

2. Run a LASSO version of logistic regression, using CV to estimate optimal shrinkage in the logistic regression parameters using minimum CV-error.

```

93 ### 2. Run a LASSO version of logistic regression, using CV to estimate optimal shrinkage
94 # in the logistic regression parameters using minimum CV-error.
95
96 # (a) Report a list of the variables included/excluded in each logit. Does
97 # the pattern seem somewhat consistent with the ANOVA results from
98 # earlier? Explain in a sentence.
99 X.train.scale = as.matrix(data.train.scale[, -19])
100 Y.train = data.train.scale[, 19]
101 X.valid.scale = as.matrix(data.valid.scale[, -19])
102 Y.valid = data.valid.scale[, 19]
103
104
105 ### Let's repeat our logistic regression analysis using the glmnet package.
106 # fit.log.glmnet = cv.glmnet(X.train.scale, Y.train, family = "multinomial")
107 all.LASSOs = cv.glmnet(X.train.scale, Y.train, family = "multinomial")
108
109 lambda.min = all.LASSOs$lambda.min
110 lambda.lse = all.LASSOs$lambda.lse
111 ### Get predictions and investigate performance. The predict() function
112 ### for glmnet() can give several different types of output. To get
113 ### predicted values, set type="class". Remember to set s=0 for logistic
114 ### regression.
115 # pred.log.glmnet = predict(fit.log.glmnet, X.valid.scale, type = "class",
116 #                           s = 0)
117 pred.log.glmnet.min = predict(fit.log.glmnet, X.valid.scale, type = "class",
118                             s = lambda.min)
119 pred.log.glmnet.lse = predict(fit.log.glmnet, X.valid.scale, type = "class",
120                             s = lambda.lse)
121
122
123 # table(Y.valid, pred.log.glmnet, dnn = c("Observed", "Predicted"))
124 table(Y.valid, pred.log.glmnet.min, dnn = c("Observed", "Predicted"))
125 table(Y.valid, pred.log.glmnet.lse, dnn = c("Observed", "Predicted"))
126
127 (misclass.log.glmnet = mean(Y.valid != pred.log.glmnet.min))
128 (misclass.log.glmnet = mean(Y.valid != pred.log.glmnet.lse))
129
131 lasso.min.coef = coef(all.LASSOs, s = "lambda.min", exact = TRUE, family = "multinomial")
132 lasso.lse.coef = coef(all.LASSOs, s = "lambda.lse", exact = TRUE, family = "multinomial")

```

(a) Report a list of the variables included/excluded in each logit. Does the pattern seem somewhat consistent with the ANOVA results from earlier? Explain in a sentence.

```

>lasso.min.coef
(Intercept)          10.47789955
Compactness          -6.43735489
Circularity           4.33621387
Distance.Circularity -0.28083026
Radius.Ratio          18.69867552
Pr.Axis.Aspect.Ratio -28.96879784
Max.Length.Aspect.Ratio -4.05453435
Scatter.Ratio         0.05192484
Elongatedness         3.17196076
Pr.Axis.Rectangularity .
Max.Length.Rectangularity -7.57664804
Scaled.Variance.Along.Major.Axis -1.51693002
Scaled.Variance.Along.Minor.Axis 15.67578846
Scaled.Radius.of.Gyration -3.10578781
Skewness.About.Major.Axis -14.41921897
Skewness.About.Minor.Axis  0.04195768

```

Kurtosis.About.Minor.Axis	0.20849110
Kurtosis.About.Major.Axis	1.07739497
Hollows.Ratio	-5.46839934

\$`4D`

19 x 1 sparse Matrix of class "dgCMatrix"

1

(Intercept)	13.5155248
Compactness	4.5758051
Circularity	-12.3226861
Distance.Circularity	0.2808303
Radius.Ratio	21.2169082
Pr.Axis.Aspect.Ratio	-28.5843066
Max.Length.Aspect.Ratio	-10.0694529
Scatter.Ratio	.
Elongatedness	-3.1719608
Pr.Axis.Rectangularity	8.0309236
Max.Length.Rectangularity	-0.8662893
Scaled.Variance.Along.Major.Axis	.
Scaled.Variance.Along.Minor.Axis	-10.6377616
Scaled.Radius.of.Gyration	6.2865542
Skewness.About.Major.Axis	-10.7250672
Skewness.About.Minor.Axis	0.1011371
Kurtosis.About.Minor.Axis	-0.2084911
Kurtosis.About.Major.Axis	-12.7160699
Hollows.Ratio	5.4683993

\$BUS

19 x 1 sparse Matrix of class "dgCMatrix"

1

(Intercept)	0.1965590
Compactness	-4.5758051
Circularity	.
Distance.Circularity	-6.5642736
Radius.Ratio	-64.6005083
Pr.Axis.Aspect.Ratio	105.6089591
Max.Length.Aspect.Ratio	4.0545344
Scatter.Ratio	.
Elongatedness	-19.7803722
Pr.Axis.Rectangularity	.
Max.Length.Rectangularity	0.8662893
Scaled.Variance.Along.Major.Axis	46.2307520
Scaled.Variance.Along.Minor.Axis	.
Scaled.Radius.of.Gyration	3.1057878
Skewness.About.Major.Axis	10.7250672
Skewness.About.Minor.Axis	-5.1044260

Kurtosis.About.Minor.Axis	3.1220789
Kurtosis.About.Major.Axis	47.7261462
Hollows.Ratio	-25.2995979

\$VAN

19 x 1 sparse Matrix of class "dgCMatrix"

1

(Intercept)	-24.18998339
Compactness	12.94557435
Circularity	.
Distance.Circularity	21.91993544
Radius.Ratio	-18.69867552
Pr.Axis.Aspect.Ratio	28.58430664
Max.Length.Aspect.Ratio	29.33960573
Scatter.Ratio	-35.12848043
Elongatedness	9.29536204
Pr.Axis.Rectangularity	-19.40725058
Max.Length.Rectangularity	37.68734783
Scaled.Variance.Along.Major.Axis	.
Scaled.Variance.Along.Minor.Axis	.
Scaled.Radius.of.Gyration	-19.04715083
Skewness.About.Major.Axis	35.85145712
Skewness.About.Minor.Axis	-0.04195768
Kurtosis.About.Minor.Axis	-3.22501509
Kurtosis.About.Major.Axis	-1.07739497
Hollows.Ratio	13.67304898

Lasso.1se.coef

(Intercept)	7.49601576
Compactness	-6.19399153
Circularity	1.88339320
Distance.Circularity	-0.01332511
Radius.Ratio	11.93254099
Pr.Axis.Aspect.Ratio	-19.56533341
Max.Length.Aspect.Ratio	-0.06776796
Scatter.Ratio	.
Elongatedness	1.95032211
Pr.Axis.Rectangularity	.
Max.Length.Rectangularity	-4.05991526
Scaled.Variance.Along.Major.Axis	-0.77740303
Scaled.Variance.Along.Minor.Axis	12.21014816
Scaled.Radius.of.Gyration	-2.43221848
Skewness.About.Major.Axis	-11.87408238
Skewness.About.Minor.Axis	0.23303965
Kurtosis.About.Minor.Axis	0.09133776

Kurtosis.About.Major.Axis	.
Hollows.Ratio	-4.22927907

\$`4D`

19 x 1 sparse Matrix of class "dgCMatrix"

1

(Intercept)	8.99540239
Compactness	3.44488105
Circularity	-11.36835569
Distance.Circularity	0.01332511
Radius.Ratio	14.84922077
Pr.Axis.Aspect.Ratio	-20.45341274
Max.Length.Aspect.Ratio	-5.35577614
Scatter.Ratio	.
Elongatedness	-1.95032211
Pr.Axis.Rectangularity	4.10408335
Max.Length.Rectangularity	-0.30553560
Scaled.Variance.Along.Major.Axis	.
Scaled.Variance.Along.Minor.Axis	-4.71025706
Scaled.Radius.of.Gyration	5.98296974
Skewness.About.Major.Axis	-9.00831800
Skewness.About.Minor.Axis	0.33918807
Kurtosis.About.Minor.Axis	-0.09133776
Kurtosis.About.Major.Axis	-10.81017188
Hollows.Ratio	4.22927907

\$BUS

19 x 1 sparse Matrix of class "dgCMatrix"

1

(Intercept)	0.73796804
Compactness	-3.44488105
Circularity	.
Distance.Circularity	-3.91462628
Radius.Ratio	-45.29986372
Pr.Axis.Aspect.Ratio	73.72073905
Max.Length.Aspect.Ratio	0.06776796
Scatter.Ratio	.
Elongatedness	-14.95283732
Pr.Axis.Rectangularity	.
Max.Length.Rectangularity	0.30553560
Scaled.Variance.Along.Major.Axis	30.29959732
Scaled.Variance.Along.Minor.Axis	.
Scaled.Radius.of.Gyration	2.43221848
Skewness.About.Major.Axis	9.00831800
Skewness.About.Minor.Axis	-3.78573575
Kurtosis.About.Minor.Axis	1.76934625

Kurtosis.About.Major.Axis	32.16611869
Hollows.Ratio	-17.62158854

\$VAN

19 x 1 sparse Matrix of class "dgCMatrix"

	1
(Intercept)	-17.2293862
Compactness	9.3044324
Circularity	.
Distance.Circularity	13.1945727
Radius.Ratio	-11.9325410
Pr.Axis.Aspect.Ratio	19.5653334
Max.Length.Aspect.Ratio	18.3302287
Scatter.Ratio	-23.2613542
Elongatedness	7.8143496
Pr.Axis.Rectangularity	-8.5801997
Max.Length.Rectangularity	26.7657963
Scaled.Variance.Along.Major.Axis	.
Scaled.Variance.Along.Minor.Axis	.
Scaled.Radius.of.Gyration	-13.5295229
Skewness.About.Major.Axis	21.9994724
Skewness.About.Minor.Axis	-0.2330397
Kurtosis.About.Minor.Axis	-2.3249012
Kurtosis.About.Major.Axis	.
Hollows.Ratio	7.3440734

The variable with numerical values (not just dot) looks like variables in a list from ANOVA

(b) Compute and **report training and test error. How does test error compare to other methods?**

```
> (miss.lasso.min = mean(Y.valid != pred.lasso.min))
[1] 0.2169811
> (miss.lasso.lse = mean(Y.valid != pred.lasso.lse))
[1] 0.2122642
```

-> The error is lower than other methods.