.alumni	-3.103958e-04	-4.103208e-04	-5.424126e-04	-7.170248e-04
## Expend	2.119270e-06	2.801507e-06	3.703354e-06	4.895487e-06

## Grad.Rate	3.642021e-04	4.814477e-04	6.364350e-04	8.413111e-04
##				
## (Intercept)	3.000378e+03	2.999972e+03	2.999436e+03	2.998728e+03
## PrivateYes	-1.261987e-01	-1.668205e-01	-2.205150e-01	-2.914872e-01
## Accept	5.010642e-05	6.623559e-05	8.755580e-05	1.157371e-04
## Enroll	1.186377e-04	1.568258e-04	2.073038e-04	2.740246e-04
## Top10perc	2.500350e-03	3.305170e-03	4.368985e-03	5.775095e-03
## Top25perc	2.311225e-03	3.055152e-03	4.038464e-03	5.338147e-03
## F.Undergrad	2.185897e-05	2.889502e-05	3.819533e-05	5.048815e-05
## P.Undergrad	3.405202e-05	4.501237e-05	5.949950e-05	7.864750e-05
## Outstate	1.623083e-06	2.145560e-06	2.836199e-06	3.749103e-06
## Room.Board	1.957804e-05	2.588008e-05	3.421030e-05	4.522114e-05
## Books	1.045095e-04	1.381480e-04	1.826106e-04	2.413779e-04
## Personal	3.436086e-05	4.542027e-05	6.003807e-05	7.935833e-05
## PhD	3.114610e-03	4.117083e-03	5.442109e-03	7.193391e-03
## Terminal	3.266802e-03	4.318245e-03	5.707987e-03	7.544785e-03
## S.F.Ratio	3.144893e-03	4.157149e-03	5.495133e-03	7.263589e-03
## perc.alumni	-9.478422e-04	-1.252952e-03	-1.656261e-03	-2.189360e-03
## Expend	6.471320e-06	8.554309e-06	1.130760e-05	1.494678e-05
## Grad.Rate	1.112132e-03	1.470116e-03	1.943308e-03	2.568767e-03
##				
## (Intercept)	2.997791e+03	2.996553e+03	2.994917e+03	2.992756e+03
## PrivateYes	-3.852928e-01	-5.092713e-01	-6.731164e-01	-8.896277e-01
## Accept	1.529861e-04	2.022187e-04	2.672862e-04	3.532756e-04
## Enroll	3.622115e-04	4.787650e-04	6.327988e-04	8.363477e-04
## Top10perc	7.633556e-03	1.008975e-02	1.333568e-02	1.762483e-02
## Top25perc	7.055903e-03	9.326068e-03	1.232603e-02	1.628995e-02
## F.Undergrad	6.673569e-05	8.820898e-05	1.165867e-04	1.540852e-04
## P.Undergrad	1.039545e-04	1.373994e-04	1.815948e-04	2.399893e-04
## Outstate	4.955772e-06	6.550676e-06	8.658624e-06	1.144447e-05
## Room.Board	5.977467e-05	7.900985e-05	1.044311e-04	1.380250e-04
## Books	3.190478e-04	4.216934e-04	5.573333e-04	7.365512e-04
## Personal	1.048922e-04	1.386355e-04	1.832228e-04	2.421309e-04
## PhD	9.507918e-03	1.256660e-02	1.660827e-02	2.194811e-02
## Terminal	9.972299e-03	1.318024e-02	1.741903e-02	2.301913e-02
## S.F.Ratio	9.600896e-03	1.268983e-02	1.677172e-02	2.216514e-02
## perc.alumni	-2.893997e-03	-3.825329e-03	-5.056223e-03	-6.682916e-03
## Expend	1.975666e-05	2.611346e-05	3.451403e-05	4.561430e-05
## Grad.Rate	3.395458e-03	4.488071e-03	5.932048e-03	7.840215e-03
##				
## (Intercept)	2.989901e+03	2.986129e+03	2.981149e+03	2.974575e+03
## PrivateYes	-1.175699e+00	-1.553617e+00	-2.052764e+00	-2.711842e+00
## Accept	4.669028e-04	6.170317e-04	8.153543e-04	1.077282e-03
## Enroll	1.105296e-03	1.460603e-03	1.929898e-03	2.549582e-03
## Top10perc	2.329173e-02	3.077762e-02	4.066407e-02	5.371687e-02
## Top25perc	2.152679e-02	2.844392e-02	3.757810e-02	4.963573e-02
## F.Undergrad	2.036293e-04	2.690773e-04	3.555148e-04	4.696386e-04
## P.Undergrad	3.171323e-04	4.190218e-04	5.535581e-04	7.311360e-04
## Outstate	1.512592e-05	1.999033e-05	2.641689e-05	3.490560e-05
## Room.Board	1.824143e-04	2.410594e-04	3.185240e-04	4.208216e-04

## Books	9.733094e-04	1.286015e-03	1.698915e-03	2.243908e-03
## Personal	3.199451e-04	4.227080e-04	5.583750e-04	7.374059e-04
## PhD	2.900181e-02	3.831720e-02	5.061558e-02	6.684535e-02
## Terminal	3.041632e-02	4.018479e-02	5.308040e-02	7.009669e-02
## S.F.Ratio	2.929038e-02	3.870160e-02	5.112883e-02	6.753273e-02
## perc.alumni	-8.832477e-03	-1.167262e-02	-1.542458e-02	-2.038004e-02
## Expend	6.027981e-05	7.965214e-05	1.052357e-04	1.390111e-04
## Grad.Rate	1.036150e-02	1.369241e-02	1.809202e-02	2.390169e-02
##				
## (Intercept)	2.965902e+03	2.954465e+03	2.939425e+03	2.919615e+03
## PrivateYes	-3.581768e+00	-4.729432e+00	-6.233352e+00	-8.219727e+00
## Accept	1.423113e-03	1.879542e-03	2.479281e-03	3.271232e-03
## Enroll	3.367556e-03	4.446752e-03	5.865580e-03	7.737187e-03
## Top10perc	7.094314e-02	9.366509e-02	1.234633e-01	1.627915e-01
## Top25perc	6.554521e-02	8.652428e-02	1.140658e-01	1.503747e-01
## F.Undergrad	6.202569e-04	8.189357e-04	1.080537e-03	1.425229e-03
## P.Undergrad	9.654103e-04	1.274283e-03	1.680920e-03	2.216118e-03
## Outstate	4.611527e-05	6.091304e-05	8.010652e-05	1.056102e-04
## Room.Board	5.558679e-04	7.340686e-04	9.681769e-04	1.277205e-03
## Books	2.962898e-03	3.910821e-03	5.157797e-03	6.799542e-03
## Personal	9.735283e-04	1.284717e-03	1.694603e-03	2.233476e-03
## PhD	8.825141e-02	1.164640e-01	1.535431e-01	2.023430e-01
## Terminal	9.253728e-02	1.221084e-01	1.610165e-01	2.121783e-01
## S.F.Ratio	8.917556e-02	1.177127e-01	1.554149e-01	2.049711e-01
## perc.alumni	-2.692313e-02	-3.555925e-02	-4.699018e-02	-6.203821e-02
## Expend	1.835826e-04	2.423682e-04	3.198170e-04	4.218074e-04
## Grad.Rate	3.157065e-02	4.168924e-02	5.503211e-02	7.261216e-02
##				
## (Intercept)	2.893589e+03	2.859464e+03	2.814835e+03	2.756666e+03
## PrivateYes	-1.083054e+01	-1.425580e+01	-1.873878e+01	-2.458760e+01
## Accept	4.313543e-03	5.683450e-03	7.480629e-03	9.832733e-03
## Enroll	1.019896e-02	1.343185e-02	1.766852e-02	2.320551e-02
## Top10perc	2.144716e-01	2.822545e-01	3.709371e-01	4.865871e-01
## Top25perc	1.980678e-01	2.605879e-01	3.423264e-01	4.488189e-01
## F.Undergrad	1.878546e-03	2.473743e-03	3.253545e-03	4.272338e-03
## P.Undergrad	2.919218e-03	3.841059e-03	5.046534e-03	6.617502e-03
## Outstate	1.391149e-04	1.830451e-04	2.404990e-04	3.153944e-04
## Room.Board	1.683756e-03	2.217793e-03	2.917902e-03	3.833352e-03
## Books	8.955991e-03	1.178274e-02	1.547823e-02	2.029249e-02
## Personal	2.940897e-03	3.867519e-03	5.077710e-03	6.652156e-03
## PhD	2.663877e-01	3.502451e-01	4.597107e-01	6.020334e-01
## Terminal	2.793131e-01	3.671995e-01	4.818947e-01	6.309656e-01
## S.F.Ratio	2.701274e-01	3.556459e-01	4.676333e-01	6.138398e-01
## perc.alumni	-8.187126e-02	-1.079857e-01	-1.423270e-01	-1.874110e-01
## Expend	5.559164e-04	7.319621e-04	9.625494e-04	1.263709e-03
## Grad.Rate	9.575097e-02	1.261645e-01	1.660685e-01	2.183030e-01
##				
## (Intercept)	2.681184e+03	2.583797e+03	2.459083e+03	2.300896e+03
## PrivateYes	-3.218693e+01	-4.200761e+01	-5.461044e+01	-7.063796e+01
## Accept	1.290158e-02	1.688954e-02	2.204524e-02	2.866700e-02

## Enroll	3.041616e-02	3.976299e-02	5.180667e-02	6.720679e-02
## Top10perc	6.367677e-01	8.307198e-01	1.079430e+00	1.395473e+00
## Top25perc	5.869311e-01	7.649916e-01	9.927957e-01	1.281364e+00
## F.Undergrad	5.598483e-03	7.316479e-03	9.528407e-03	1.235378e-02
## P.Undergrad	8.655572e-03	1.128410e-02	1.464823e-02	1.891113e-02
## Outstate	4.126165e-04	5.381488e-04	6.991516e-04	9.039448e-04
## Room.Board	5.026378e-03	6.574448e-03	8.572188e-03	1.113233e-02
## Books	2.653542e-02	3.458251e-02	4.487442e-02	5.790499e-02
## Personal	8.690125e-03	1.131056e-02	1.465064e-02	1.885968e-02
## PhD	7.861062e-01	1.022544e+00	1.323529e+00	1.702228e+00
## Terminal	8.236776e-01	1.071060e+00	1.385717e+00	1.781174e+00
## S.F.Ratio	8.039645e-01	1.049912e+00	1.365901e+00	1.768257e+00
## perc.alumni	-2.464668e-01	-3.235980e-01	-4.239498e-01	-5.538557e-01
## Expend	1.655575e-03	2.163013e-03	2.816057e-03	3.649908e-03
## Grad.Rate	2.864733e-01	3.750986e-01	4.897532e-01	6.371694e-01
##				
## (Intercept)	2.102699e+03	1.858193e+03	1.562357e+03	1.212892e+03
## PrivateYes	-9.078521e+01	-1.157402e+02	-1.460801e+02	-1.821231e+02
## Accept	3.710153e-02	4.773555e-02	6.097600e-02	7.721862e-02
## Enroll	8.670785e-02	1.111015e-01	1.411562e-01	1.775092e-01
## Top10perc	1.792445e+00	2.283925e+00	2.881583e+00	3.592695e+00
## Top25perc	1.642289e+00	2.086509e+00	2.622293e+00	3.252482e+00
## F.Undergrad	1.592648e-02	2.038706e-02	2.586886e-02	3.247677e-02
## P.Undergrad	2.424391e-02	3.080575e-02	3.871160e-02	4.798583e-02
## Outstate	1.161831e-03	1.482801e-03	1.876995e-03	2.354183e-03
## Room.Board	1.438485e-02	1.847328e-02	2.354723e-02	2.975085e-02
## Books	7.418981e-02	9.420610e-02	1.182961e-01	1.465317e-01
## Personal	2.408536e-02	3.044880e-02	3.800512e-02	4.668916e-02
## PhD	2.171534e+00	2.741839e+00	3.417595e+00	4.192702e+00
## Terminal	2.270499e+00	2.863871e+00	3.564872e+00	4.365507e+00
## S.F.Ratio	2.274636e+00	2.902360e+00	3.665465e+00	4.570278e+00
## perc.alumni	-7.209162e-01	-9.339432e-01	-1.202659e+00	-1.537056e+00
## Expend	4.704175e-03	6.020961e-03	7.641498e-03	9.601295e-03
## Grad.Rate	8.252645e-01	1.063036e+00	1.360278e+00	1.727082e+00
##				
## (Intercept)	8.119302e+02	3.676396e+02	-1.048738e+02	-5.840186e+02
## PrivateYes	-2.237438e+02	-2.701898e+02	-3.199612e+02	-3.708258e+02
## Accept	9.680577e-02	1.199816e-01	1.468575e-01	1.774035e-01
## Enroll	2.205249e-01	2.701413e-01	3.257407e-01	3.860895e-01
## Top10perc	4.417174e+00	5.344901e+00	6.354404e+00	7.413996e+00
## Top25perc	3.971227e+00	4.760922e+00	5.590446e+00	6.415854e+00
## F.Undergrad	4.025977e-02	4.918130e-02	5.909490e-02	6.973446e-02
## P.Undergrad	5.850527e-02	6.994230e-02	8.172671e-02	9.305006e-02
## Outstate	2.923321e-03	3.592227e-03	4.366961e-03	5.249677e-03
## Room.Board	3.720795e-02	4.600595e-02	5.618143e-02	6.770887e-02
## Books	1.785530e-01	2.134149e-01	2.494996e-01	2.845599e-01
## Personal	5.625145e-02	6.619958e-02	7.577062e-02	8.396591e-02
## PhD	5.045248e+00	5.932914e+00	6.791138e+00	7.536361e+00
## Terminal	5.240559e+00	6.142699e+00	7.000602e+00	7.722594e+00
## S.F.Ratio	5.609676e+00	6.757072e+00	7.962231e+00	9.151880e+00

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## perc.alumni -1.946351e+00 -2.437586e+00 -3.014132e+00 -3.674509e+00
## Expend      1.192436e-02  1.461787e-02  1.766922e-02  2.104743e-02
## Grad.Rate   2.173117e+00  2.706728e+00  3.333853e+00  4.056757e+00
##
## (Intercept) -1.043992e+03 -1.458263e+03 -1.804182e+03 -2.066353e+03
## PrivateYes  -4.200536e+02 -4.648071e+02 -5.025740e+02 -5.318671e+02
## Accept      2.114821e-01  2.489066e-01  2.894948e-01  3.332152e-01
## Enroll      4.494218e-01  5.134601e-01  5.757010e-01  6.336451e-01
## Top10perc   8.483925e+00  9.530257e+00  1.052380e+01  1.144745e+01
## Top25perc   7.184314e+00  7.841425e+00  8.337870e+00  8.636505e+00
## F.Undergrad 8.072989e-02  9.162011e-02  1.019144e-01  1.111174e-01
## P.Undergrad 1.029346e-01  1.103401e-01  1.143603e-01  1.143717e-01
## Outstate    6.236439e-03  7.305875e-03  8.407063e-03  9.454535e-03
## Room.Board  8.050435e-02  9.438862e-02  1.091033e-01  1.243056e-01
## Books       3.159534e-01  3.409711e-01  3.573767e-01  3.637870e-01
## Personal    8.965814e-02  9.181531e-02  8.971699e-02  8.312910e-02
## PhD         8.076269e+00  8.323631e+00  8.216664e+00  7.731740e+00
## Terminal    8.207483e+00  8.360791e+00  8.115584e+00  7.447622e+00
## S.F.Ratio   1.023339e+01  1.112056e+01  1.174331e+01  1.205718e+01
## perc.alumni -4.410776e+00 -5.211799e+00 -6.060334e+00 -6.930192e+00
## Expend      2.470949e-02  2.860713e-02  3.269798e-02  3.694744e-02
## Grad.Rate   4.872354e+00  5.769915e+00  6.729143e+00  7.718258e+00
##
## (Intercept) -2.239443e+03 -2.326351e+03 -2.339483e+03 -2.293252e+03
## PrivateYes  -5.518657e+02 -5.634574e+02 -5.672886e+02 -5.651963e+02
## Accept      3.800196e-01  4.302041e-01  4.838692e-01  5.415447e-01
## Enroll      6.844843e-01  7.257911e-01  7.549247e-01  7.696982e-01
## Top10perc   1.231498e+01  1.314437e+01  1.399350e+01  1.492178e+01
## Top25perc   8.719261e+00  8.579376e+00  8.233177e+00  7.693607e+00
## F.Undergrad 1.187470e-01  1.243525e-01  1.275520e-01  1.279847e-01
## P.Undergrad 1.101826e-01  1.020863e-01  9.089490e-02  7.777716e-02
## Outstate    1.030837e-02  1.077920e-02  1.065072e-02  9.652874e-03
## Room.Board  1.395062e-01  1.541583e-01  1.676565e-01  1.793774e-01
## Books       3.599089e-01  3.465923e-01  3.255758e-01  2.990406e-01
## Personal    7.253859e-02  5.891017e-02  4.376305e-02  2.867105e-02
## PhD         6.894306e+00  5.767155e+00  4.452005e+00  3.046101e+00
## Terminal    6.386378e+00  5.007848e+00  3.424575e+00  1.760167e+00
## S.F.Ratio   1.209978e+01  1.191289e+01  1.163805e+01  1.139402e+01
## perc.alumni -7.799258e+00 -8.622008e+00 -9.368269e+00 -9.984593e+00
## Expend      4.131960e-02  4.578365e-02  5.027078e-02  5.471620e-02
## Grad.Rate   8.693721e+00  9.606567e+00  1.040288e+01  1.104309e+01
##
## (Intercept) -2.202704e+03 -2.084290e+03 -1.949206e+03 -1.805760e+03
## PrivateYes  -5.606455e+02 -5.533251e+02 -5.465423e+02 -5.407085e+02
## Accept      6.037569e-01  6.707425e-01  7.422987e-01  8.176940e-01
## Enroll      7.673029e-01  7.460597e-01  7.046287e-01  6.449477e-01
## Top10perc   1.596728e+01  1.723459e+01  1.872919e+01  2.048255e+01
## Top25perc   6.972337e+00  6.084209e+00  5.044212e+00  3.869387e+00
## F.Undergrad 1.254490e-01  1.199131e-01  1.117175e-01  1.012262e-01
## P.Undergrad 6.407257e-02  5.125971e-02  4.038183e-02  3.212733e-02

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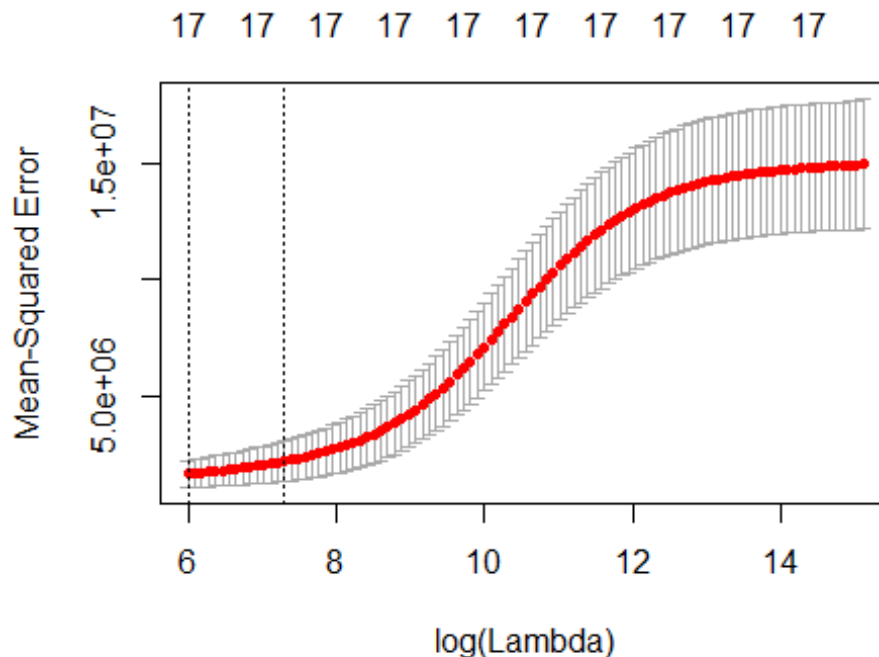
## Outstate	7.599283e-03	4.210320e-03	-4.361216e-04	-6.268056e-03
## Room.Board	1.889080e-01	1.957722e-01	2.001440e-01	2.022578e-01
## Books	2.696155e-01	2.391602e-01	2.098600e-01	1.827551e-01
## Personal	1.497891e-02	4.003623e-03	-3.891415e-03	-8.631141e-03
## PhD	1.643095e+00	3.111222e-01	-8.959688e-01	-1.964431e+00
## Terminal	1.490101e-01	-1.316956e+00	-2.554886e+00	-3.538636e+00
## S.F.Ratio	1.120432e+01	1.126906e+01	1.145528e+01	1.172025e+01
## perc.alumni	-1.039968e+01	-1.060097e+01	-1.052992e+01	-1.017522e+01
## Expend	5.904154e-02	6.315135e-02	6.690717e-02	7.017382e-02
## Grad.Rate	1.150245e+01	1.178011e+01	1.187639e+01	1.180568e+01
##				
## (Intercept)	-1.660613e+03	-1.517138e+03	-1.378126e+03	-1.246208e+03
## PrivateYes	-5.348672e+02	-5.299185e+02	-5.259011e+02	-5.225365e+02
## Accept	8.963028e-01	9.763456e-01	1.055837e+00	1.132347e+00
## Enroll	5.658504e-01	4.699828e-01	3.594529e-01	2.392115e-01
## Top10perc	2.256065e+01	2.491605e+01	2.748685e+01	3.017251e+01
## Top25perc	2.547772e+00	1.099547e+00	-4.520017e-01	-2.060498e+00
## F.Undergrad	8.924180e-02	7.672094e-02	6.482096e-02	5.441166e-02
## P.Undergrad	2.690040e-02	2.446828e-02	2.434292e-02	2.583796e-02
## Outstate	-1.328091e-02	-2.108063e-02	-2.927710e-02	-3.746180e-02
## Room.Board	2.019966e-01	1.999132e-01	1.964290e-01	1.920703e-01
## Books	1.579315e-01	1.357697e-01	1.162840e-01	9.935594e-02
## Personal	-1.012061e-02	-9.031903e-03	-6.037431e-03	-1.891095e-03
## PhD	-2.913702e+00	-3.755372e+00	-4.506115e+00	-5.176592e+00
## Terminal	-4.251564e+00	-4.706505e+00	-4.934534e+00	-4.983937e+00
## S.F.Ratio	1.220338e+01	1.276267e+01	1.333607e+01	1.386374e+01
## perc.alumni	-9.602712e+00	-8.838767e+00	-7.939901e+00	-6.967000e+00
## Expend	7.294585e-02	7.515295e-02	7.680655e-02	7.795251e-02
## Grad.Rate	1.162342e+01	1.135576e+01	1.104009e+01	1.070623e+01
##				
## (Intercept)	-1.123888e+03	-1.012783e+03	-913.94835456	-826.99736562
## PrivateYes	-5.195540e+02	-5.167427e+02	-514.02650361	-511.46633915
## Accept	1.203749e+00	1.268440e+00	1.32554820	1.37516661
## Enroll	1.131081e-01	-1.298535e-02	-0.13479022	-0.24930224
## Top10perc	3.285961e+01	3.544319e+01	37.83907977	40.00784506
## Top25perc	-3.672347e+00	-5.228199e+00	-6.67978792	-8.00010082
## F.Undergrad	4.631037e-02	4.062809e-02	0.03728163	0.03586360
## P.Undergrad	2.826309e-02	3.103741e-02	0.03375669	0.03620796
## Outstate	-4.526283e-02	-5.239849e-02	-0.05869930	-0.06411550
## Room.Board	1.872442e-01	1.823499e-01	0.17764696	0.17327607
## Books	8.485524e-02	7.253803e-02	0.06219630	0.05355615
## Personal	2.759499e-03	7.404768e-03	0.01171931	0.01556643
## PhD	-5.771209e+00	-6.294093e+00	-6.74729818	-7.13727588
## Terminal	-4.905766e+00	-4.749943e+00	-4.55665726	-4.35364685
## S.F.Ratio	1.432001e+01	1.468453e+01	14.95828735	15.14733557
## perc.alumni	-5.983495e+00	-5.038068e+00	-4.16612125	-3.38214013
## Expend	7.868238e-02	7.907562e-02	0.07922345	0.07918792
## Grad.Rate	1.038301e+01	1.008311e+01	9.81592001	9.58026314
##				
## (Intercept)	-753.28822577	-692.11373494	-640.86788169	-599.03612139

## PrivateYes	-508.95287354	-506.54977167	-504.41689891	-502.53082427
## Accept	1.41686360	1.45118499	1.47959696	1.50259631
## Enroll	-0.35373409	-0.44604226	-0.52706158	-0.59621253
## Top10perc	41.88632597	43.46814236	44.80196206	45.89581751
## Top25perc	-9.15553164	-10.13809652	-10.97066941	-11.65784920
## F.Undergrad	0.03616475	0.03760631	0.03968098	0.04205273
## P.Undergrad	0.03823363	0.03982386	0.04106082	0.04198650
## Outstate	-0.06863348	-0.07232103	-0.07530760	-0.07768552
## Room.Board	0.16938050	0.16601715	0.16311030	0.16067042
## Books	0.04655094	0.04095334	0.03644576	0.03288578
## Personal	0.01879932	0.02142721	0.02358055	0.02529262
## PhD	-7.45949941	-7.72008191	-7.93307674	-8.10262469
## Terminal	-4.16370623	-3.99761613	-3.85556809	-3.73904265
## S.F.Ratio	15.27689427	15.36155302	15.40751864	15.43027045
## perc.alumni	-2.70716550	-2.14083469	-1.66401003	-1.27310742
## Expend	0.07907971	0.07894462	0.07877944	0.07862053
## Grad.Rate	9.38772754	9.23438255	9.10753698	9.00682496
##				
## (Intercept)	-565.93241041	-539.61642150	-518.83668369	-503.09448370
## PrivateYes	-500.87592417	-499.48621216	-498.34469883	-497.40998332
## Accept	1.52071649	1.53504316	1.54630392	1.55483604
## Enroll	-0.65317376	-0.69987628	-0.73770495	-0.76699616
## Top10perc	46.76519645	47.45697849	48.00307048	48.41749265
## Top25perc	-12.20824480	-12.64839655	-12.99727434	-13.26405131
## F.Undergrad	0.04440861	0.04659999	0.04854416	0.05014443
## P.Undergrad	0.04265623	0.04314293	0.04349538	0.04374464
## Outstate	-0.07954675	-0.08099957	-0.08212718	-0.08298876
## Room.Board	0.15869563	0.15709570	0.15581189	0.15483153
## Books	0.03013345	0.02799823	0.02634917	0.02511358
## Personal	0.02660925	0.02763117	0.02842147	0.02900635
## PhD	-8.23365331	-8.33556080	-8.41443219	-8.47301799
## Terminal	-3.64701175	-3.57464094	-3.51831111	-3.47619076
## S.F.Ratio	15.44156191	15.44420669	15.44192424	15.43959147
## perc.alumni	-0.96224314	-0.71441463	-0.51828778	-0.36857065
## Expend	0.07849033	0.07837943	0.07828706	0.07822280
## Grad.Rate	8.93045478	8.87125786	8.82547563	8.79248754
##				
## (Intercept)	-491.02115481	-481.25296842	-474.09675513	-468.37823221
## PrivateYes	-496.69865637	-496.15524903	-495.78686337	-495.46990089
## Accept	1.56134911	1.56657487	1.57037976	1.57342391
## Enroll	-0.78972459	-0.80827042	-0.82190613	-0.83293277
## Top10perc	48.73383014	48.98870976	49.17293410	49.32167492
## Top25perc	-13.46893193	-13.63322040	-13.75355389	-13.84963169
## F.Undergrad	0.05144140	0.05254153	0.05337216	0.05405741
## P.Undergrad	0.04392375	0.04405842	0.04415063	0.04422150
## Outstate	-0.08364474	-0.08415210	-0.08452506	-0.08481543
## Room.Board	0.15408038	0.15346445	0.15302332	0.15266006
## Books	0.02418557	0.02344931	0.02292695	0.02250281
## Personal	0.02944411	0.02979663	0.03004294	0.03024412
## PhD	-8.51723703	-8.55291705	-8.57846765	-8.59908438

## Terminal	-3.44448941	-3.41901063	-3.40094464	-3.38619805
## S.F.Ratio	15.43471108	15.42775084	15.41933783	15.41385709
## perc.alumni	-0.25298023	-0.16035411	-0.09148242	-0.03745197
## Expend	0.07817436	0.07812618	0.07809507	0.07806537
## Grad.Rate	8.76788370	8.74712195	8.73272168	8.72061765
##				
## (Intercept)	-463.91980972	-461.02111173	-458.34025487	-456.44652837
## PrivateYes	-495.23105966	-495.11642787	-494.93282045	-494.81566173
## Accept	1.57578864	1.57730842	1.57874959	1.57976411
## Enroll	-0.84158303	-0.84714316	-0.85245946	-0.85622367
## Top10perc	49.43730296	49.51060748	49.58148153	49.63148611
## Top25perc	-13.92421114	-13.97241791	-14.01806211	-14.05006991
## F.Undergrad	0.05460602	0.05496090	0.05530371	0.05554883
## P.Undergrad	0.04427393	0.04430365	0.04433646	0.04435735
## Outstate	-0.08503650	-0.08517874	-0.08531769	-0.08541079
## Room.Board	0.15237501	0.15220337	0.15202450	0.15190020
## Books	0.02217631	0.02196800	0.02176842	0.02162334
## Personal	0.03040016	0.03049064	0.03058891	0.03065658
## PhD	-8.61505414	-8.62524452	-8.63477946	-8.64159239
## Terminal	-3.37481733	-3.36763436	-3.36077039	-3.35585393
## S.F.Ratio	15.40887244	15.40299688	15.40200143	15.40084137
## perc.alumni	0.00448985	0.03192099	0.05740087	0.07492396
## Expend	0.07804044	0.07802653	0.07801279	0.07800149
## Grad.Rate	8.71102185	8.70498909	8.69950463	8.69529939
##				
## (Intercept)	-455.20105116	-453.98456323	-452.86297188	-451.84835886
## PrivateYes	-494.74972770	-494.67935924	-494.59702893	-494.51499270
## Accept	1.58042733	1.58107161	1.58167350	1.58222178
## Enroll	-0.85868268	-0.86108039	-0.86334136	-0.86542543
## Top10perc	49.66388915	49.69546400	49.72537149	49.75293930
## Top25perc	-14.07104685	-14.09143941	-14.11038988	-14.12750748
## F.Undergrad	0.05570944	0.05586697	0.05601734	0.05615822
## P.Undergrad	0.04437017	0.04438371	0.04439723	0.04440950
## Outstate	-0.08547123	-0.08553156	-0.08558732	-0.08563580
## Room.Board	0.15182225	0.15174238	0.15166282	0.15158682
## Books	0.02152805	0.02144149	0.02136005	0.02128466
## Personal	0.03069906	0.03074167	0.03078510	0.03082703
## PhD	-8.64605668	-8.65033865	-8.65432575	-8.65797767
## Terminal	-3.35268272	-3.34962934	-3.34675317	-3.34409686
## S.F.Ratio	15.39928193	15.39797135	15.39783801	15.39831030
## perc.alumni	0.08653536	0.09804528	0.10855891	0.11783496
## Expend	0.07799458	0.07798773	0.07798042	0.07797237
## Grad.Rate	8.69255472	8.69002005	8.68759740	8.68521947
##				
## (Intercept)	-450.93718935	-450.12509864	-449.40828988	-448.78203801
## PrivateYes	-494.44372716	-494.38698125	-494.34369837	-494.31092043
## Accept	1.58271243	1.58314568	1.58352431	1.58385256
## Enroll	-0.86731493	-0.86900691	-0.87050793	-0.87183019
## Top10perc	49.77774680	49.79963953	49.81867734	49.83505781
## Top25perc	-14.14268361	-14.15596655	-14.16748474	-14.17740294

## F.Undergrad	0.05628848	0.05640780	0.05651630	0.05661440
## P.Undergrad	0.04441994	0.04442850	0.04443533	0.04444067
## Outstate	-0.08567637	-0.08570968	-0.08573690	-0.08575920
## Room.Board	0.15151662	0.15145333	0.15139718	0.15134787
## Books	0.02121738	0.02115933	0.02111048	0.02107003
## Personal	0.03086568	0.03090023	0.03093052	0.03095675
## PhD	-8.66127163	-8.66418687	-8.66671380	-8.66886205
## Terminal	-3.34170848	-3.33962422	-3.33785463	-3.33638505
## S.F.Ratio	15.39869727	15.39869024	15.39829058	15.39762940
## perc.alumni	0.12596523	0.13310284	0.13936925	0.14485142
## Expend	0.07796379	0.07795514	0.07794689	0.07793936
## Grad.Rate	8.68290574	8.68071987	8.67872326	8.67695243
##				
## (Intercept)	-448.23991308	-447.77403369	-447.37579398	-447.03658576
## PrivateYes	-494.28566020	-494.26563405	-494.24931729	-494.23574722
## Accept	1.58413542	1.58437807	1.58458556	1.58476260
## Enroll	-0.87298890	-0.87400038	-0.87488088	-0.87564584
## Top10perc	49.84904766	49.86093457	49.87099879	49.87949892
## Top25perc	-14.18589778	-14.19314405	-14.19930682	-14.20453726
## F.Undergrad	0.05670268	0.05678179	0.05685247	0.05691543
## P.Undergrad	0.04444476	0.04444782	0.04445003	0.04445156
## Outstate	-0.08577755	-0.08579271	-0.08580527	-0.08581569
## Room.Board	0.15130484	0.15126744	0.15123502	0.15120696
## Books	0.02103687	0.02100986	0.02098793	0.02097018
## Personal	0.03097926	0.03099848	0.03101481	0.03102865
## PhD	-8.67065946	-8.67214544	-8.67336362	-8.67435648
## Terminal	-3.33518427	-3.33421412	-3.33343638	-3.33281648
## S.F.Ratio	15.39684653	15.39604270	15.39527541	15.39457095
## perc.alumni	0.14962049	0.15374478	0.15729384	0.16033700
## Expend	0.07793270	0.07792694	0.07792202	0.07791786
## Grad.Rate	8.67541692	8.67410665	8.67300081	8.67207451
##				
## (Intercept)	-446.74830401	-446.50361548	-446.29605124	-446.11999448
## PrivateYes	-494.22430860	-494.21458243	-494.20626019	-494.19910151
## Accept	1.58491344	1.58504187	1.58515120	1.58524428
## Enroll	-0.87630955	-0.87688492	-0.87738348	-0.87781541
## Top10perc	49.88666627	49.89270385	49.89778753	49.90206822
## Top25perc	-14.20897065	-14.21272600	-14.21590665	-14.21860143
## F.Undergrad	0.05697139	0.05702103	0.05706499	0.05710387
## P.Undergrad	0.04445255	0.04445312	0.04445338	0.04445340
## Outstate	-0.08582434	-0.08583154	-0.08583752	-0.08584251
## Room.Board	0.15118272	0.15116179	0.15114373	0.15112815
## Books	0.02095584	0.02094428	0.02093498	0.02092752
## Personal	0.03104034	0.03105021	0.03105852	0.03106553
## PhD	-8.67516240	-8.67581455	-8.67634098	-8.67676502
## Terminal	-3.33232477	-3.33193652	-3.33163139	-3.33139275
## S.F.Ratio	15.39393765	15.39337506	15.39287905	15.39244418
## perc.alumni	0.16294053	0.16516527	0.16706547	0.16868856
## Expend	0.07791437	0.07791146	0.07790904	0.07790702
## Grad.Rate	8.67130271	8.67066221	8.67013233	8.66969506

```
cv.out = cv.glmnet(x,y,alpha =0)
plot(cv.out)
```



```
bestlam =cv.out$lambda.min
bestlam
```

```
## [1] 400.4766
```

```
predict(ridge.mod,s=bestlam,type="coefficients")[1:18,]
```

```
## (Intercept) PrivateYes Accept Enroll Top10perc
## -1.512577e+03 -5.297867e+02 9.789533e-01 4.663570e-01 2.500038e+01
## Top25perc F.Undergrad P.Undergrad Outstate Room.Board
## 1.048650e+00 7.633057e-02 2.446417e-02 -2.134950e-02 1.997989e-01
## Books Personal PhD Terminal S.F.Ratio
## 1.351305e-01 -8.933673e-03 -3.779999e+00 -4.713986e+00 1.278148e+01
## perc.alumni Expend Grad.Rate
## -8.809281e+00 7.520720e-02 1.134540e+01
```

*# (e) Now split the data set into a training set and a test set.
 # i. Fit the best models obtained in the best subset selection (according to C_p , BIC or adjusted R^2) to the training set, and report the test error obtained.*

```
set.seed(1)
train= sample(c(TRUE,FALSE), nrow(College), rep=TRUE)
test=(!train)
```



```

regfit.best = regsubsets(Apps~., data=College[train,], nvmax=18)
regfit.summary = summary(regfit.best)

par(mfrow = c(2,2))

#1
plot(regfit.summary$cp ,xlab = " Number of Variables ",ylab="Cp", type="l")
points(which.min (regfit.summary$cp), regfit.summary$cp [which.min
(regfit.summary$cp)], col ="red",cex =2, pch =20)

#2
plot(regfit.summary$bic ,xlab=" Number of Variables ",ylab=" BIC", type="l")
points (which.min (regfit.summary$bic ), regfit.summary$bic [which.min
(regfit.summary$bic )], col =" red",cex =2, pch =20)

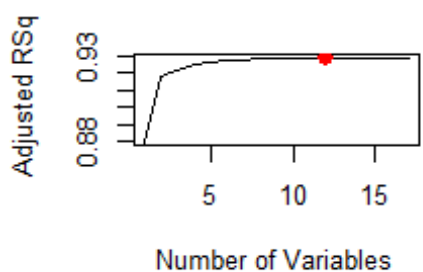
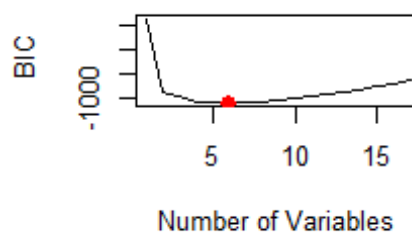
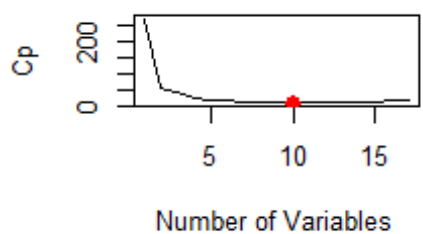
#3
plot(regfit.summary$adjr2 ,xlab = " Number of Variables ", ylab=" Adjusted
RSq",type="l")
points (which.max(regfit.summary$adjr2),
regfit.summary$adjr2[which.max(regfit.summary$adjr2)], col ="red",cex =2, pch
=20)

#building an "X" matrix from test data
test.mat = model.matrix(Apps~.,data=College[test,])

coefi=coef(regfit.best,id=which.min (regfit.summary$bic))
pred=test.mat[,names(coefi)]%*%coefi
mean((College$Apps[test]-pred)^2)

## [1] 1635683

```



ii. Fit a lasso model to the training set, with λ chosen by cross validation. Report the test error obtained.

```
library(glmnet)
set.seed(1)
lasso.mod = glmnet(x[train,], y[train], alpha = 1, lambda = grid)
cv.out = cv.glmnet(x[train,], y[train], alpha = 1)
bestlam = cv.out$lambda.min
lasso.pred = predict(lasso.mod, s = bestlam, newx = x[test,])
mean((lasso.pred - y[test])^2)

## [1] 1612366
```

iii. Fit a ridge regression model to the training set, with λ chosen by cross validation. Report the test error obtained.

```
ridge.mod = glmnet(x[train,], y[train], alpha = 0, lambda = grid)
ridge.pred = predict(ridge.mod, s = 4, newx = x[test,])
mean((ridge.pred - y[test])^2)

## [1] 1547703
```

iv. Compare the test errors obtained in the above analysis (i-iii) and determine the optimal model.

The following are the test errors for the above three models :
best subset selection - 1635683

```

# lasso model - 1612366
# ridge model - 1547703
# Ridge regression model is the optimal model.

# In the lab, a classification tree was applied to the Carseats data set
after converting Sales into a binary response variable. This question will
seek to predict Sales using regression trees and related approaches, treating
the response as a quantitative variable (that is, without the conversion).
# (a) Split the data set into a training set and a test set.

library(ISLR)
set.seed(1)
train <- sample(1:nrow(Carseats), nrow(Carseats) / 2)
Carseats_train <- Carseats[train, ]
Carseats_test <- Carseats[-train, ]

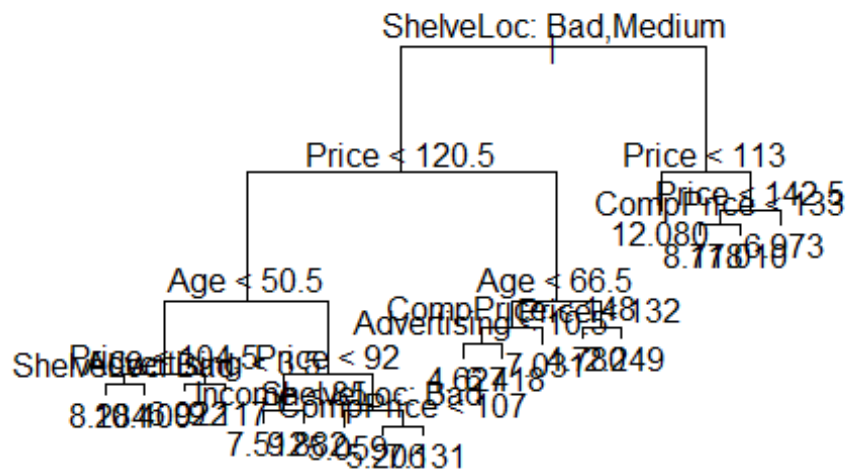
# (b) Fit a regression tree to the training set. Plot the tree, and interpret
the results. Then compute the test MSE.

library(tree)
tree_carseats <- tree(Sales ~ ., data = Carseats_train)
summary(tree_carseats)

##
## Regression tree:
## tree(formula = Sales ~ ., data = Carseats_train)
## Variables actually used in tree construction:
## [1] "ShelveLoc" "Price" "Age" "Advertising" "Income"
## [6] "CompPrice"
## Number of terminal nodes: 18
## Residual mean deviance: 2.36 = 429.5 / 182
## Distribution of residuals:
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## -4.2570 -1.0360 0.1024 0.0000 0.9301 3.9130

plot(tree_carseats)
text(tree_carseats, pretty = 0)

```

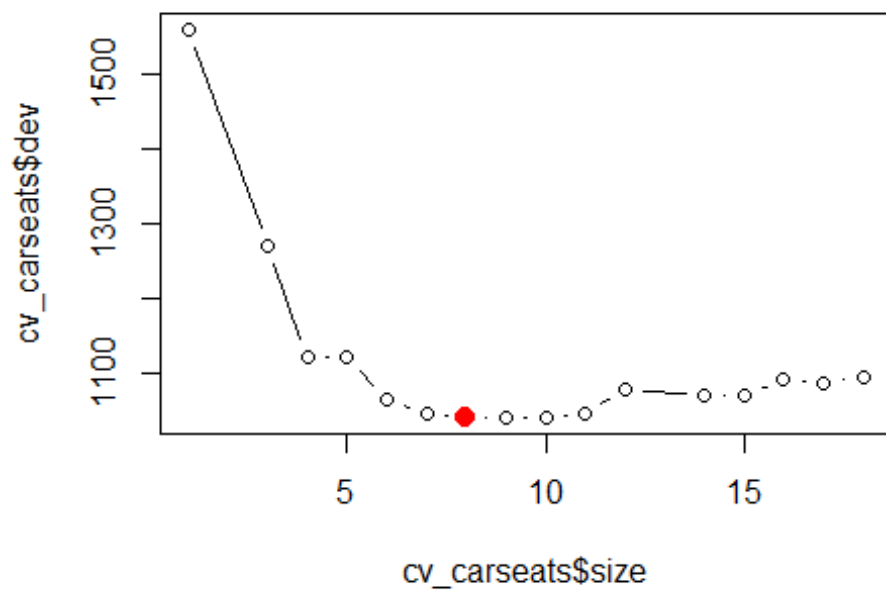


```
yhat <- predict(tree_carseats, Carseats_test)
mean((yhat - Carseats_test$Sales)^2)
```

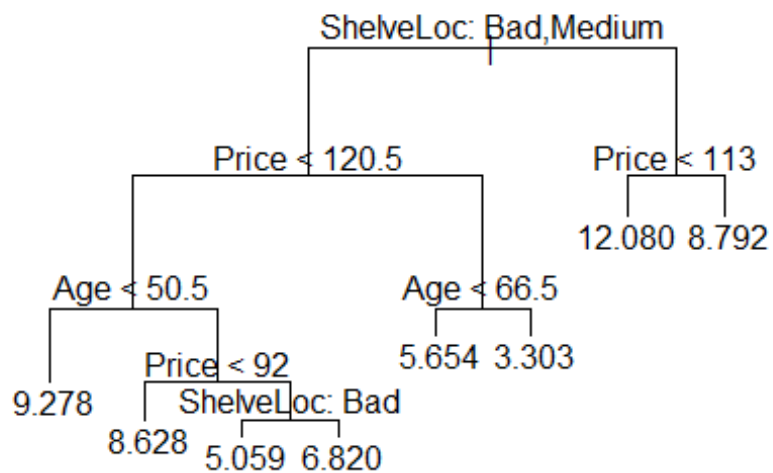
```
## [1] 4.148897
```

(c) Prune the tree obtained in (b). Use cross validation to determine the optimal level of tree complexity. Plot the pruned tree and interpret the results. Compute the test MSE of the pruned tree. Does pruning improve the test error?

```
cv_carseats <- cv.tree(tree_carseats)
plot(cv_carseats$size, cv_carseats$dev, type = "b")
tree_min <- which.min(cv_carseats$dev)
points(tree_min, cv_carseats$dev[tree_min], col = "red", cex = 2, pch = 20)
```



```
prune_carseats <- prune.tree(tree_carseats, best = 8)
plot(prune_carseats)
text(prune_carseats, pretty = 0)
```



##(d) Use the bagging approach to analyze the data. What test MSE do you obtain? Determine which variables are most important.

```
library(randomForest)

## randomForest 4.6-14

## Type rfNews() to see new features/changes/bug fixes.

bag.carseats <- randomForest(Sales ~ ., data = Carseats_train, mtry = 10,
                             ntree = 500)
yhat.bag <- predict(bag.carseats, newdata = Carseats_test)
mean((yhat.bag - Carseats_test$Sales)^2)

## [1] 2.553118
```

```
importance(bag.carseats)
```

```
##              IncNodePurity
## CompPrice      130.725006
## Income         77.495031
## Advertising    124.129450
## Population     62.078103
## Price          505.500563
## ShelfLoc       325.698430
## Age            192.250330
## Education      40.007558
## Urban          8.741943
## US             14.682941
```

We may conclude that "Price" and "ShelfLoc" are the two most important variables.

(e) Use random forests to analyze the data. What test MSE do you obtain? Determine which variables are most important.

```
rf.carseats <- randomForest(Sales ~ ., data = Carseats_train, mtry = 3, ntree
                             = 500, importance = TRUE)
yhat.rf <- predict(rf.carseats, newdata = Carseats_test)
mean((yhat.rf - Carseats_test$Sales)^2)

## [1] 3.286144
```

```
importance(rf.carseats)
```

```
##              %IncMSE IncNodePurity
## CompPrice      8.3370363    129.69721
## Income         4.0540506    127.70485
## Advertising    11.9343057    139.03662
## Population     0.6266783     97.93305
## Price          38.4595875    388.45041
## ShelfLoc       29.8867032    234.02006
```

```
## Age          18.1019639    197.17165
## Education    2.3638614     63.07024
## Urban        -0.5623597     15.93606
## US           6.2576901     32.71135
```

We may conclude that, in this case also, "Price" and "ShelveLoc" are the two most important variables.

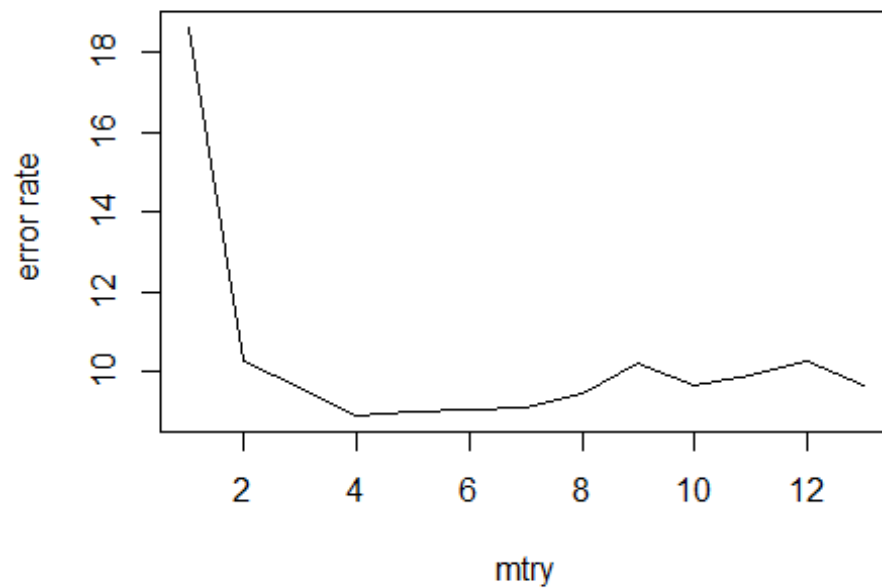
```
library (MASS)
library (randomForest)
set.seed (100)
train = sample (1: nrow(Boston), nrow(Boston )/2)
x_train = Boston[train, -14]
x_test = Boston[-train, -14]
y_train = Boston[train, 14]
y_test = Boston[-train, 14]
test_error1 = rep(0,13)

for (i in 1:13){
  bag.boston = randomForest(x_train, y_train, mtry= i, ntree =100, importance
=TRUE)
  yhat.bag = predict(bag.boston, newdata = Boston[-train,])
  test_error1[i] = mean(( yhat.bag - y_test)^2)
}

test_error1

## [1] 18.611813 10.282254 9.624311 8.896569 8.990730 9.052735 9.084105
## [8] 9.429814 10.228730 9.661823 9.908464 10.251536 9.673689

mtry = 1:13
plot(mtry, test_error1, xlab = " mtry ", ylab=" error rate",type="l")
```



```
test_error2 = rep(0,196)
for (i in 5:200){
  bag.boston = randomForest(medv~., data=Boston[train,] , mtry= 6, ntree = i ,
  importance =TRUE)
  yhat.bag = predict (bag.boston, newdata = Boston[-train,])
  test_error2[i-4] = mean(( yhat.bag - y_test)^2)
}

ntree = 5:200
plot(ntree, test_error2, xlab = " ntree ", ylab=" error rate",type="l")
```