

Homework 3

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Problem Set 13, Applications

1

(a)

```
vehdata <- read.csv("vehicle.csv")  
summary(vehdata)
```

```

## Compactness      Circularity      Distance.Circularity      Radius.Ratio
## Min.   : 73.00    Min.   :33.00    Min.   : 40.00    Min.   :104.0
## 1st Qu.: 87.00    1st Qu.:40.00    1st Qu.: 70.00    1st Qu.:141.0
## Median : 93.00    Median :44.00    Median : 80.00    Median :167.0
## Mean   : 93.68    Mean   :44.86    Mean   : 82.09    Mean   :168.9
## 3rd Qu.:100.00    3rd Qu.:49.00    3rd Qu.: 98.00    3rd Qu.:195.0
## Max.   :119.00    Max.   :59.00    Max.   :112.00    Max.   :333.0
## Pr.Axis.Aspect.Ratio Max.Length.Aspect.Ratio Scatter.Ratio Elongatedness
## Min.   : 47.00    Min.   : 2.000    Min.   :112.0    Min.   :26.00
## 1st Qu.: 57.00    1st Qu.: 7.000    1st Qu.:146.2    1st Qu.:33.00
## Median : 61.00    Median : 8.000    Median :157.0    Median :43.00
## Mean   : 61.69    Mean   : 8.567    Mean   :168.8    Mean   :40.93
## 3rd Qu.: 65.00    3rd Qu.:10.000    3rd Qu.:198.0    3rd Qu.:46.00
## Max.   :138.00    Max.   :55.000    Max.   :265.0    Max.   :61.00
## Pr.Axis.Rectangularity Max.Length.Rectangularity
## Min.   :17.00    Min.   :118
## 1st Qu.:19.00    1st Qu.:137
## Median :20.00    Median :146
## Mean   :20.58    Mean   :148
## 3rd Qu.:23.00    3rd Qu.:159
## Max.   :29.00    Max.   :188
## Scaled.Variance.Along.Major.Axis Scaled.Variance.Along.Minor.Axis
## Min.   :130.0    Min.   : 184.0
## 1st Qu.:167.0    1st Qu.: 318.2
## Median :178.5    Median : 364.0
## Mean   :188.6    Mean   : 439.9
## 3rd Qu.:217.0    3rd Qu.: 587.0
## Max.   :320.0    Max.   :1018.0
## Scaled.Radius.of.Gyration Skewness.About.Major.Axis Skewness.About.Minor.Axis
## Min.   :109.0    Min.   : 59.00    Min.   : 0.000
## 1st Qu.:149.0    1st Qu.: 67.00    1st Qu.: 2.000
## Median :173.0    Median : 71.50    Median : 6.000
## Mean   :174.7    Mean   : 72.46    Mean   : 6.377
## 3rd Qu.:198.0    3rd Qu.: 75.00    3rd Qu.: 9.000
## Max.   :268.0    Max.   :135.00    Max.   :22.000
## Kurtosis.About.Minor.Axis Kurtosis.About.Major.Axis Hollows.Ratio
## Min.   : 0.0    Min.   :176.0    Min.   :181.0
## 1st Qu.: 5.0    1st Qu.:184.0    1st Qu.:190.2
## Median :11.0    Median :188.0    Median :197.0
## Mean   :12.6    Mean   :188.9    Mean   :195.6
## 3rd Qu.:19.0    3rd Qu.:193.0    3rd Qu.:201.0
## Max.   :41.0    Max.   :206.0    Max.   :211.0
## class
## Min.   :1.000
## 1st Qu.:1.250
## Median :2.000
## Mean   :2.478
## 3rd Qu.:3.000
## Max.   :4.000

```

(b)

```
vehdata$class = factor(vehdata$class, labels=c('2D', '4D', 'BUS', 'VAN'))  
summary(vehdata$class)
```

```
##  2D  4D BUS VAN  
## 212 217 218 199
```

(c)

```
cor(vehdata[,1:18])
```

##	Compactness	Circularity	Distance.Circularity
## Compactness	1.00000000	0.69286923	0.7924440
## Circularity	0.69286923	1.00000000	0.7984920
## Distance.Circularity	0.79244402	0.79849200	1.0000000
## Radius.Ratio	0.69165856	0.62277837	0.7716439
## Pr.Axis.Aspect.Ratio	0.09322213	0.14969187	0.1615292
## Max.Length.Aspect.Ratio	0.14824919	0.24746673	0.2643086
## Scatter.Ratio	0.81300326	0.86036714	0.9072801
## Elongatedness	-0.78864702	-0.82875480	-0.9123072
## Pr.Axis.Rectangularity	0.81343702	0.85792532	0.8953261
## Max.Length.Rectangularity	0.67614317	0.96577578	0.7745238
## Scaled.Variance.Along.Major.Axis	0.76441546	0.80849631	0.8644323
## Scaled.Variance.Along.Minor.Axis	0.81863161	0.85267941	0.8896611
## Scaled.Radius.of.Gyration	0.58534709	0.93608041	0.7058181
## Skewness.About.Major.Axis	-0.25298400	0.05866929	-0.2316598
## Skewness.About.Minor.Axis	0.23369282	0.14843285	0.1164777
## Kurtosis.About.Minor.Axis	0.15677928	-0.01548177	0.2645222
## Kurtosis.About.Major.Axis	0.29778044	-0.11304723	0.1479800
## Hollows.Ratio	0.36555185	0.03867702	0.3354525
##	Radius.Ratio	Pr.Axis.Aspect.Ratio	
## Compactness	0.69165856	0.09322213	
## Circularity	0.62277837	0.14969187	
## Distance.Circularity	0.77164394	0.16152916	
## Radius.Ratio	1.00000000	0.66540656	
## Pr.Axis.Aspect.Ratio	0.66540656	1.00000000	
## Max.Length.Aspect.Ratio	0.44804838	0.64809643	
## Scatter.Ratio	0.73846048	0.10606364	
## Elongatedness	-0.79255741	-0.18508583	
## Pr.Axis.Rectangularity	0.71149706	0.07998168	
## Max.Length.Rectangularity	0.57015440	0.12916829	
## Scaled.Variance.Along.Major.Axis	0.79758793	0.27470468	
## Scaled.Variance.Along.Minor.Axis	0.72578142	0.09218201	
## Scaled.Radius.of.Gyration	0.53906903	0.12440783	
## Skewness.About.Major.Axis	-0.18241222	0.15230355	
## Skewness.About.Minor.Axis	0.05053425	-0.05680698	
## Kurtosis.About.Minor.Axis	0.17429937	-0.03417976	
## Kurtosis.About.Major.Axis	0.38075719	0.23819525	
## Hollows.Ratio	0.46927820	0.26738123	
##	Max.Length.Aspect.Ratio	Scatter.Ratio	
## Compactness	0.14824919	0.813003257	
## Circularity	0.24746673	0.860367138	
## Distance.Circularity	0.26430861	0.907280057	
## Radius.Ratio	0.44804838	0.738460476	
## Pr.Axis.Aspect.Ratio	0.64809643	0.106063643	
## Max.Length.Aspect.Ratio	1.00000000	0.166769266	
## Scatter.Ratio	0.16676927	1.000000000	
## Elongatedness	-0.18004430	-0.973385344	
## Pr.Axis.Rectangularity	0.16186089	0.992088347	
## Max.Length.Rectangularity	0.30594269	0.810647613	
## Scaled.Variance.Along.Major.Axis	0.31942961	0.951862122	
## Scaled.Variance.Along.Minor.Axis	0.14460641	0.996318013	
## Scaled.Radius.of.Gyration	0.18979054	0.800931523	

## Skewness.About.Major.Axis	0.29454772	-0.028618420
## Skewness.About.Minor.Axis	0.01521789	0.074347173
## Kurtosis.About.Minor.Axis	0.04337924	0.210706455
## Kurtosis.About.Major.Axis	-0.02731663	0.004053166
## Hollows.Ratio	0.14391873	0.119949827
##	Elongatedness	Pr.Axis.Rectangularity
## Compactness	-0.78864702	0.81343702
## Circularity	-0.82875480	0.85792532
## Distance.Circularity	-0.91230719	0.89532606
## Radius.Ratio	-0.79255741	0.71149706
## Pr.Axis.Aspect.Ratio	-0.18508583	0.07998168
## Max.Length.Aspect.Ratio	-0.18004430	0.16186089
## Scatter.Ratio	-0.97338534	0.99208835
## Elongatedness	1.00000000	-0.95051244
## Pr.Axis.Rectangularity	-0.95051244	1.00000000
## Max.Length.Rectangularity	-0.77553091	0.81330473
## Scaled.Variance.Along.Major.Axis	-0.93839190	0.93826639
## Scaled.Variance.Along.Minor.Axis	-0.95652176	0.99234619
## Scaled.Radius.of.Gyration	-0.76614632	0.79828199
## Skewness.About.Major.Axis	0.10487465	-0.01633331
## Skewness.About.Minor.Axis	-0.05334635	0.08234563
## Kurtosis.About.Minor.Axis	-0.18518873	0.21308767
## Kurtosis.About.Major.Axis	-0.11204551	-0.02205601
## Hollows.Ratio	-0.21672508	0.09851910
##	Max.Length.Rectangularity	
## Compactness	0.676143173	
## Circularity	0.965775776	
## Distance.Circularity	0.774523840	
## Radius.Ratio	0.570154405	
## Pr.Axis.Aspect.Ratio	0.129168289	
## Max.Length.Aspect.Ratio	0.305942689	
## Scatter.Ratio	0.810647613	
## Elongatedness	-0.775530913	
## Pr.Axis.Rectangularity	0.813304735	
## Max.Length.Rectangularity	1.000000000	
## Scaled.Variance.Along.Major.Axis	0.747151993	
## Scaled.Variance.Along.Minor.Axis	0.798071495	
## Scaled.Radius.of.Gyration	0.866478540	
## Skewness.About.Major.Axis	0.040339043	
## Skewness.About.Minor.Axis	0.137473338	
## Kurtosis.About.Minor.Axis	0.001183035	
## Kurtosis.About.Major.Axis	-0.107718191	
## Hollows.Ratio	0.076769617	
##	Scaled.Variance.Along.Major.Axis	
## Compactness	0.76441546	
## Circularity	0.80849631	
## Distance.Circularity	0.86443228	
## Radius.Ratio	0.79758793	
## Pr.Axis.Aspect.Ratio	0.27470468	
## Max.Length.Aspect.Ratio	0.31942961	
## Scatter.Ratio	0.95186212	
## Elongatedness	-0.93839190	

```

## Pr.Axis.Rectangularity          0.93826639
## Max.Length.Rectangularity        0.74715199
## Scaled.Variance.Along.Major.Axis 1.00000000
## Scaled.Variance.Along.Minor.Axis 0.94988787
## Scaled.Radius.of.Gyration        0.78096248
## Skewness.About.Major.Axis        0.11129593
## Skewness.About.Minor.Axis        0.03913734
## Kurtosis.About.Minor.Axis        0.19194816
## Kurtosis.About.Major.Axis        0.01118902
## Hollows.Ratio                    0.08553952
##                                Scaled.Variance.Along.Minor.Axis
## Compactness                      0.818631612
## Circularity                       0.852679411
## Distance.Circularity              0.889661065
## Radius.Ratio                      0.725781417
## Pr.Axis.Aspect.Ratio              0.092182007
## Max.Length.Aspect.Ratio          0.144606405
## Scatter.Ratio                     0.996318013
## Elongatedness                     -0.956521755
## Pr.Axis.Rectangularity            0.992346188
## Max.Length.Rectangularity         0.798071495
## Scaled.Variance.Along.Major.Axis  0.949887874
## Scaled.Variance.Along.Minor.Axis  1.000000000
## Scaled.Radius.of.Gyration         0.797539612
## Skewness.About.Major.Axis         -0.019872473
## Skewness.About.Minor.Axis         0.076435424
## Kurtosis.About.Minor.Axis         0.202893719
## Kurtosis.About.Major.Axis         0.005379651
## Hollows.Ratio                     0.105432191
##                                Scaled.Radius.of.Gyration
## Compactness                       0.58534709
## Circularity                       0.93608041
## Distance.Circularity              0.70581805
## Radius.Ratio                      0.53906903
## Pr.Axis.Aspect.Ratio              0.12440783
## Max.Length.Aspect.Ratio           0.18979054
## Scatter.Ratio                     0.80093152
## Elongatedness                     -0.76614632
## Pr.Axis.Rectangularity            0.79828199
## Max.Length.Rectangularity         0.86647854
## Scaled.Variance.Along.Major.Axis  0.78096248
## Scaled.Variance.Along.Minor.Axis  0.79753961
## Scaled.Radius.of.Gyration         1.000000000
## Skewness.About.Major.Axis         0.19120943
## Skewness.About.Minor.Axis         0.16800281
## Kurtosis.About.Minor.Axis         -0.05621812
## Kurtosis.About.Major.Axis         -0.22736551
## Hollows.Ratio                     -0.11780842
##                                Skewness.About.Major.Axis
## Compactness                       -0.25298400
## Circularity                       0.05866929
## Distance.Circularity              -0.23165977

```

```

## Radius.Ratio -0.18241222
## Pr.Axis.Aspect.Ratio 0.15230355
## Max.Length.Aspect.Ratio 0.29454772
## Scatter.Ratio -0.02861842
## Elongatedness 0.10487465
## Pr.Axis.Rectangularity -0.01633331
## Max.Length.Rectangularity 0.04033904
## Scaled.Variance.Along.Major.Axis 0.11129593
## Scaled.Variance.Along.Minor.Axis -0.01987247
## Scaled.Radius.of.Gyration 0.19120943
## Skewness.About.Major.Axis 1.00000000
## Skewness.About.Minor.Axis -0.09253891
## Kurtosis.About.Minor.Axis -0.12615735
## Kurtosis.About.Major.Axis -0.75414191
## Hollows.Ratio -0.80539854
## Skewness.About.Minor.Axis
## Compactness 0.23369282
## Circularity 0.14843285
## Distance.Circularity 0.11647771
## Radius.Ratio 0.05053425
## Pr.Axis.Aspect.Ratio -0.05680698
## Max.Length.Aspect.Ratio 0.01521789
## Scatter.Ratio 0.07434717
## Elongatedness -0.05334635
## Pr.Axis.Rectangularity 0.08234563
## Max.Length.Rectangularity 0.13747334
## Scaled.Variance.Along.Major.Axis 0.03913734
## Scaled.Variance.Along.Minor.Axis 0.07643542
## Scaled.Radius.of.Gyration 0.16800281
## Skewness.About.Major.Axis -0.09253891
## Skewness.About.Minor.Axis 1.00000000
## Kurtosis.About.Minor.Axis -0.03936531
## Kurtosis.About.Major.Axis 0.11693185
## Hollows.Ratio 0.09914923
## Kurtosis.About.Minor.Axis
## Compactness 0.156779278
## Circularity -0.015481769
## Distance.Circularity 0.264522153
## Radius.Ratio 0.174299370
## Pr.Axis.Aspect.Ratio -0.034179762
## Max.Length.Aspect.Ratio 0.043379243
## Scatter.Ratio 0.210706455
## Elongatedness -0.185188734
## Pr.Axis.Rectangularity 0.213087671
## Max.Length.Rectangularity 0.001183035
## Scaled.Variance.Along.Major.Axis 0.191948164
## Scaled.Variance.Along.Minor.Axis 0.202893719
## Scaled.Radius.of.Gyration -0.056218120
## Skewness.About.Major.Axis -0.126157345
## Skewness.About.Minor.Axis -0.039365312
## Kurtosis.About.Minor.Axis 1.000000000
## Kurtosis.About.Major.Axis 0.079089910

```

## Hollows.Ratio	0.204923603	
##	Kurtosis.About.Major.Axis	Hollows.Ratio
## Compactness	0.297780441	0.36555185
## Circularity	-0.113047232	0.03867702
## Distance.Circularity	0.147979965	0.33545253
## Radius.Ratio	0.380757189	0.46927820
## Pr.Axis.Aspect.Ratio	0.238195253	0.26738123
## Max.Length.Aspect.Ratio	-0.027316627	0.14391873
## Scatter.Ratio	0.004053166	0.11994983
## Elongatedness	-0.112045506	-0.21672508
## Pr.Axis.Rectangularity	-0.022056010	0.09851910
## Max.Length.Rectangularity	-0.107718191	0.07676962
## Scaled.Variance.Along.Major.Axis	0.011189021	0.08553952
## Scaled.Variance.Along.Minor.Axis	0.005379651	0.10543219
## Scaled.Radius.of.Gyration	-0.227365512	-0.11780842
## Skewness.About.Major.Axis	-0.754141913	-0.80539854
## Skewness.About.Minor.Axis	0.116931854	0.09914923
## Kurtosis.About.Minor.Axis	0.079089910	0.20492360
## Kurtosis.About.Major.Axis	1.000000000	0.89409781
## Hollows.Ratio	0.894097812	1.00000000

The following variables have strong correlations (beyond ± 0.7):

Compactness and Distance.Circularity

Compactness and Scatter.Ratio

Compactness and Elongatedness

Compactness and Pr.Axis.Rectangularity

Compactness and Scaled.Variance.Along.Major.Axis

Compactness and Scaled.Variance.Along.Minor.Axis

Circularity and Distance.Circularity

Circularity and Scatter.Ratio

Circularity and Elongatedness

Circularity and Pr.Axis.Rectangularity

Circularity and Scaled.Variance.Along.Major.Axis

Circularity and Scaled.Variance.Along.Minor.Axis

Distance.Circularity and Radius.Ratio

Distance.Circularity and Pr.Axis.Rectangularity

Distance.Circularity and Max.Length.Rectangularity

Distance.Circularity and Scaled.Variance.Along.Major.Axis

Distance.Circularity and Scaled.Variance.Along.Minor.Axis

Distance.Circularity and Scaled.Radius.of.Gyration

Radius.Ratio and Pr.Axis.Aspect.Ratio

Radius.Ratio and Scatter.Ratio

Radius.Ratio and Elongatedness

Radius.Ratio and Pr.Axis.Rectangularity

Radius.Ratio and Scaled.Variance.Along.Major.Axis

Radius.Ratio and Scaled.Variance.Along.Minor.Axis

Scatter.Ratio and Max.Length.Rectangularity

Scatter.Ratio and Scaled.Radius.of.Gyration

Elongatedness and Max.Length.Rectangularity

Elongatedness and Scaled.Radius.of.Gyration

Pr.Axis.Rectangularity and Max.Length.Rectangularity
Pr.Axis.Rectangularity and Scaled.Radius.of.Gyration
Max.Length.Rectangularity and Scaled.Variance.Along.Major.Axis
Max.Length.Rectangularity and Scaled.Variance.Along.Minor.Axis
Max.Length.Rectangularity and Scaled.Radius.of.Gyration
Scaled.Variance.Along.Major.Axis and Scaled.Radius.of.Gyration
Scaled.Variance.Along.Minor.Axis and Scaled.Radius.of.Gyration
Skewness.About.Major.Axis and Kurtosis.About.Major.Axis
Skewness.About.Major.Axis and Hollows.Ratio
Kurtosis.About.Major.Axis and Hollows.Ratio

The following variables have especially strong correlations (beyond ± 0.9):

Circularity and Max.Length.Rectangularity
Circularity and Scaled.Radius.of.Gyration
Distance.Circularity and Scatter.Ratio
Distance.Circularity and Elongatedness
Scatter.Ratio and Elongatedness
Scatter.Ratio and Pr.Axis.Rectangularity
Scatter.Ratio and Scaled.Variance.Along.Major.Axis
Scatter.Ratio and Scaled.Variance.Along.Minor.Axis
Elongatedness and Pr.Axis.Rectangularity
Elongatedness and Scaled.Variance.Along.Major.Axis
Elongatedness and Scaled.Variance.Along.Minor.Axis
Pr.Axis.Rectangularity and Scaled.Variance.Along.Major.Axis
Pr.Axis.Rectangularity and Scaled.Variance.Along.Minor.Axis
Scaled.Variance.Along.Major.Axis and Scaled.Variance.Along.Minor.Axis

2

```
set.seed(46685326, kind = "Mersenne-Twister")
perm <- sample(x = nrow(vehdata))
set1 <- vehdata[which(perm <= 3*nrow(vehdata)/4), ]
set2 <- vehdata[which(perm > 3*nrow(vehdata)/4), ]
head(set1)
```

```
## Compactness Circularity Distance.Circularity Radius.Ratio
## 1          95          48          83          178
## 2          91          41          84          141
## 3         104          50         106          209
## 4          93          41          82          159
## 5          85          44          70          205
## 7          97          43          73          173
## Pr.Axis.Aspect.Ratio Max.Length.Aspect.Ratio Scatter.Ratio Elongatedness
## 1              72              10              162              42
## 2              57              9              149              45
## 3              66              10              207              32
## 4              63              9              144              46
## 5             103              52              149              45
## 7              65              6              153              42
## Pr.Axis.Rectangularity Max.Length.Rectangularity
## 1              20              159
## 2              19              143
## 3              23              158
## 4              19              143
## 5              19              144
## 7              19              143
## Scaled.Variance.Along.Major.Axis Scaled.Variance.Along.Minor.Axis
## 1              176              379
## 2              170              330
## 3              223              635
## 4              160              309
## 5              241              325
## 7              176              361
## Scaled.Radius.of.Gyration Skewness.About.Major.Axis Skewness.About.Minor.Axis
## 1              184              70              6
## 2              158              72              9
## 3              220              73              14
## 4              127              63              6
## 5              188              127              9
## 7              172              66              13
## Kurtosis.About.Minor.Axis Kurtosis.About.Major.Axis Hollows.Ratio class
## 1              16              187              197 VAN
## 2              14              189              199 VAN
## 3              9              188              196 4D
## 4              10              199              207 VAN
## 5              11              180              183 BUS
## 7              1              200              204 BUS
```

```
head(set2)
```

```

##      Compactness Circularity Distance.Circularity Radius.Ratio
## 6          107          57          106          172
## 11          86          36          70          143
## 12          90          34          66          136
## 14          89          42          85          144
## 15          94          49          79          203
## 16          96          55          103          201
##      Pr.Axis.Aspect.Ratio Max.Length.Aspect.Ratio Scatter.Ratio Elongatedness
## 6              50              6              255              26
## 11              61              9              133              50
## 12              55              6              123              54
## 14              58              10             152              44
## 15              71              5              174              37
## 16              65              9              204              32
##      Pr.Axis.Rectangularity Max.Length.Rectangularity
## 6              28              169
## 11              18              130
## 12              17              118
## 14              19              144
## 15              21              154
## 16              23              166
##      Scaled.Variance.Along.Major.Axis Scaled.Variance.Along.Minor.Axis
## 6              280              957
## 11              153              266
## 12              148              224
## 14              173              345
## 15              196              465
## 16              227              624
##      Scaled.Radius.of.Gyration Skewness.About.Major.Axis
## 6              264              85
## 11              127              66
## 12              118              65
## 14              161              72
## 15              206              71
## 16              246              74
##      Skewness.About.Minor.Axis Kurtosis.About.Minor.Axis
## 6              5              9
## 11              2              10
## 12              5              26
## 14              8              13
## 15              6              2
## 16              6              2
##      Kurtosis.About.Major.Axis Hollows.Ratio class
## 6              181              183    BUS
## 11              194              202    VAN
## 12              196              202     4D
## 14              187              197    VAN
## 15              197              199    BUS
## 16              186              194     2D

```

3

```
library(FNN)

### Split up the predictor variables from the class labels.
X.train.raw = set1[, -19]
X.valid.raw = set2[, -19]
Y.train = set1[, 19]
Y.valid = set2[, 19]

### KNN is based on distances. If variables are measured on different
### scales, we can change which points are neighbours by measuring
### in different units.
### a function we can use to rescale the columns of
### a data frame to have mean 0 and SD 1. We can also use it to rescale
### a data frame based on the means and SDs of another (this is useful
### for scaling the validation set to match the training set).

### Rescale x1 using the means and SDs of x2
scale.1 <- function(x1, x2) {
  for (col in 1:ncol(x1)) {
    a <- mean(x2[, col])
    b <- sd(x2[, col])
    x1[, col] <- (x1[, col] - a) / b
  }
  x1
}

### Rescale our training and validation sets
X.train <- scale.1(X.train.raw, X.train.raw)
X.valid <- scale.1(X.valid.raw, X.train.raw) # Watch the order

### Now we can fit a KNN model using the knn function in the FNN
### package. The syntax of the knn function is a bit different from
### what we're used to. The first two inputs are the training and
### validation predictor matrices. The third input is the class labels
### for the training set. We can also set k to the number of
### neighbours we want. The function then outputs predicted class
### labels for the validation set. Let's use 1 neighbour.
pred.knn <- knn(X.train, X.valid, Y.train, k = 1)
```

(a)

```
### Let's make a confusion matrix. We get this using the table()
### function and providing both the predicted and true class labels
### for the validation set. We can also set the axis labels using
### the dnn input.
table(pred.knn, Y.valid, dnn = c("Predicted", "Observed"))
```

```
##           Observed
## Predicted 2D 4D BUS VAN
##           2D 25 18  1  0
##           4D 26 29  1  2
##           BUS  0  1 46  6
##           VAN  4  6  1 46
```

There are 55 vehicles of type 2D, 54 vehicles of type 4D, 59 vehicles of type BUS and 54 vehicles of type VAN. In particular, it seems to be difficult to distinguish between 2D and 4D vehicles, whereas BUS and VAN vehicles are easy to predict.

(b)

```
### Next, Let's get the misclassification rate
(misclass.knn <- mean(pred.knn != Y.valid))
```

```
## [1] 0.3113208
```

```
(se.knn <- sapply(misclass.knn, function(r) {
  sqrt(r * (1 - r) / nrow(X.train))
}))
```

```
## [1] 0.01838941
```

The test misclassification rate is about 31%, and the standard error is approximately $\pm 1.8\%$

Problem Set 14, Applications

1

(a)

```
### Rescale the columns of x1 so that the columns of x2 fall between 0 and 1
rescale <- function(x1, x2) {
  for (col in 1:ncol(x1)) {
    a <- min(x2[, col])
    b <- max(x2[, col])
    x1[, col] <- (x1[, col] - a) / (b - a)
  }
  x1
}

### Create copies of our datasets and rescale
data.train.scale <- set1
data.valid.scale <- set2
data.train.scale[, -19] <- rescale(data.train.scale[, -19], set1[, -19])
data.valid.scale[, -19] <- rescale(data.valid.scale[, -19], set1[, -19])
summary(data.train.scale[,1:3])
```

```
## Compactness      Circularity      Distance.Circularity
## Min.      :0.0000   Min.      :0.0000   Min.      :0.0000
## 1st Qu.:0.3261   1st Qu.:0.2692   1st Qu.:0.4286
## Median :0.4348   Median :0.4231   Median :0.5714
## Mean    :0.4483   Mean    :0.4512   Mean     :0.5973
## 3rd Qu.:0.5652   3rd Qu.:0.6154   3rd Qu.:0.8000
## Max.    :1.0000   Max.    :1.0000   Max.     :1.0000
```

```
summary(data.train.scale[,1:3])
```

```
## Compactness      Circularity      Distance.Circularity
## Min.      :0.0000   Min.      :0.0000   Min.      :0.0000
## 1st Qu.:0.3261   1st Qu.:0.2692   1st Qu.:0.4286
## Median :0.4348   Median :0.4231   Median :0.5714
## Mean    :0.4483   Mean    :0.4512   Mean     :0.5973
## 3rd Qu.:0.5652   3rd Qu.:0.6154   3rd Qu.:0.8000
## Max.    :1.0000   Max.    :1.0000   Max.     :1.0000
```

It appears that the data has been correctly scaled, to lie between 0 and 1.

(b)

(i)

```
library(nnet)
library(car)
```

```
## Loading required package: carData
```

```
library(glmnet)
```

```
## Loading required package: Matrix
```

```
## Loaded glmnet 4.1-8
```

```
library(MASS)
```

```
fit.log.nnet <- multinom(class ~ ., data = data.train.scale)
```

```
## # weights:  80 (57 variable)
## initial  value 878.910625
## iter   10 value 540.126072
## iter   20 value 264.002381
## iter   30 value 222.740935
## iter   40 value 211.568230
## iter   50 value 203.944730
## iter   60 value 200.943742
## iter   70 value 199.287545
## iter   80 value 197.947131
## iter   90 value 196.809842
## iter  100 value 196.152373
## final   value 196.152373
## stopped after 100 iterations
```

```
Anova(fit.log.nnet)
```

```
## Analysis of Deviance Table (Type II tests)
##
## Response: class
##
##          LR Chisq Df Pr(>Chisq)
## Compactness      55.741  3  4.770e-12 ***
## Circularity       18.992  3  0.0002744 ***
## Distance.Circularity 19.143  3  0.0002554 ***
## Radius.Ratio     131.680  3  < 2.2e-16 ***
## Pr.Axis.Aspect.Ratio 141.223  3  < 2.2e-16 ***
## Max.Length.Aspect.Ratio 15.429  3  0.0014845 **
## Scatter.Ratio      2.759  3  0.4302978
## Elongatedness      4.330  3  0.2279767
## Pr.Axis.Rectangularity 8.232  3  0.0414498 *
## Max.Length.Rectangularity 28.668  3  2.630e-06 ***
## Scaled.Variance.Along.Major.Axis 31.655  3  6.186e-07 ***
## Scaled.Variance.Along.Minor.Axis 1.058  3  0.7871357
## Scaled.Radius.of.Gyration 26.215  3  8.599e-06 ***
## Skewness.About.Major.Axis 28.790  3  2.479e-06 ***
## Skewness.About.Minor.Axis 13.571  3  0.0035514 **
## Kurtosis.About.Minor.Axis 8.339  3  0.0395014 *
## Kurtosis.About.Major.Axis 99.194  3  < 2.2e-16 ***
## Hollows.Ratio     61.632  3  2.634e-13 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

According to Anova, the following variables are important (at 95% confidence level):

Compactness

Circularity

Distance.Circularity

Radius.Ratio

Pr.Axis.Aspect.Ratio

Max.Length.Aspect.Ratio

Max.Length.Rectangularity

Scaled.Variance.Along.Major.Axis

Scaled.Radius.of.Gyration

Skewness.About.Major.Axis

Skewness.About.Minor.Axis

Kurtosis.About.Minor.Axis

Kurtosis.About.Major.Axis

Hollows.Ratio

So that's 14 out of 18 variables. The other 4 are relatively unimportant.

(ii)

```
pred.log.nnet <- predict(fit.log.nnet, data.valid.scale)
(misclass.log.nnet <- mean(pred.log.nnet != Y.valid))
```

```
## [1] 0.2122642
```



```
(misclass.log.se <- sapply(misclass.log.nnet, function(r) {
  sqrt(r * (1 - r) / nrow(X.train))
}))
```

```
## [1] 0.01623992
```

The misclassification rate is still just over 21%. The standard error for the multinomial classification is about 1.6%, slightly lower than KNN but not by much.

(iii)

```
table(Y.valid, pred.log.nnet, ### Confusion matrix
      dnn = c("Observed", "Predicted"))
)
```

```
##          Predicted
## Observed 2D 4D BUS VAN
##      2D  34 19   1   1
##      4D  18 33   1   2
##      BUS   1  1 47   0
##      VAN   1  0  0 53
```

```
misclass.log.nnet
```

```
## [1] 0.2122642
```

Note that the misclassification rate is lower (just above 21%). We still observe that 2D and 4D vehicles are hard to distinguish and that BUS and VAN are much easier to distinguish.

2

(a)

```
### The glmnet() function uses predictor matrix/response vector syntax,
### so we need to extract these from our training and validation sets.
### We also have to convert the predictors to a matrix using the
### as.matrix() function.
X.train.scale <- as.matrix(data.train.scale[, -19])
Y.train <- data.train.scale[, 19]
X.valid.scale <- as.matrix(data.valid.scale[, -19])
Y.valid <- data.valid.scale[, 19]

### While we're looking at the glmnet package, Let's do LASSO. We need to
### choose lambda using CV. Fortunately, the cv.glmnet() function does this
### for us. The syntax for cv.glmnet() is the same as for glmnet().
fit.CV.lasso <- cv.glmnet(X.train.scale, Y.train, family = "multinomial")

### The CV-min values are stored in the output from
### cv.glmnet()
lambda.min <- fit.CV.lasso$lambda.min

### Let's check which predictors are included in each "best" model. We
### can get the coefficients using the coef() function, setting s to
### the appropriate lambda value.
coef(fit.CV.lasso, s = lambda.min)
```

```

## $`2D`
## 19 x 1 sparse Matrix of class "dgCMatrix"
##                                     1
## (Intercept)                      10.21104200
## Compactness                      -6.45355109
## Circularity                      4.17746021
## Distance.Circularity             -0.26588534
## Radius.Ratio                    17.96378752
## Pr.Axis.Aspect.Ratio             -27.99934280
## Max.Length.Aspect.Ratio          -3.63129111
## Scatter.Ratio                    0.02509100
## Elongatedness                    3.08830452
## Pr.Axis.Rectangularity           .
## Max.Length.Rectangularity        -7.33835179
## Scaled.Variance.Along.Major.Axis -1.49417719
## Scaled.Variance.Along.Minor.Axis 15.44674226
## Scaled.Radius.of.Gyration        -3.02963705
## Skewness.About.Major.Axis        -14.18581458
## Skewness.About.Minor.Axis         0.05707199
## Kurtosis.About.Minor.Axis         0.20129341
## Kurtosis.About.Major.Axis         0.97398338
## Hollows.Ratio                   -5.38678590
##
## $`4D`
## 19 x 1 sparse Matrix of class "dgCMatrix"
##                                     1
## (Intercept)                      13.1396119
## Compactness                      4.4747208
## Circularity                     -12.2757752
## Distance.Circularity             0.2658853
## Radius.Ratio                    20.5065471
## Pr.Axis.Aspect.Ratio             -27.6999366
## Max.Length.Aspect.Ratio          -9.6494204
## Scatter.Ratio                    .
## Elongatedness                   -3.0883045
## Pr.Axis.Rectangularity           7.8058323
## Max.Length.Rectangularity        -0.8026757
## Scaled.Variance.Along.Major.Axis .
## Scaled.Variance.Along.Minor.Axis -10.3225408
## Scaled.Radius.of.Gyration         6.3002800
## Skewness.About.Major.Axis        -10.5404710
## Skewness.About.Minor.Axis         0.1181783
## Kurtosis.About.Minor.Axis        -0.2012934
## Kurtosis.About.Major.Axis        -12.6231148
## Hollows.Ratio                    5.3867859
##
## $BUS
## 19 x 1 sparse Matrix of class "dgCMatrix"
##                                     1
## (Intercept)                      0.2702513
## Compactness                      -4.4747208
## Circularity                      .

```

```
## Distance.Circularity      -6.2573078
## Radius.Ratio              -62.9143391
## Pr.Axis.Aspect.Ratio      102.7336033
## Max.Length.Aspect.Ratio    3.6312911
## Scatter.Ratio              .
## Elongatedness             -19.3133250
## Pr.Axis.Rectangularity     .
## Max.Length.Rectangularity   0.8026757
## Scaled.Variance.Along.Major.Axis 44.7123659
## Scaled.Variance.Along.Minor.Axis .
## Scaled.Radius.of.Gyration    3.0296370
## Skewness.About.Major.Axis   10.5404710
## Skewness.About.Minor.Axis   -4.9814126
## Kurtosis.About.Minor.Axis   3.0009161
## Kurtosis.About.Major.Axis   46.2391073
## Hollows.Ratio              -24.6221721
##
## $VAN
## 19 x 1 sparse Matrix of class "dgCMatrix"
##                                1
## (Intercept)                   -23.62090512
## Compactness                   12.59825716
## Circularity                    .
## Distance.Circularity           21.05765272
## Radius.Ratio                  -17.96378752
## Pr.Axis.Aspect.Ratio           27.69993664
## Max.Length.Aspect.Ratio        28.02519655
## Scatter.Ratio                  -33.56834420
## Elongatedness                  9.32075772
## Pr.Axis.Rectangularity         -18.23841429
## Max.Length.Rectangularity       36.55491373
## Scaled.Variance.Along.Major.Axis .
## Scaled.Variance.Along.Minor.Axis .
## Scaled.Radius.of.Gyration      -18.54739118
## Skewness.About.Major.Axis       34.48477108
## Skewness.About.Minor.Axis      -0.05707199
## Kurtosis.About.Minor.Axis      -3.14142057
## Kurtosis.About.Major.Axis      -0.97398338
## Hollows.Ratio                  12.97849013
```

```
### Now we can get predictions for both "best" models
pred.lasso.min <- predict(fit.CV.lasso, X.valid.scale,
  s = lambda.min,
  type = "class"
)

table(Y.valid, pred.lasso.min, dnn = c("Obs", "Pred"))
```

```
##      Pred
## Obs   2D 4D BUS VAN
##   2D  33 20   1   1
##   4D  18 34   1   1
##   BUS   0  1  48   0
##   VAN   1  0   2  51
```

For predicting 2D, 17 of the 18 variables are important, except for Pr.Axis.Rectangularity.

For predicting 4D, 16 of the 18 variables are important, except for Scatter.Ratio and

Scaled.Variance.Along.Major.Axis.

For predicting BUS, 14 of the 18 variables are important, except for Circularity, Scatter.Ratio,

Pr.Axis.Rectangularity, and Scaled.Variance.Along.Minor.Axis.

For predicting VAN, 15 of the 18 variables are important, except for Circularity, Scaled.Variance.Along.Major.Axis, and Scaled.Variance.Along.Minor.Axis.

Recall from ANOVA, that Scatter.Ratio, Elongatedness, Pr.Axis.Rectangularity, and

Scaled.Variance.Along.Minor.Axis are unimportant variables, but the LASSO logit model suggests that Circularity would be unimportant for predicting BUS and VAN, and Elongatedness is important in all cases. The other variables generally match with ANOVA.

(b)

```
(miss.lasso.min <- mean(Y.valid != pred.lasso.min))
```

```
## [1] 0.2169811
```

```
(se.miss.lasso.min <- sapply(miss.lasso.min, function(r) {
  sqrt(r * (1 - r) / nrow(X.train.scale))
}))
```

```
## [1] 0.01637014
```

The misclassification rate is now at around 22%. The error is still at 1.6%. No improvement has been made.

3

(a)

```

### For discriminant analysis, it's best to scale predictors
### to have mean 0 and SD 1 (this makes the results easier to
### interpret). We can do this using using the following function.

### Rescale x1 using the means and SDs of x2
scale.1 <- function(x1, x2) {
  for (col in 1:ncol(x1)) {
    a <- mean(x2[, col])
    b <- sd(x2[, col])
    x1[, col] <- (x1[, col] - a) / b
  }
  x1
}

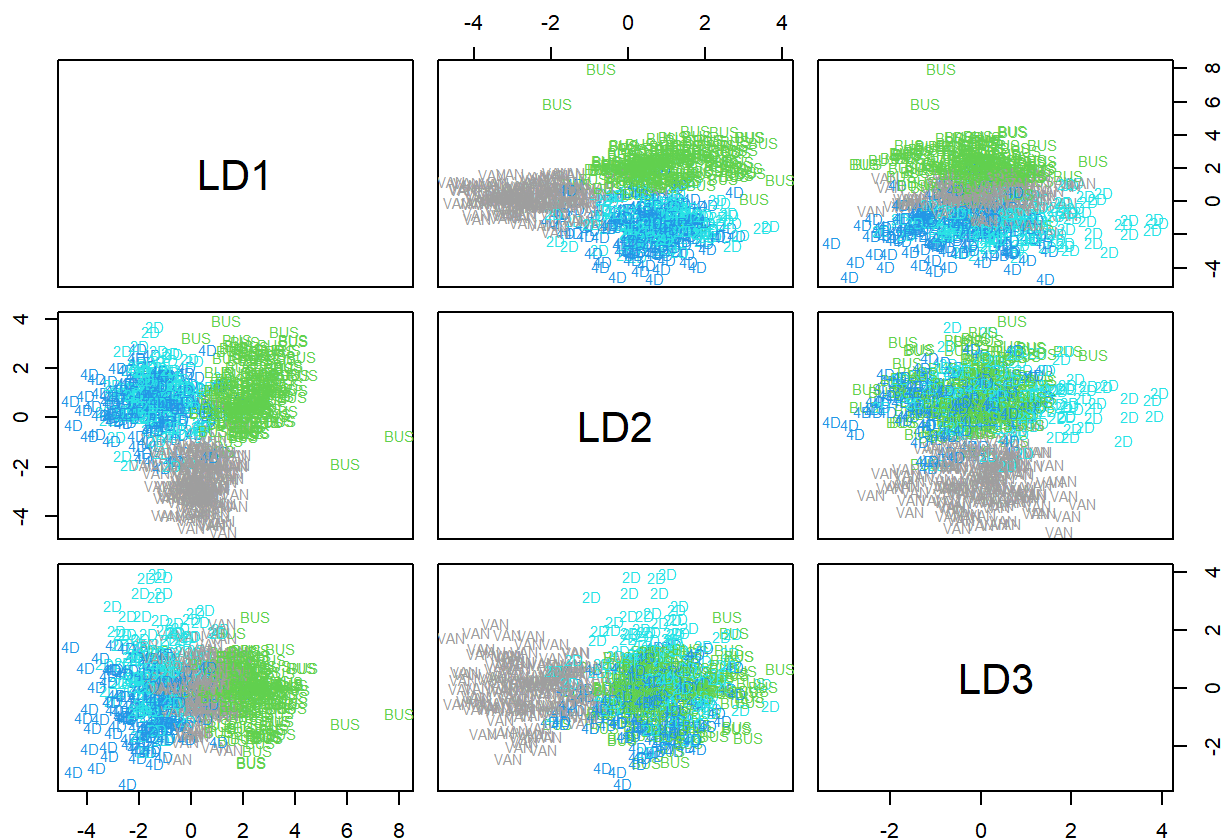
X.train.DA <- scale.1(set1[, -19], set1[, -19])
X.valid.DA <- scale.1(set2[, -19], set1[, -19])

class.col <- ifelse(set1$class=='2D', y = 53,
n = ifelse(set1$class=='4D', y = 68, n = ifelse(set1$class=='BUS',y=203,n=464)))

### Fit an LDA model using the lda() funtion from the MASS package. This
### function uses predictor/response syntax.
fit.lda <- lda(X.train.DA, Y.train)

### We can plot the data using the linear discriminants. It's best to
### include colors. Let's just recycle the colors from above.
### There is no simple way to change the axis labels. Sometimes we just need
### to live with the defaults.
plot(fit.lda, col = class.col)

```



LD1 seems to be separating vehicles of class BUS from the rest.

LD2 seems to be separating vehicles of class 2D and 4D from the rest. It cannot really distinguish between the 2 classes.

LD3 seems to be separating vehicles of class VAN from the rest.

(b)

```
### We get predictions by extracting the class object from the predict()
```

```
### function's output.
```

```
pred.lda <- predict(fit.lda, X.valid.DA)$class
```

```
table(Y.valid, pred.lda, dnn = c("Obs", "Pred"))
```

```
##      Pred
## Obs   2D 4D BUS VAN
##  2D  34 19  1  1
##  4D  15 33  3  3
##  BUS   0  0 49  0
##  VAN   0  0  0 54
```

```
(miss.lda <- mean(Y.valid != pred.lda))
```

```
## [1] 0.1981132
```

```
(se.miss.lda <- sapply(miss.lda, function(r) {  
  sqrt(r * (1 - r) / nrow(X.train.DA))  
}))
```

```
## [1] 0.01582955
```

The misclassification rate is just under 20%. The error is just under 1.6%, so it is a very slight improvement compare to the rest.