

# Homework 3

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## 1. Problem Set 13, Applications

```
library(FNN)
```

```
## Warning: package 'FNN' was built under R version 4.1.2
```

### 1(a).

```
# read the data  
vehicle <- read.csv("vehicle.csv")  
# print summary  
summary(vehicle)
```

```

## 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

```

## 1(b).

```
# convert class to factor
vehicle$class <- factor(vehicle$class, labels=c("2D", "4D", "Bus", "Van"))
# print summary
summary(vehicle)
```

```
## 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 class
## Min.   : 0.0    Min.   :176.0    Min.   :181.0    2D :212
## 1st Qu.: 5.0    1st Qu.:184.0    1st Qu.:190.2    4D :217
## Median :11.0    Median :188.0    Median :197.0    Bus:218
## Mean   :12.6    Mean   :188.9    Mean   :195.6    Van:199
## 3rd Qu.:19.0    3rd Qu.:193.0    3rd Qu.:201.0
## Max.   :41.0    Max.   :206.0    Max.   :211.0
```

**1(c).**

```
corMatrix <- cor(vehicle[, -19])  
corMatrix
```

##	Compactness	Circularity	Distance.Circularity
##	Compactness	1.00000000	0.69286923
##	Circularity	0.69286923	1.00000000
##	Distance.Circularity	0.79244402	0.79849200
##	Radius.Ratio	0.69165856	0.62277837
##	Pr.Axis.Aspect.Ratio	0.09322213	0.14969187
##	Max.Length.Aspect.Ratio	0.14824919	0.24746673
##	Scatter.Ratio	0.81300326	0.86036714
##	Elongatedness	-0.78864702	-0.82875480
##	Pr.Axis.Rectangularity	0.81343702	0.85792532
##	Max.Length.Rectangularity	0.67614317	0.96577578
##	Scaled.Variance.Along.Major.Axis	0.76441546	0.80849631
##	Scaled.Variance.Along.Minor.Axis	0.81863161	0.85267941
##	Scaled.Radius.of.Gyration	0.58534709	0.93608041
##	Skewness.About.Major.Axis	-0.25298400	0.05866929
##	Skewness.About.Minor.Axis	0.23369282	0.14843285
##	Kurtosis.About.Minor.Axis	0.15677928	-0.01548177
##	Kurtosis.About.Major.Axis	0.29778044	-0.11304723
##	Hollows.Ratio	0.36555185	0.03867702
##	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

Variables with correlations between  $\pm 0.7$  and  $\pm 0.9$ :

- Compactness and Circularity is correlated with Distance.Circularity, Scatter.Ratio, Elongatedness, Pr.Axis.Rectangularity, Scaled.Variance.Along.Major.Axis, and Scaled.Variance.Along.Minor.Axis.
- Distance.Circularity is correlated with Radius.Ratio, Pr.Axis.Rectangularity, Max.Length.Rectangularity, Scaled.Variance.Along.Major.Axis, Scaled.Variance.Along.Minor.Axis, and Scaled.Radius.of.Gyration.
- Radius.Ratio is correlated with Scatter.Ratio, Elongatedness, Pr.Axis.Rectangularity, Scaled.Variance.Along.Major.Axis, and Scaled.Variance.Along.Minor.Axis.
- Scatter.Ratio, Elongatedness, and Pr.Axis.Rectangularity is correlated with Max.Length.Rectangularity and Scaled.Radius.of.Gyration.
- Max.Length.Rectangularity, Scaled.Variance.Along.Major.Axis, Scaled.Variance.Along.Minor.Axis, and Scaled.Radius.of.Gyration are correlated with each other.
- Skewness.About.Major.Axis, Kurtosis.About.Major.Axis and Hollows.Ratio are correlated with each other.

Variables with correlations greater than  $\pm 0.9$ :

- Circularity is highly correlated with Max.Length.Rectangularity, and Scaled.Radius.of.Gyration.
- Distance.Circularity is highly correlated with Scatter.Ratio and Elongatedness.
- Scatter.Ratio, Elongatedness, Pr.Axis.Rectangularity, Scaled.Variance.Along.Major.Axis, and Scaled.Variance.Along.Minor.Axis are highly correlated with each other.

## 2.

```
# sample the data into train and test sets
set.seed(46685326, kind="Mersenne-Twister")
perm <- sample(x=nrow(vehicle))
# print set 1
set1 <- vehicle[which(perm <= 3*nrow(vehicle)/4), ]
head(set1, n = 6)
```

##	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

```
# print set 2
set2 <- vehicle[which(perm > 3*nrow(vehicle)/4), ]
head(set2, n = 6)
```

##	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(a).

```
# scale the X variables using mean and sd from training set
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.set1.unscaled <- as.matrix(set1[, -19])
x.set2.unscaled <- as.matrix(set2[, -19])
x.set1 <- scale.1(x.set1.unscaled, x.set1.unscaled)
x.set2 <- scale.1(x.set2.unscaled, x.set1.unscaled)
# fit a knn using m/k= 1
model.knn <- knn(train = x.set1, test = x.set2, cl=set1[, 19], k=1)
# create confusion matrix
table(model.knn, set2[, 19], 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 is a high misclassification rate between classes “2D” and “4D” which indicates that these two classes are harder to separate as these two classes overlap each other. Class “Bus” has the lowest misclassifications which indicates that this class is easier to separate from other classes; however, there is a small overlap with class “Van”. We also see a small number misclassifications in class “Van” which indicates that this class is also easier to separate from other classes; however, there is a small overlap with classes “2D” and “4D”.

### 3(b).

```
# calculate misclassification rate
(misclass.model.knn <-
  mean(ifelse(model.knn == set2[, 19], yes = 0, no = 1)))
```

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

```
# calculate aprox. se of misclassification
(se <- sqrt(misclass.model.knn * (1 - misclass.model.knn) / nrow(set2)))
```

```
## [1] 0.03180128
```

## 2. Problem Set 14, Applications

```
library(nnet)
```

```
## Warning: package 'nnet' was built under R version 4.1.2
```

```
library(car)
```

```
## Warning: package 'car' was built under R version 4.1.2
```

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

```
## Warning: package 'carData' was built under R version 4.1.2
```

```
library(glmnet)
```

```
## Warning: package 'glmnet' was built under R version 4.1.2
```

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

```
## Warning: package 'Matrix' was built under R version 4.1.2
```

```
## Loaded glmnet 4.1-7
```

```
library(MASS)
```

```
## Warning: package 'MASS' was built under R version 4.1.2
```

## 1(a).

```
# scale the values in training set 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
}
set1.rescale <- data.frame(cbind(rescale(set1[, -19], set1[, -19]),
                                class = set1$class))
set2.rescale <- data.frame(cbind(rescale(set2[, -19], set1[, -19]),
                                class = set2$class))

#summary(set1.rescale)
#summary(set2.rescale)
```

```
summary(set1.rescale)[, 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(set2.rescale)[, 1:3]
```

```
## Compactness      Circularity      Distance.Circularity
## Min.      :0.06522   Min.      :0.0000   Min.      :0.1429
## 1st Qu.:0.30435   1st Qu.:0.2692   1st Qu.:0.4286
## Median :0.39130   Median :0.4615   Median :0.5643
## Mean    :0.45324   Mean    :0.4713   Mean    :0.6133
## 3rd Qu.:0.60870   3rd Qu.:0.6923   3rd Qu.:0.8286
## Max.    :0.93478   Max.    :0.9615   Max.    :1.0286
```

## 1(b)

i.

```
# fit logistic regression
mod.logit <- multinom(
  data = set1.rescale, formula = class ~ .,
  trace = TRUE
)
```

```
## # 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
```

```
#summary(mod.logit)
# run anova on the model
Anova(mod.logit)
```



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

```
# compute se for test set
(se.logit <- sqrt(mul.misclass.test * (1 - mul.misclass.test) / nrow(set2)))
```

```
## [1] 0.02808411
```

The test error for logistic regression using 'multinom()' is lower than that of the optimal KNN, suggesting that logistic regression performs better in making predictions for this case.

### iii.

```
# create confusion matrix
table(set2$class, pred.class.2, dnn = c("Obs", "Pred"))
```

```
##      Pred
## Obs   2D 4D Bus Van
## 2D   34 19  1  1
## 4D   18 33  1  2
## Bus   1  1 47  0
## Van   1  0  0 53
```

The confusion matrix shows that there is a high misclassification rate in classes '2D' and '4D' and it seems hard to separate these two classes from each other as we see that 19 cases (out of total 55) of class '2D' are misclassified as class '4D' and 18 cases (out of total 54) of class '4D' are misclassified as class '2D'. On the other hand, classes 'Bus' and 'Van' have a very low misclassification rate where we see only two misclassified cases (out of total 49) for class 'Bus' and only one misclassified case (out of total 54) for 'Van'.

## 2(a).

```
# fit lasso
model.logit.lasso <- cv.glmnet(
  x = as.matrix(set1.rescale[, 1:18]),
  y = set1.rescale[, 19], family = "multinomial"
)
#model.logit.lasso
#plot(model.logit.lasso)
# find nonzero lasso coefficients
c <- coef(model.logit.lasso, s = model.logit.lasso$lambda.min)
cmat <- cbind(
  as.matrix(c[[1]]), as.matrix(c[[2]]),
  as.matrix(c[[3]]), as.matrix(c[[4]])
)
cmat != 0
```



##	1	1	1	1
## (Intercept)	TRUE	TRUE	TRUE	TRUE
## Compactness	TRUE	TRUE	TRUE	TRUE
## Circularity	TRUE	TRUE	FALSE	FALSE
## Distance.Circularity	TRUE	TRUE	TRUE	TRUE
## Radius.Ratio	TRUE	TRUE	TRUE	TRUE
## Pr.Axis.Aspect.Ratio	TRUE	TRUE	TRUE	TRUE
## Max.Length.Aspect.Ratio	TRUE	TRUE	TRUE	TRUE
## Scatter.Ratio	FALSE	FALSE	FALSE	TRUE
## Elongatedness	TRUE	TRUE	TRUE	TRUE
## Pr.Axis.Rectangularity	FALSE	TRUE	FALSE	TRUE
## Max.Length.Rectangularity	TRUE	TRUE	TRUE	TRUE
## Scaled.Variance.Along.Major.Axis	TRUE	FALSE	TRUE	FALSE
## Scaled.Variance.Along.Minor.Axis	TRUE	TRUE	FALSE	FALSE
## Scaled.Radius.of.Gyration	TRUE	TRUE	TRUE	TRUE
## Skewness.About.Major.Axis	TRUE	TRUE	TRUE	TRUE
## Skewness.About.Minor.Axis	TRUE	TRUE	TRUE	TRUE
## Kurtosis.About.Minor.Axis	TRUE	TRUE	TRUE	TRUE
## Kurtosis.About.Major.Axis	TRUE	TRUE	TRUE	TRUE
## Hollows.Ratio	TRUE	TRUE	TRUE	TRUE

The pattern is somewhat similar to ANOVA, because the variables that consistently appear in all four logistic regression models with LASSO are the same variables identified as important in ANOVA; and the only exception is the variable 'Elongatedness,' which, while deemed unimportant in ANOVA, appears to influence all four logistic regression models in the LASSO.

## 2(b).

```
# compute train and test error
lasso.pred.train <- predict(
  object = model.logit.lasso, type = "class",
  s = model.logit.lasso$lambda.min,
  newx = as.matrix(set1.rescale[, 1:18])
)
(lassomisclass.train <-
  mean(ifelse(lasso.pred.train == set1$class, yes = 0, no = 1)))
```

```
## [1] 0.1671924
```

```
lasso.pred.test <- predict(model.logit.lasso,
  type = "class",
  s = model.logit.lasso$lambda.min,
  newx = as.matrix(set2.rescale[, 1:18])
)
(lassomisclass.test <-
  mean(ifelse(lasso.pred.test == set2$class, yes = 0, no = 1)))
```

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

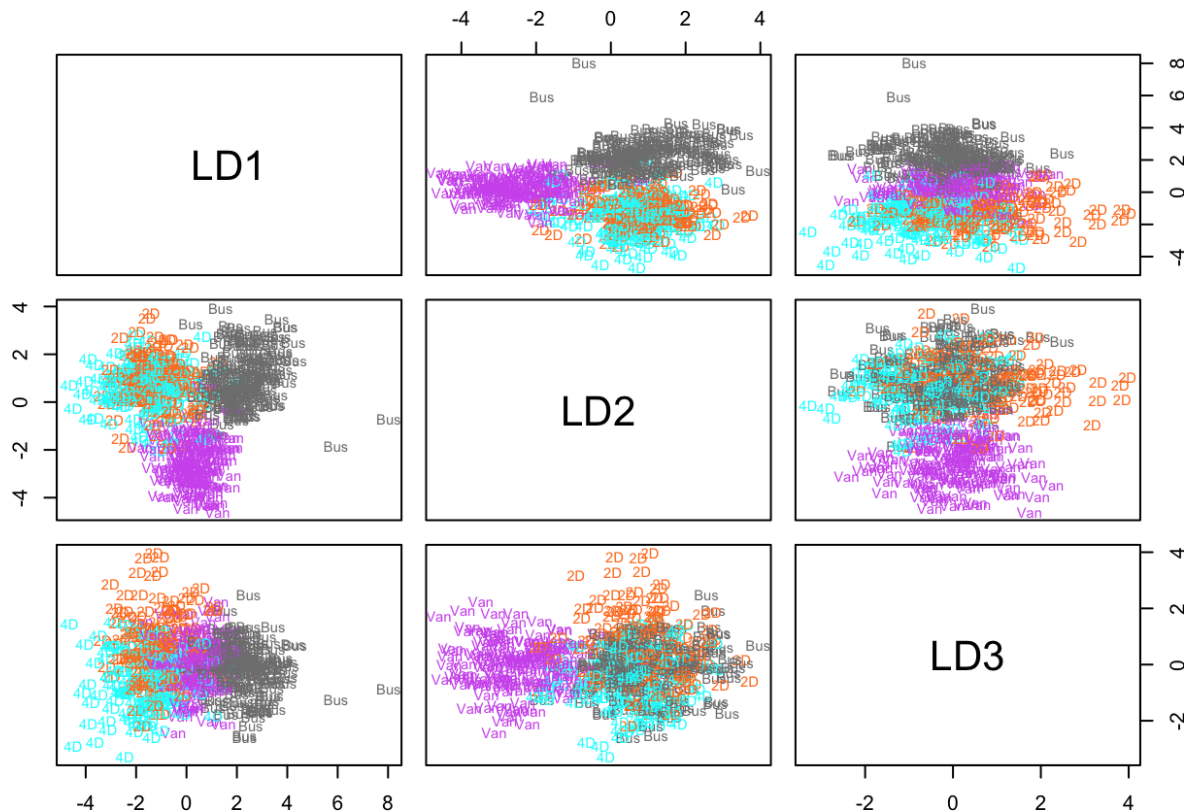
```
# compute standard error for test set
(se.lasso <- sqrt(lassomisclass.test * (1 - lassomisclass.test) / nrow(set2)))
```

```
## [1] 0.0283093
```

The test error for the LASSO version of logistic regression closely mirrors the test error observed in logistic regression using 'multinom()'. Therefore, the LASSO version of logistic regression outperforms KNN in terms of predictive accuracy, although its performance is comparable to logistic regression using 'multinom()'.

### 3(a).

```
# fit lda
model.lda <- lda(x = set1[, -19], grouping = set1$class)
# plot results
class.col <- ifelse(set1$class=="2D", y = 53, n =
                    ifelse(set1$class=="4D", y = 68, n =
                            ifelse(set1$class=="Bus", y = 203, n = 464)))
plot(model.lda, col = colors()[class.col])
```



LD1 seems to mostly separate classes 'Bus' and 'Van'.

LD2 seems to mostly separate classes 'Bus' and '4D'.

LD3 seems to mostly separate classes 'Van' and '4D'.

### 3(b).

```
# compute train and test error
lda.pred.train <- predict(model.lda, newdata = set1[, -19])$class
(ldamisclass.train <- mean(ifelse(lda.pred.train == set1$class, yes = 0, no = 1)))
```

```
## [1] 0.1971609
```

```
lda.pred.test <- predict(model.lda, newdata = set2[, -19])$class
(ldamisclass.test <- mean(ifelse(lda.pred.test == set2$class, yes = 0, no = 1)))
```

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

```
# compute se for test set
(se.lda <- sqrt(ldamisclass.test * (1 - ldamisclass.test) / nrow(set2)))
```

```
## [1] 0.02737444
```

The test error for Linear Discriminant Analysis (LDA) is the lowest among all the models, suggesting that LDA outperforms KNN, logistic regression using 'multinom()', and the LASSO version of logistic regression in terms of predictive accuracy.