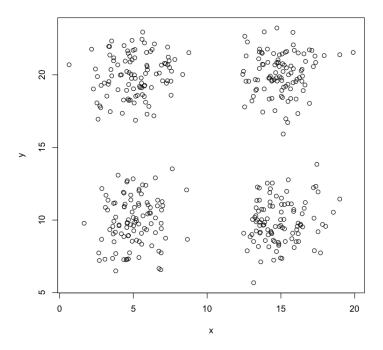
$Math\ 5365$

Data Mining 1

Homework 17

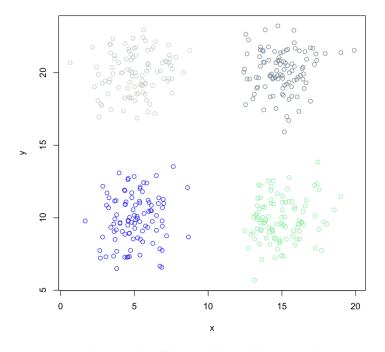
Mary Barker

1. Generate a data set similar to the one displayed, where each of the four clusters has 100 points.

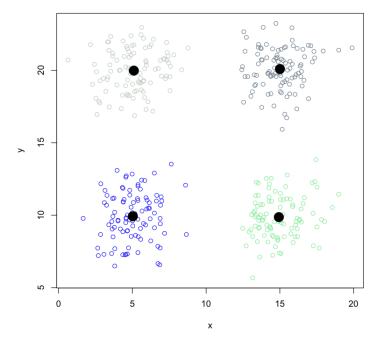


- (a) Perform a K-means clustering with K=4 and 1000 repetitions.
- (b) Plot the points and color them based on which clusters they are in.

clusters using 1000 repetitions



clusters using 1000 repetitions with centers shown



(c) Find the total, total within, and between sums of squares

The total sum of squares is 21777.85.

The total within sum of squares is 1711.74.

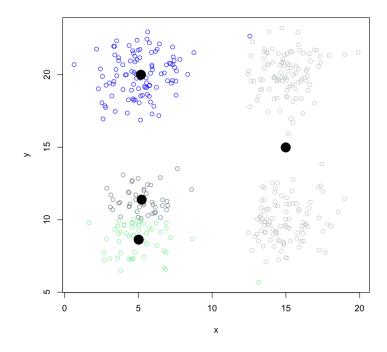
The total between sum of squares is 20066.11.

(d) Find the centers of the clusters.

The centers are in the table below

	X	У
1	5.034003	9.927154
2	15.004838	20.085618
3	14.931673	9.870831
4	5.100832	19.972832

- (e) Suppose we want at least one of our repetitions of K-means to have the property that every cluster contains exactly one initial centroid. How many repetitions would be necessary to ensure that this happens with at least 99% probability? The minimum number of necessary repetitions is 47.
- (f) Can you find a choice of initial centers that does not result in the optimal clusters? What is the total within sum of squares for that clustering? The center of each cluster is plotted together with the colored points



The total within sum of squares for this case is 6728.091.

2. Perform a K-means clustering with K = 2 and 1000 repetitions for the wdbc data set. What classification accuracy would be achieved if the clusters were used to predict the diagnosis in this data set?

The classification accuracy using K=2 and 1000 repetitions is roughly 85.41301%

1. Generate a data set similar to the one displayed, where each of the

```
four clusters has 100 points.
   #
10
   set.seed(0)
   x \leftarrow c(rnorm(100, 5, 1.5), rnorm(100, 15, 1.5),
12
           rnorm(100, 5, 1.5), rnorm(100, 15, 1.5))
13
   y \leftarrow c(rnorm(100, 10, 1.5), rnorm(100, 10, 1.5),
           rnorm(100, 20, 1.5), rnorm(100, 20, 1.5))
15
   plot(x, y)
   points \leftarrow data.frame(x = x, y = y)
17
18
        (a) Perform a K-means clustering with K = 4 and 1000 repetitions.
19
20
   kmeans_reps <- function(dset, centers, reps){</pre>
21
     w_s = Inf
22
     for(i in 1:reps){
23
       k_cluster <- kmeans(x = dset, centers = centers)</pre>
24
        if((k_cluster$tot.withinss) < w_ss){</pre>
25
          ssw = k_cluster$tot.withinss
26
          my_k_cluster <- k_cluster
       }
28
     }
     return(my_k_cluster)
30
   }
31
   reps = 1000
32
   mycluster <- kmeans_reps(points, 4, reps)</pre>
33
        (b) Plot the points and color them based on which clusters they are in.
35
   plot(x, y, col=c('blue',
```

```
'slategray',
37
                    'lightgreen',
38
                    'honeydew3',
39
                    'orange',
40
                    'brown') [mycluster$cluster],
41
                    main='clusters using 1000 repetitions')
42
   plot(x, y, col=c('blue',
                    'slategray',
44
                    'lightgreen',
45
                    'honeydew3',
46
                    'orange',
47
                    'brown') [mycluster$cluster],
48
                    main = 'clusters using 1000 repetitions with centers shown')
49
   lines(mycluster$centers,type='p',pch=16,col='black', cex = 2.5)
50
        (c) Find the total, total within, and between sums of squares
51
   mycluster$totss
   mycluster$tot.withinss
53
   mycluster$betweenss
55
   #
       (d) Find the centers of the clusters.
56
   mycluster$centers
57
58
   #
        (e) Suppose we want at least one of our repetitions of K-means to have
59
   #
            the property that every cluster contains exactly one initial centroid.
60
            How many repetitions would be necessary to ensure that this happens
            with at least 99% probability?
62
63
```

```
minimum_reps <- function(k, err){
     ceiling(log(err) / log(1 - factorial(k) / (k^k)))
   }
66
   minimum_num = minimum_reps(4, 0.01)
67
   #
       (f) Can you find a choice of initial centers that does not result in the
69
            optimal clusters? What is the total within sum of squares for that
70
            clustering?
71
72
   # try with alternate centers:
73
   centers04 <- cbind(c(10, 15, 10, 5), c(10, 15, 20, 15))
74
   newcluster04 <- kmeans(x = points, centers = centers04)</pre>
75
   plot(x, y, col=c('blue',
76
                     'slategray',
                     'lightgreen',
78
                     'honeydew3',
79
                     'orange',
80
                     'brown') [newcluster04$cluster],
                    main = 'clusters using 4 centers')
82
   lines(newcluster04$centers,type='p',pch=16,col='black', cex = 2.5)
83
84
   centers03 <- cbind(c(4, 15, 5, 16), c(10, 20, 10, 20))
85
   newcluster03 <- kmeans(x = points, centers = centers03)</pre>
   plot(x, y, col=c('blue',
87
                     'slategray',
                     'lightgreen',
89
                     'honeydew3',
90
```

```
'orange',
91
                     'brown') [newcluster03$cluster],
92
                     main = 'clusters using 4 centers')
93
    lines(newcluster03$centers,type='p',pch=16,col='black', cex = 2.5)
94
95
    centers02 <- cbind(c(5, 15, 15, 15), c(10, 15, 20, 12))
96
    newcluster02 <- kmeans(x = points, centers = centers02)</pre>
    plot(x, y, col=c('blue',
98
                     'slategray',
                     'lightgreen',
100
                     'honeydew3',
101
                     'orange',
102
                     'brown') [newcluster02$cluster],
103
                     main = 'clusters using 4 centers')
104
    lines(newcluster02$centers,type='p',pch=16,col='black', cex = 2.5)
105
106
    centers01 <- cbind(c(2, 10, 10, 20), c(2, 10, 20, 25))
107
    newcluster01 <- kmeans(x = points, centers = centers01)</pre>
    colors = 5 * c(1:50)
109
    plot(x, y, col=c('blue',
                     'slategray',
111
                     'lightgreen',
112
                     'honeydew3',
113
                     'orange',
114
                     'brown') [newcluster01$cluster],
115
                     main = 'clusters using 4 centers')
116
    lines(newcluster01$centers,type='p',pch=16,col='black', cex = 2.5)
```

```
118
    # try with k = 1
119
    newcluster1 <- kmeans(x = points, centers = 1)</pre>
120
    plot(x, y, col=c('blue',
121
                      'slategray',
122
                      'lightgreen',
123
                      'honeydew3',
124
                      'orange',
125
                      'brown') [newcluster1$cluster],
126
                      main = 'clusters using 1 center')
127
    lines(newcluster1$centers,type='p',pch=16,col='black', cex = 2.5)
128
129
    # try with k = 2
130
    newcluster2 <- kmeans(x = points, centers = 2)</pre>
131
    plot(x, y, col=c('blue',
132
                      'slategray',
133
                      'lightgreen',
134
                      'honeydew3',
135
                      'orange',
136
                      'brown') [newcluster2$cluster],
137
                      main = 'clusters using 2 centers')
138
    lines(newcluster2$centers,type='p',pch=16,col='black', cex = 2.5)
139
140
    # try with k = 2
141
    newcluster21 <- kmeans(x = points, centers = cbind(c(5, 16), c(10, 21)))
    plot(x, y, col=c('blue',
143
                      'slategray',
144
```

```
'lightgreen',
145
                      'honeydew3',
146
                      'orange',
147
                      'brown') [newcluster21$cluster],
148
                      main = 'clusters using 2 centers')
149
    lines(newcluster21$centers,type='p',pch=16,col='black', cex = 2.5)
150
151
    # try with k = 6
152
    newcluster6 <- kmeans(x = points, centers = 6)</pre>
153
    plot(x, y, col=c('blue',
154
                      'slategray',
155
                      'lightgreen',
156
                      'honeydew3',
157
                      'orange',
158
                      'brown',
159
                      'yellow',
160
                      'plum') [newcluster6$cluster],
161
                      main = 'clusters using 6 centers')
162
    lines(newcluster6$centers,type='p',pch=16,col='black', cex = 2.5)
163
164
    # try with k = 8
165
    newcluster8 <- kmeans(x = points, centers = 8)</pre>
166
    plot(x, y, col=c('blue',
167
                      'slategray',
168
                      'lightgreen',
169
                      'honeydew3',
170
                      'orange',
171
```

```
'brown',
172
                      'yellow',
173
                      'plum') [newcluster8$cluster],
174
                      main = 'clusters using 8 centers')
175
    lines(newcluster8$centers,type='p',pch=16,col='black', cex = 2.5)
176
177
    # try with k = 10
178
    newcluster10 <- kmeans(x = points, centers = 10)</pre>
179
    plot(x, y, col=colors[newcluster10$cluster],
180
                      main = 'clusters using 10 centers')
181
    lines(newcluster10$centers,type='p',pch=16,col='black', cex = 2.5)
182
183
    #brute force:
184
    d \leftarrow function(x1, x2){
185
      return(sqrt(sum((x1 - x2)^2)))
186
    }
187
    count = 1
188
    list_of_clusters <- list()</pre>
    keep_going = TRUE
190
    for(i in 1:100000){
191
      if(keep_going){
192
        newcluster05 <- kmeans(x = points, centers = 4)</pre>
193
        plot(x, y, col=c('blue',
194
                           'slategray',
195
                           'lightgreen',
196
                           'honeydew3',
197
                           'orange',
198
```

```
'brown') [newcluster05$cluster])
199
        lines(newcluster05$centers,type='p',pch=16,col='black', cex = 2.5)
200
        x1 <- as.numeric(newcluster05$centers[1,])</pre>
201
        x2 <- as.numeric(newcluster05$centers[2,])</pre>
202
        x3 <- as.numeric(newcluster05$centers[3,])</pre>
203
        x4 <- as.numeric(newcluster05$centers[4,])
204
        mindist <- min(c(</pre>
205
           d(x1, x2), d(x1, x3), d(x1, x4),
206
           d(x2, x3), d(x2, x4),
207
           d(x3, x4)
208
        ))
209
        if(mindist < 5){
210
                 list_of_clusters[[count]] <- newcluster05</pre>
211
                 count = count + 1
212
                 keep_going = FALSE
213
        }
214
      }
215
    }
216
217
    \# 2. Perform a K-means clustering with K = 2 and 1000 repetitions for the
218
         wdbc data set. What classification accuracy would be achieved if the
219
         clusters were used to predict the diagnosis in this data set?
220
221
    wdbc_cluster <- kmeans_reps(wdbc[,2:ncol(wdbc)], 2, 1000)</pre>
222
    acc <- confmatrix(wdbc$V2, wdbc_cluster$cluster)</pre>
223
```