# 46-926 Homework 4, Part I

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```
library(quantmod)
load("/Users/apple/retmat.Robj")
stockdata = getSymbols("MSFT", auto.assign=FALSE, from="2016-1-1", to="2016-12-31")
msft = dailyReturn(stockdata[,6])
msft = as.numeric(msft)
library(glmnet)
```

#### 1 Lasso Procedure

```
retmat = as.matrix(retmat)
glmnetout = glmnet(retmat, msft)
print(glmnetout)
##
## Call: glmnet(x = retmat, y = msft)
##
##
           Df
                 %Dev
                         Lambda
##
     [1,]
            0 0.00000 0.0118900
##
     [2,]
            2 0.06192 0.0113500
##
     [3,]
            2 0.11840 0.0108300
##
     [4,]
            2 0.16990 0.0103400
##
     [5,]
            2 0.21680 0.0098720
##
     [6,]
            2 0.25950 0.0094230
##
     [7,]
            2 0.29850 0.0089950
##
     [8,]
            2 0.33390 0.0085860
##
     [9,]
            3 0.36640 0.0081960
##
    [10,]
            3 0.39590 0.0078230
##
    [11,]
            3 0.42290 0.0074680
##
   [12,]
            3 0.44740 0.0071280
   [13,]
            3 0.46980 0.0068040
##
   [14,]
            3 0.49020 0.0064950
   [15,]
            3 0.50880 0.0062000
##
##
   [16,]
            3 0.52570 0.0059180
  [17,]
            3 0.54110 0.0056490
## [18,]
            3 0.55520 0.0053920
  [19,]
##
            3 0.56800 0.0051470
##
  [20,]
            3 0.57960 0.0049130
##
  [21,]
            3 0.59030 0.0046900
##
   [22,]
            3 0.59990 0.0044770
##
   [23,]
            3 0.60880 0.0042730
##
   [24,]
            3 0.61680 0.0040790
##
  [25,]
            3 0.62410 0.0038940
   [26,]
##
            3 0.63080 0.0037170
## [27,]
            3 0.63690 0.0035480
  [28,]
            3 0.64240 0.0033860
##
   [29,]
            3 0.64750 0.0032330
```

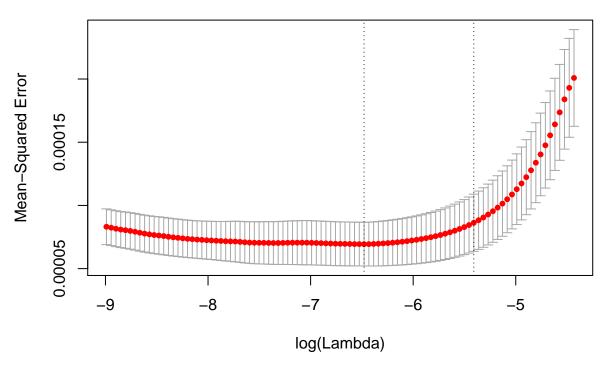
```
[30,]
##
            3 0.65210 0.0030860
##
    [31,]
            3 0.65630 0.0029450
    [32,]
            3 0.66010 0.0028120
    [33,]
            3 0.66350 0.0026840
##
##
    [34,]
            3 0.66670 0.0025620
##
    [35,]
            3 0.66960 0.0024450
    [36,]
            3 0.67220 0.0023340
    [37,]
            3 0.67460 0.0022280
##
##
    [38,]
            3 0.67680 0.0021270
##
    [39,]
            3 0.67880 0.0020300
    [40,]
            3 0.68060 0.0019380
##
    [41,]
            3 0.68230 0.0018500
    [42,]
##
            3 0.68380 0.0017660
##
    [43,]
            3 0.68510 0.0016850
##
    [44,]
            3 0.68640 0.0016090
##
    [45,]
            4 0.68760 0.0015360
##
    [46,]
            4 0.68890 0.0014660
##
    [47,]
            5 0.69040 0.0013990
##
    [48,]
            5 0.69200 0.0013360
##
    [49,]
            5 0.69330 0.0012750
##
    [50,]
            6 0.69470 0.0012170
    [51,]
            7 0.69770 0.0011620
    [52,]
            7 0.70050 0.0011090
##
##
    ſ53.1
            8 0.70320 0.0010590
##
    [54,]
            8 0.70580 0.0010100
    [55,]
           10 0.70830 0.0009645
##
    [56,]
           15 0.71200 0.0009206
    [57,]
           15 0.71680 0.0008788
##
    [58,]
           16 0.72120 0.0008389
    [59,]
           18 0.72590 0.0008007
##
    [60,]
           21 0.73120 0.0007643
##
    [61,]
           24 0.73730 0.0007296
##
    [62,]
           26 0.74430 0.0006964
    [63,]
           28 0.75100 0.0006648
##
##
    [64,]
           32 0.75770 0.0006346
##
    [65,]
           33 0.76440 0.0006057
##
    [66,]
           34 0.77060 0.0005782
##
    [67,]
           39 0.77670 0.0005519
##
    [68,]
           42 0.78280 0.0005268
##
    [69,]
           45 0.78910 0.0005029
    [70,]
           47 0.79540 0.0004800
##
    [71,]
           49 0.80130 0.0004582
    [72,]
           54 0.80700 0.0004374
##
    [73,]
           58 0.81300 0.0004175
    [74,]
           63 0.81950 0.0003985
    [75,]
##
           71 0.82600 0.0003804
    [76,]
##
           74 0.83250 0.0003631
##
    [77,]
           80 0.83890 0.0003466
    [78,]
           85 0.84530 0.0003309
##
    [79,]
           90 0.85210 0.0003158
##
    [80,]
           90 0.85870 0.0003015
##
    [81,]
           91 0.86460 0.0002878
##
    [82,]
           95 0.87010 0.0002747
##
    [83,]
           99 0.87520 0.0002622
```

```
[84,] 101 0.88010 0.0002503
##
    [85,] 106 0.88510 0.0002389
    [86,] 109 0.88950 0.0002281
##
    [87,] 112 0.89390 0.0002177
##
##
    [88,] 115 0.89830 0.0002078
    [89,] 119 0.90240 0.0001983
##
##
    [90,] 122 0.90630 0.0001893
    [91,] 124 0.91000 0.0001807
##
##
    [92,] 132 0.91380 0.0001725
##
    [93,] 135 0.91780 0.0001647
    [94,] 141 0.92180 0.0001572
    [95,] 144 0.92580 0.0001500
##
    [96,] 148 0.92960 0.0001432
##
##
   [97,] 151 0.93330 0.0001367
##
   [98,] 153 0.93680 0.0001305
##
   [99,] 157 0.94010 0.0001246
## [100,] 160 0.94340 0.0001189
```

### 2 Five-Fold Cross Validation

```
cvglmout = cv.glmnet(retmat, msft)
plot(cvglmout)
```





## 3 Lambda Selection

```
optlambda=cvglmout$lambda.1se
print(optlambda)
```

## [1] 0.004476724

I would use

 $\lambda = 0.03716655$ 

in this case.

## 4 Selected ETF

optlambdapos=which(cvglmout\$glmnet.fit\$lambda==optlambda)
glmnetout\$beta[glmnetout\$beta[,optlambdapos]!=0,optlambdapos]

## IYW TECS XLK ## 0.17730140 -0.07376143 0.35970019

Therefore ITY, TECS, XLK are included in the final model.