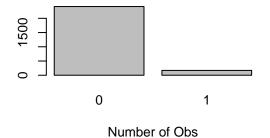
557_Project_2BS

Ben Straub, Hillary Koch, Jiawei Huang, Arif Masrur 3/15/2017

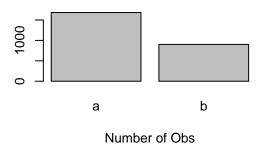
No Command Lines Ever. Whoa

What the Factor Variables look like

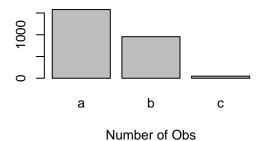
Class/Response Distribution



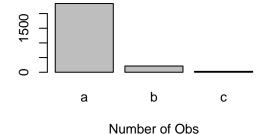
Seismic Distribution



Seismoacoustic Distribution

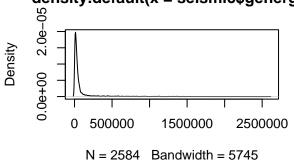


Ghazard Distribution

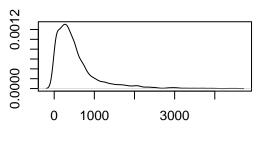


What the Continuous Variables look like

density.default(x = seismic\$genergy)



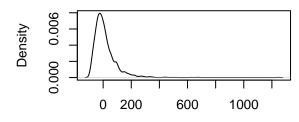
density.default(x = seismic\$gpuls)

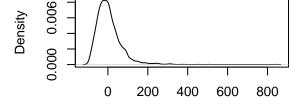


N = 2584 Bandwidth = 66.84

density.default(x = seismic\$gdenergy

density.default(x = seismic\$gdpuls)



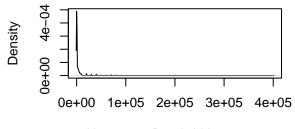


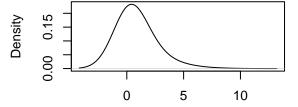
N = 2584 Bandwidth = 10.47

N = 2584 Bandwidth = 9.244

density.default(x = seismic\$maxenerg\$nsity.default(x = seismic\$nbumps, adjus

Density

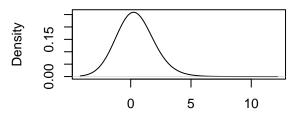


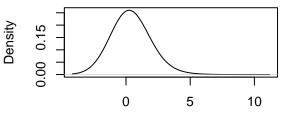


N = 2584 Bandwidth = 279.1

N = 2584 Bandwidth = 1.395

nsity.default(x = seismic\$nbumps2, adjusnsity.default(x = seismic\$nbumps3, adjus





N = 2584 Bandwidth = 1.395

N = 2584 Bandwidth = 1.395

Call:

```
lm(formula = class ~ ., data = seismic)
```

Residuals:

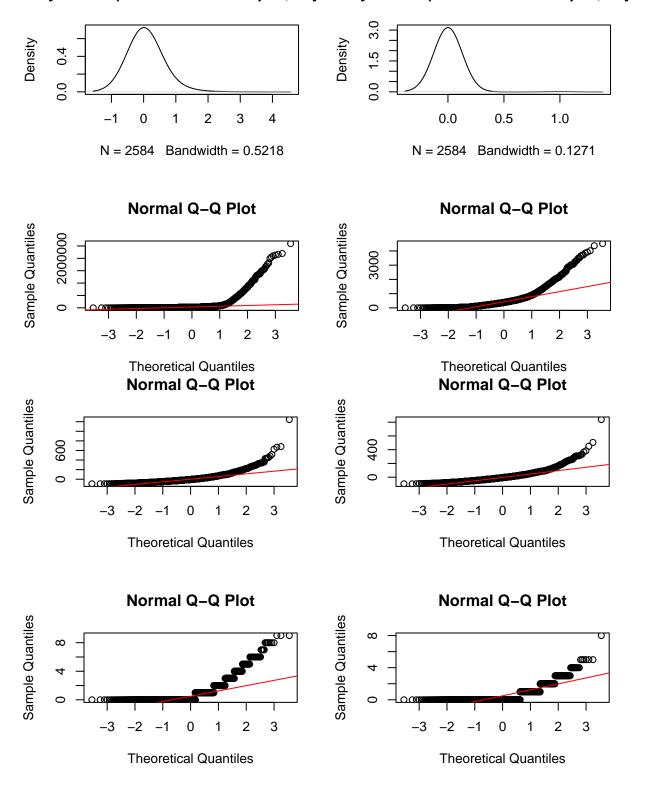
Coefficients:

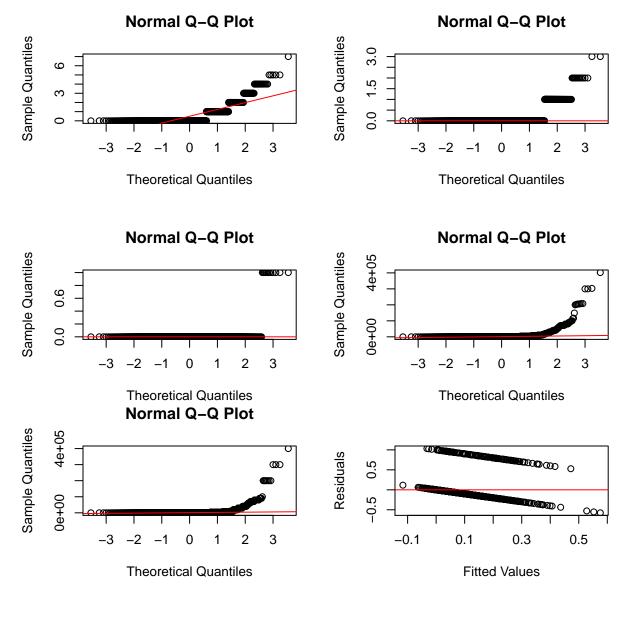
Estimate Std. Error t value Pr(>|t|) (Intercept) -2.393e-02 2.565e-02 -0.933 0.35090 seismic 1.869e-02 1.076e-02 1.737 0.08254 . seismoacoustic 2.610e-03 1.002e-02 0.260 0.79457 6.190e-04 1.157e-02 0.054 0.95732 shift genergy -8.698e-08 3.459e-08 -2.514 0.01199 * 1.019e-04 1.670e-05 6.102 1.2e-09 *** gpuls -6.943e-05 1.006e-04 -0.690 0.49009 gdenergy gdpuls -1.942e-04 1.368e-04 -1.420 0.15583 -1.394e-02 1.608e-02 -0.867 0.38618 ghazard nbumps 4.674e-01 1.680e-01 2.783 0.00543 ** nbumps2 -4.282e-01 1.682e-01 -2.546 0.01096 * -4.260e-01 1.681e-01 -2.535 0.01131 * nbumps3 nbumps4 -4.622e-01 1.708e-01 -2.706 0.00685 ** nbumps5 -2.963e-01 2.332e-01 -1.270 0.20408 2.536e-07 2.395e-06 0.106 0.91568 energy -1.054e-06 2.333e-06 -0.452 0.65164 maxenergy

Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1

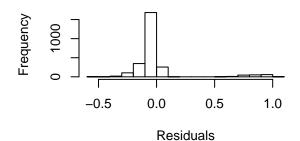
Residual standard error: 0.2371 on 2568 degrees of freedom Multiple R-squared: 0.09128, Adjusted R-squared: 0.08597 F-statistic: 17.2 on 15 and 2568 DF, p-value: < 2.2e-16

nsity.default(x = seismic\$nbumps4, adjusnsity.default(x = seismic\$nbumps5, adjus





Histogram of res

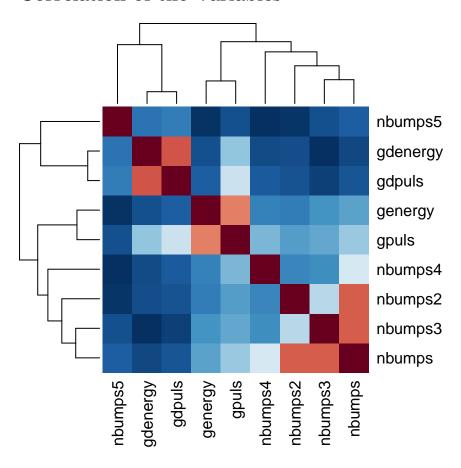


Lots of multicollinearity to worry about during variable selection

vif(fit)

##	seismic	seismoacoustic	shift	genergy	gpuls
##	1.209814	1.286183	1.411216	2.889651	4.057018
##	gdenergy	gdpuls	ghazard	nbumps	nbumps2
##	3.000282	3.430524	1.395598	2414.689538	798.964152
##	nbumps3	nbumps4	nbumps5	energy	maxenergy
##	769.131960	104.402690	11.562237	110.283444	93.762895

Correlation of the Variables

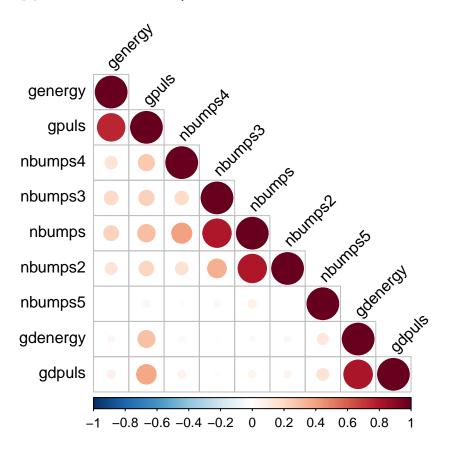


\$r

	genergy	gpuls	nbumps4	${\tt nbumps3}$	${\tt nbumps}$	${\tt nbumps2}$	${\tt nbumps5}$	gdenergy
genergy	1							
gpuls	0.75	1						
nbumps4	0.15	0.26	1					
nbumps3	0.19	0.23	0.18	1				
nbumps	0.22	0.3	0.4	0.8	1			
nbumps2	0.14	0.21	0.16	0.35	0.8	1		
nbumps5	-0.0099	0.049	-0.017	0.046	0.07	-0.0053	1	
gdenergy	0.049	0.29	0.037	-0.012	0.03	0.041	0.12	1

```
0.072 0.38
                          0.066
                                  0.015 0.058
                                                  0.051
                                                           0.14
                                                                    0.81
gdpuls
         gdpuls
genergy
gpuls
nbumps4
nbumps3
nbumps
nbumps2
nbumps5
gdenergy
gdpuls
              1
$p
         genergy gpuls nbumps4 nbumps3 nbumps nbumps2 nbumps5 gdenergy
genergy
               0
gpuls
                     0
nbumps4 1.4e-14
                     0
                              0
               0
                     0
                              0
                                      0
nbumps3
                     0
                              0
                                      0
                                             0
nbumps
               0
                                                      0
nbumps2 2.2e-13
                     0
                              0
                                      0
                                             0
            0.62 0.012
                                  0.018 4e-04
                                                   0.79
nbumps5
                            0.4
                                                              0
                                                  0.036 3.3e-10
gdenergy
           0.014
                          0.061
                                   0.54
                                          0.13
                                                                       0
         0.00027
                     0 0.00076
                                   0.45 0.0032 0.0094 5.9e-13
                                                                       0
gdpuls
         gdpuls
genergy
gpuls
nbumps4
nbumps3
nbumps
nbumps2
nbumps5
gdenergy
              0
gdpuls
$sym
         genergy gpuls nbumps4 nbumps3 nbumps nbumps2 nbumps5 gdenergy
genergy
gpuls
nbumps4
                        1
nbumps3
                                1
nbumps
                                        1
nbumps2
                                               1
nbumps5
                                                        1
gdenergy
                                                                1
gdpuls
         gdpuls
genergy
gpuls
nbumps4
nbumps3
nbumps
nbumps2
nbumps5
gdenergy
```

```
gdpuls 1
attr(,"legend")
[1] 0 ' ' 0.3 '.' 0.6 ',' 0.8 '+' 0.9 '*' 0.95 'B' 1
```



```
$r
        row
              column
                         cor
1
               gpuls
                      0.7500 0.0e+00
    genergy
2
    genergy
             nbumps4
                      0.1500 1.4e-14
                      0.2600 0.0e+00
3
             nbumps4
      gpuls
4
                      0.1900 0.0e+00
    genergy
             nbumps3
5
             nbumps3
                      0.2300 0.0e+00
     gpuls
6
    nbumps4
             nbumps3
                      0.1800 0.0e+00
7
              nbumps
                      0.2200 0.0e+00
    genergy
8
      gpuls
              nbumps
                      0.3000 0.0e+00
9
    nbumps4
              nbumps
                     0.4000 0.0e+00
10
    nbumps3
              nbumps
                      0.8000 0.0e+00
             nbumps2
                      0.1400 2.2e-13
11
    genergy
12
      gpuls
             nbumps2
                      0.2100 0.0e+00
13
             nbumps2 0.1600 0.0e+00
    nbumps4
14
    nbumps3
             nbumps2
                      0.3500 0.0e+00
             nbumps2 0.8000 0.0e+00
15
     nbumps
16
             nbumps5 -0.0099 6.2e-01
    genergy
17
      gpuls
             nbumps5 0.0490 1.2e-02
18
             nbumps5 -0.0170 4.0e-01
    nbumps4
19
    nbumps3
             nbumps5 0.0460 1.8e-02
20
     nbumps
             nbumps5 0.0700 4.0e-04
    nbumps2
             nbumps5 -0.0053 7.9e-01
```

```
22 genergy gdenergy 0.0490 1.4e-02
23
     gpuls gdenergy 0.2900 0.0e+00
24 nbumps4 gdenergy 0.0370 6.1e-02
25 nbumps3 gdenergy -0.0120 5.4e-01
   nbumps gdenergy 0.0300 1.3e-01
27 nbumps2 gdenergy 0.0410 3.6e-02
28 nbumps5 gdenergy 0.1200 3.3e-10
29 genergy gdpuls 0.0720 2.7e-04
30 gpuls gdpuls 0.3800 0.0e+00
31 nbumps4 gdpuls 0.0660 7.6e-04
32 nbumps3 gdpuls 0.0150 4.5e-01
   nbumps gdpuls 0.0580 3.2e-03
33
34 nbumps2 gdpuls 0.0510 9.4e-03
35 nbumps5 gdpuls 0.1400 5.9e-13
36 gdenergy gdpuls 0.8100 0.0e+00
$р
NULL
$sym
NULL
```

[1] 646 16

Separating into Test and Training Sets

```
## Setting up Test and Training Sets
##-----
n <- dim(seismic)[1]</pre>
p <- dim(seismic)[2]</pre>
set.seed(2016)
test <- sample(n, round(n/4))</pre>
train <- (1:n)[-test]
seismic.train <- seismic[train,]</pre>
seismic.test <- seismic[test,]</pre>
dim(seismic)
[1] 2584
           16
dim(seismic.train)
[1] 1938
           16
dim(seismic.test)
```

```
#View(seismic.train)
#View(seismic.test)
```

Linear regression of an indicator matrix

```
## Fit Linear Regression to Indicator Matrix
##-----
fit.lm <- lm(class~., data=seismic.train)</pre>
summary(fit.lm)
##
## Call:
## lm(formula = class ~ ., data = seismic.train)
## Residuals:
       Min
                1Q Median
                                 30
## -0.53532 -0.08061 -0.03978 -0.00442 1.03254
## Coefficients:
##
                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                -6.001e-02 2.980e-02 -2.014 0.0442 *
## seismic
                 3.007e-02 1.269e-02 2.370 0.0179 *
## seismoacoustic 1.309e-02 1.183e-02 1.106 0.2688
## shift
                1.359e-02 1.333e-02 1.019 0.3084
            -3.179e-08 4.150e-08 -0.766 0.4437
## genergy
                8.221e-05 2.004e-05 4.102 4.27e-05 ***
## gpuls
## gdenergy
               -3.462e-05 1.170e-04 -0.296 0.7673
## gdpuls
                -2.266e-04 1.598e-04 -1.418 0.1564
## ghazard
                -1.850e-02 1.931e-02 -0.958
                                              0.3382
## nbumps
                1.003e+00 2.397e-01 4.186 2.97e-05 ***
                -9.670e-01 2.398e-01 -4.033 5.73e-05 ***
## nbumps2
## nbumps3
                -9.681e-01 2.398e-01 -4.037 5.63e-05 ***
                -1.004e+00 2.425e-01 -4.139 3.63e-05 ***
## nbumps4
## nbumps5
                -7.912e-01 3.091e-01 -2.560
                                              0.0106 *
## energy
                2.757e-06 3.419e-06 0.806
                                             0.4201
## maxenergy
                -3.511e-06 3.346e-06 -1.050
                                              0.2940
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2394 on 1922 degrees of freedom
## Multiple R-squared: 0.09781,
                                Adjusted R-squared: 0.09077
## F-statistic: 13.89 on 15 and 1922 DF, p-value: < 2.2e-16
fit.probs <- predict(fit.lm, type="response")</pre>
# Train Data
fit.pred=rep("0",1938)
fit.pred[fit.probs >.5]="1"
```

```
confusion <- table(fit.pred ,seismic.train$class)</pre>
mean(fit.pred==seismic.train$class)
## [1] 0.9318885
sensitivity <- confusion[2,2]/sum(confusion[,2])</pre>
specificity <- confusion[1,1]/sum(confusion[,1])</pre>
## Sensitivity is very bad! Dramatically underpredict 1s
confusion
##
## fit.pred 0 1
      0 1805 130
##
          1 2 1
sensitivity
## [1] 0.007633588
specificity
## [1] 0.9988932
# Test Data
fit.probs <- predict(fit.lm, newdata=seismic.test, type="response")</pre>
fit.pred=rep("0",646)
fit.pred[fit.probs >.5]="1"
confusion <- table(fit.pred, seismic.test$class)</pre>
mean(fit.pred==seismic.test$class)
## [1] 0.9380805
sensitivity <- confusion[2,2]/sum(confusion[,2])</pre>
specificity <- confusion[1,1]/sum(confusion[,1])</pre>
## Sensitivity is very bad! Dramatically underpredict 1s
confusion
##
## fit.pred 0 1
##
      0 606 39
##
          1 1 0
sensitivity
```

[1] 0

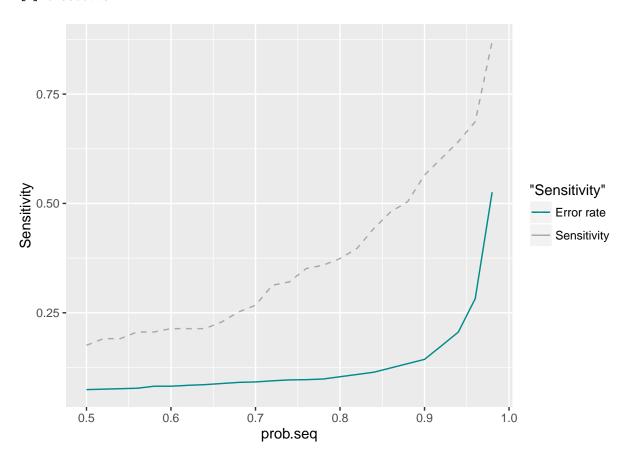
specificity

[1] 0.9983526

lda.class 0 1 0 1771 108 1 36 23

[1] 0.1755725

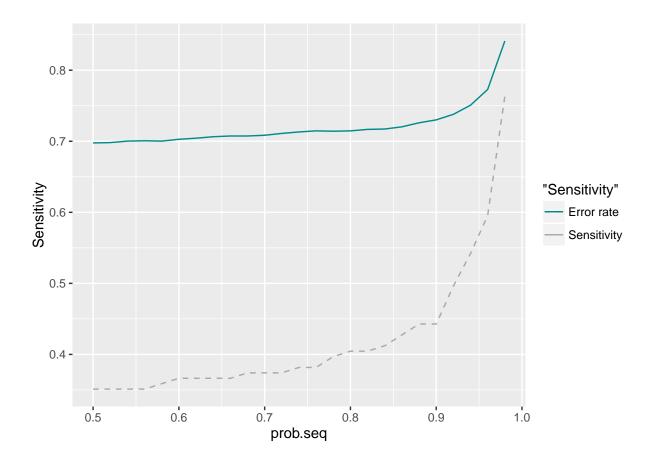
[1] 0.9800775



1da.class 0 1 0 591 34 1 16 5

[1] 0.1282051

[1] 0.9736409



Regularized Discriminant Analysis

Logistic Regression

```
Call:
glm(formula = class ~ ., family = binomial, data = seismic.train)
Deviance Residuals:
    Min
              1Q
                   Median
                                        Max
-1.8471 -0.3860
                  -0.2851 -0.1566
                                     3.0825
Coefficients:
                 Estimate Std. Error z value Pr(>|z|)
(Intercept)
               -6.343e+00 7.721e-01 -8.215 < 2e-16 ***
seismic
                4.808e-01
                          2.111e-01
                                       2.278 0.022727 *
seismoacoustic 2.159e-01
                          1.993e-01
                                       1.084 0.278524
shift
                1.179e+00 3.573e-01
                                       3.301 0.000965 ***
               -2.471e-07
                          5.044e-07
                                     -0.490 0.624239
genergy
gpuls
                7.095e-04
                          2.474e-04
                                       2.868 0.004136 **
gdenergy
               -1.904e-04
                           2.177e-03
                                     -0.087 0.930292
gdpuls
               -2.997e-03
                          3.093e-03
                                     -0.969 0.332500
ghazard
               -2.335e-01
                          3.509e-01 -0.666 0.505671
nbumps
                1.807e+01 5.354e+02
                                       0.034 0.973080
```

```
      nbumps2
      -1.773e+01
      5.354e+02
      -0.033
      0.973590

      nbumps3
      -1.771e+01
      5.354e+02
      -0.033
      0.973611

      nbumps4
      -1.806e+01
      5.354e+02
      -0.034
      0.973097

      nbumps5
      -1.604e+01
      5.354e+02
      -0.030
      0.976095

      energy
      1.622e-06
      4.033e-05
      0.040
      0.967929

      maxenergy
      -7.101e-06
      3.969e-05
      -0.179
      0.858012
```

Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 958.82 on 1937 degrees of freedom Residual deviance: 813.40 on 1922 degrees of freedom

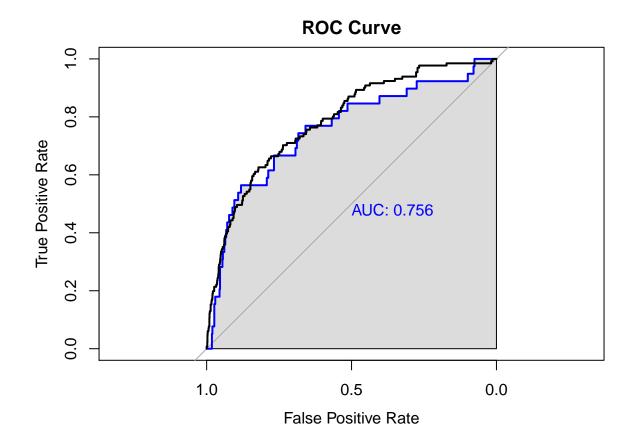
AIC: 845.4

Number of Fisher Scoring iterations: 12

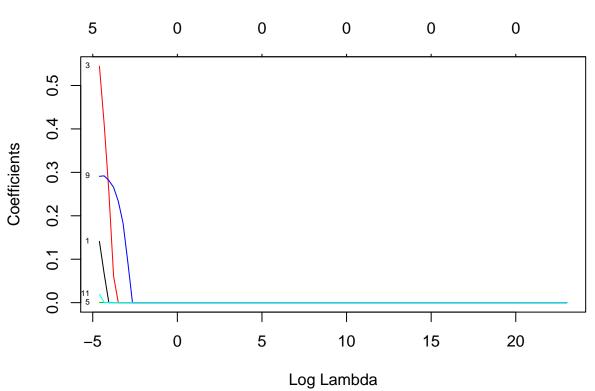
[1] 0.9329205

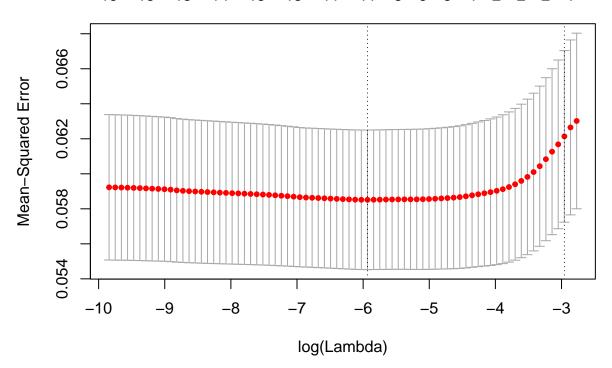
- [1] 0.04580153
- [1] 0.997233
- [1] 0.9349845

- [1] 0
- [1] 0.9950577



Variable Selection



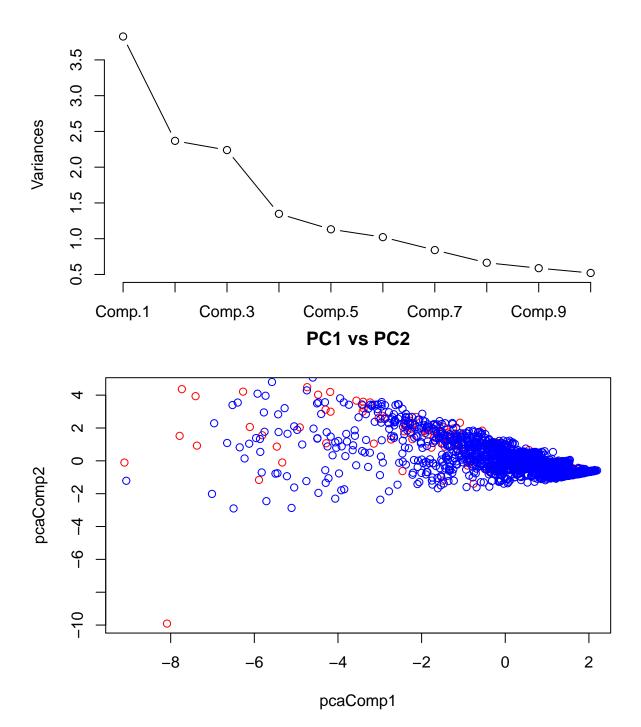


[1] 8.670049

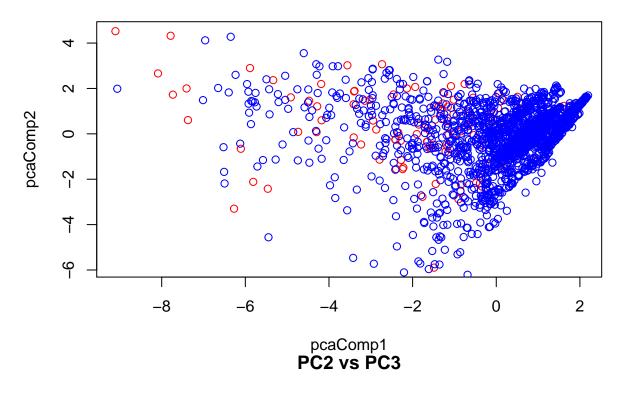
genergy	shift	seismoacoustic	seismic	(Intercept)
0.00000e+00	7.977504e-03	0.000000e+00	8.800484e-03	-8.144581e-03
nbumps	ghazard	gdpuls	gdenergy	gpuls
3.117955e-02	0.000000e+00	0.000000e+00	0.000000e+00	4.677101e-05
energy	nbumps5	nbumps4	nbumps3	nbumps2
0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00	0.000000e+00
				maxenergy
				0.000000e+00

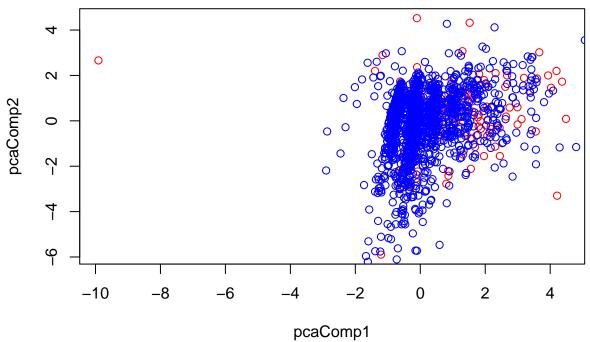
(Intercept) seismic shift gpuls nbumps -8.144581e-03 8.800484e-03 7.977504e-03 4.677101e-05 3.117955e-02





PC1 vs PC3





Data: X dimension: 2584 15 Y dimension: 2584 1

Fit method: svdpc

Number of components considered: 15

VALIDATION: RMSEP

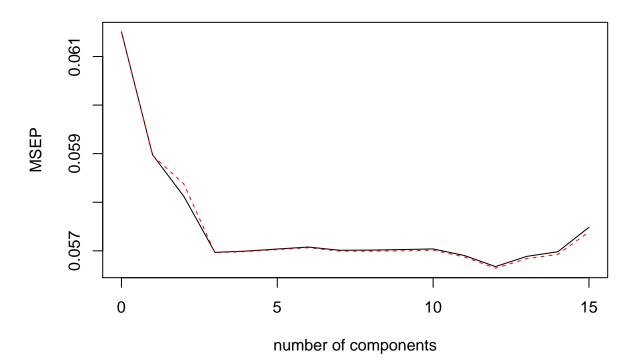
Cross-validated using 10 random segments.

	(Interce	pt) 1 co	mps 2 c	omps	3 com	ps	4 comps	5 comps	6 comps
CV	0.	248 0.2	428 0.	2411	0.23	87	0.2387	0.2388	0.2389
adjCV	0.	248 0.2	428 0.	2416	0.23	87	0.2387	0.2388	0.2389
	7 comps	8 comps	9 comps	10	comps	11	comps	12 comps	13 comps
CV	0.2388	0.2388	0.2388	0	.2388	(0.2385	0.2381	0.2385
adjCV	0.2387	0.2387	0.2387	0	.2388	(0.2385	0.2380	0.2384
	14 comps	15 comp	s						
CV	0.2387	0.239	8						
adiCV	0 2386	0 239	6						

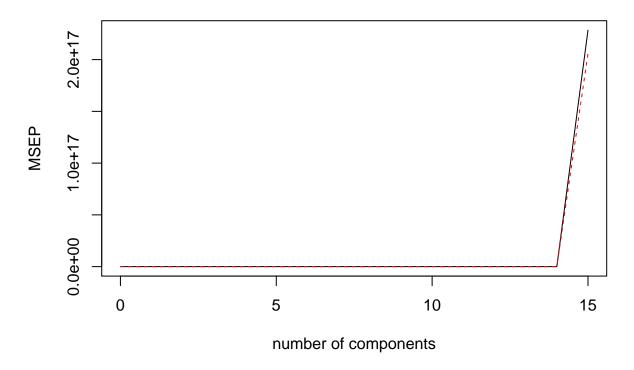
${\tt TRAINING:~\%~variance~explained}$

	1 comps	2 comps	3 comps	4 comps	5 comps	6 comps	7 cc	omps
X	25.306	40.680	55.704	64.926	72.401	79.396	85.	. 185
class	4.225	5.285	7.573	7.577	7.584	7.592	7.	.792
	8 comps	9 comps	10 comps	11 comps	s 12 comp	ps 13 cc	mps	14 comps
X	89.627	93.557	97.005	98.398	3 99.29	94 99	9.97	99.998
class	7.917	8.022	8.026	8.289	8.84	47 8	3.87	8.872
	15 comps							
X	100.000							
class	9.128							

class



class



[1] 0.05357258

Logistic Regression after Variable Selection

```
Call:
glm(formula = class ~ seismic + shift + gpuls + nbumps, family = binomial,
   data = seismic.train)
Deviance Residuals:
             1Q
                  Median
                               3Q
                                       Max
-1.6270 -0.3846 -0.2947 -0.1627
                                    2.9781
Coefficients:
             Estimate Std. Error z value Pr(>|z|)
(Intercept) -5.9508244 0.6490468 -9.169 < 2e-16 ***
seismic
            0.3641160 0.1944250
                                   1.873 0.061098 .
                                   3.342 0.000832 ***
shift
            1.1371057 0.3402674
            0.0004913
                       0.0001283
                                   3.829 0.000129 ***
gpuls
nbumps
            0.3231048 0.0507286
                                   6.369 1.9e-10 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 958.82 on 1937 degrees of freedom
Residual deviance: 828.98 on 1933 degrees of freedom
```

AIC: 838.98

Number of Fisher Scoring iterations: 6

[1] 0.9318885

glm.pred 0 1 0 1803 128 1 4 3

[1] 0.02290076

[1] 0.9977864

[1] 0.9380805

glm.pred 0 1 0 606 39 1 1 0

[1] 0

[1] 0.9983526

