Carprice prediction Lasso Regression

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Load the data

```
load("rda/carPrice.rda")
```

library(glmnet)

Create a matrix "x" of all independent variables and store dependent variable in "y".

```
x <- model.matrix(price~.,data=carPrice)[,-1]
y <-carPrice$price</pre>
```

Divide you data in 70:30

```
set.seed(1)
train= sample(1:nrow(x), 0.7*nrow(x))
```

Store indices into test which is not present in train

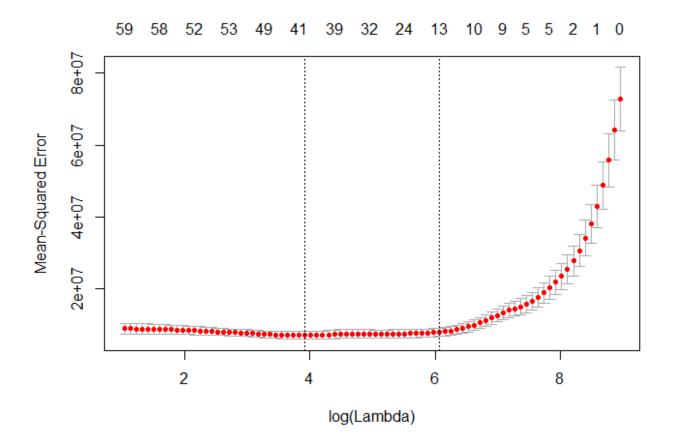
```
test = (-train)
```

Dependend variable for test data

```
y.test = y[test]
```

Crossvalidation for finding the lambda values

```
cv.out <- cv.glmnet(x[train,],y[train],alpha=1)
plot(cv.out)</pre>
```



Optimal lamda store it into "minlamda_lasso" object

```
minlamda_lasso <-cv.out$lambda.min
minlamda_lasso
## [1] 50.55804</pre>
```

But when we check in the plot, MSE value is constant upto log(6) #6= log(lambda). Working with value equivalent to log lambda =6 will help to eliminate some redundant variable is an advantage.

Lambda equivalent to log lambda = 6

```
lambda <- exp(6)
lambda
## [1] 403.4288
```

Modeling

```
lasso.mod <- glmnet(x[train,],y[train],alpha=1,lambda = 403.4)</pre>
```

Prediction

```
lasso.pred <- predict(lasso.mod,s= 403.4,newx=x[test,])</pre>
```

MSE

```
mean((lasso.pred-y.test)^2)
```

```
## [1] 6480779
```

All the coefficents from the model at optimal lamda, s=403.4

```
lasso.coef <- predict(lasso.mod,type="coefficients",s=403.4)
lasso.coef</pre>
```

```
## 65 x 1 sparse Matrix of class "dgCMatrix"
##
                                       1
## (Intercept)
                           -32376.420860
## symboling
## carCompanyaudi
                             5476.945846
## carCompanybmw
## carCompanychevrolet
## carCompanydodge
## carCompanyhonda
## carCompanyisuzu
## carCompanyjaguar
## carCompanymazda
## carCompanymercedes-benz 4394.739442
## carCompanymercury
## carCompanymitsubishi
                            -228.632583
## carCompanynissan
                            -559.445212
## carCompanypeugot
## carCompanyplymouth
## carCompanyporsche
                           2799.236344
## carCompanyrenault
## carCompanysaab
## carCompanysubaru
## carCompanytoyota
```

```
## car.companyvorkswagen
## carCompanyvolvo
## fueltypegas
## aspirationturbo
## doornumbertwo
## carbodyhardtop
## carbodyhatchback
                           -174.074806
## carbodysedan
## carbodywagon
## drivewheelfwd
## drivewheelrwd
                           1422.239768
## enginelocationrear
                           5542.015048
## wheelbase
## carlength
## carwidth
                              457.841142
## carheight
## curbweight
                                1.725973
## enginetypedohcv
## enginetypel
## enginetypeohc
## enginetypeohcf
## enginetypeohcv
## enginetyperotor
## cylindernumberfive
## cylindernumberfour
                          -1378.247822
## cylindernumbersix
## cylindernumberthree
## cylindernumbertwelve
## cylindernumbertwo
## enginesize
                               74.264085
## fuelsystem2bbl
## fuelsystem4bbl
## fuelsystemidi
## fuelsystemmfi
## fuelsystemmpfi
## fuelsystemspdi
## fuelsystemspfi
## boreratio
## stroke
## compressionratio
## horsepower
                               17.272555
## peakrpm
## citympg
## highwaympg
```

Non zero coefficients in final model

```
lasso.coef <- predict(lasso.mod,type="coefficients",s=403.4)[1:65,]
lasso.coef[lasso.coef!=0]</pre>
```

##	(Intercept)	carCompanybmw	carCompanymercedes-benz
##	-32376.420860	5476.945846	4394.739442
##	carCompanymitsubishi	carCompanynissan	carCompanyporsche
##	-228.632583	-559.445212	2799.236344
##	carbodyhatchback	drivewheelrwd	enginelocationrear
##	-174.074806	1422.239768	5542.015048
##	carwidth	curbweight	cylindernumberfour
##	457.841142	1.725973	-1378.247822
##	enginesize	horsepower	
##	74.264085	17.272555	