## Big Data Analytics Using R

Session 9 Clustering In-Class Exercise Solution

**Question**: Suppose that we have 5 observations, for which we compute a dissimilarity (distance) matrix as follows:

 $\mathbf{D}_{0} = \begin{bmatrix} A & B & C & D & E \\ B & 0 & & & \\ D & 0 & & & \\ D & 3 & 7 & 0 & & \\ B & 0 & 5 & 9 & 0 & \\ 11 & 10 & 2 & 8 & 0 \end{bmatrix}$ 

On the basis of the dissimilarity matrix, sketch the dendrogram that results from hierarchically clustering these 5 observations using complete linkage.

## Solution:

- Step 1: Initially, there are 5 clusters: {A}, {B}, {C}, {D}, {E}. Find the smallest distance in the distance matrix D<sub>0</sub>: d(C, E) = 2. So we fuse C and E. There are 4 clusters left: {A}, {B}, {C, E}, {D}.
- Step 2(a): Now we construct the new distance matrix:

$$\mathbf{D}_{1} = \begin{array}{ccc} A & B & CE & D \\ \\ A \\ B \\ CE \\ D \end{array} \begin{pmatrix} 0 & & & \\ 9 & 0 & & \\ 8 & \dagger & 0 \\ 6 & 5 & \# & 0 \end{array} \right)$$

The values shown above are derived from the previous matrix  $\mathbb{D}_0$ . We are to calculate the three unknown values.

$$* : d({A}, {C, E}) = \max\{d(A, C), d(A, E)\} = \max\{3, 11\} = 11$$
  
$$+ : d({B}, {CE}) = \max\{d(B, C), d(B, E)\} = \max\{7, 10\} = 10$$
  
$$# : d({D}, {C, E}) = \max\{d(D, C), d(D, E)\} = \max\{9, 8\} = 9$$

Thus, the new distance matrix is

$$\mathbf{D}_{1} = \begin{array}{ccc} A & B & CE & D \\ A & 0 & & \\ B & 0 & & \\ CE & 11 & 10 & 0 \\ D & 6 & 5 & 9 & 0 \end{array} \right)$$

- Step 2(b): Find the smallest distance in the distance matrix  $\mathbf{D}_1$ : d(B, D) = 5. So we fuse B and D. There are 3 clusters left:  $\{A\}, \{B, D\}, \{C, E\}$ .
- Step 3(a): Now we construct the new distance matrix:

$$\mathbf{D}_{2} = \begin{array}{c} A & BD & CE \\ BD \\ CE \end{array} \begin{pmatrix} 0 & & \\ ** & 0 & \\ 11 & \dagger \dagger & 0 \end{array}$$

The values shown above are derived from the previous matrix  $\mathbb{D}_1$ . We are to calculate the two unknown values.

\*\* : 
$$d({A}, {BD}) = \max\{d(A, B), d(A, D)\} = \max\{9, 6\} = 9$$
  
†† : $d({B, D}, {C, E}) = \max\{d(B, {C, E}), d(D, {C, E})\} = \max\{10, 9\} = 10$ 

Thus, the new distance matrix is

$$\mathbf{D}_{2} = \begin{array}{c} A & BD & CE \\ A & \begin{pmatrix} 0 & & \\ 9 & 0 & \\ CE & \\ 11 & 10 & 0 \end{array} \right)$$

- Step 3(b): Find the smallest distance in the distance matrix  $\mathbf{D}_2$ :  $d(A, \{B, D\}) = 9$ . So we fuse A and  $\{B, D\}$ . There are 2 clusters left:  $\{A, B, D\}, \{C, E\}$ .
- Step 4: There is only one way to fuse the two clusters. It remains to calculate the distance  $d(\{A, B, D\}, \{C, E\})$ .

 $d(\{A,B,D\},\{C,E\}) = \max\{d(A,\{C,E\}),d(\{B,D\},\{C,E\})\} = \max\{11,10\} = 11$ 

As a result, we can obtain the dendrogram as shown in Fig. 1.



Figure 1: The dendrogram derived by R

The R code for this dendrogram:

```
> distance <- matrix(0,5,5) #create a 5*5 matrix initially all 0
#Create a lower triangular matrix, filled by columns by default
> distance[lower.tri(distance, diag = 0)] <- c(9,3,6,11,7,5,10,9,2,8)
#as.dist() converts it to the distance matrix form that hclust() will recognise
> distance <- as.dist(distance)
> hc.complete <- hclust(distance, method="complete")
# hc.single <- hclust(distance, method="single")
# hc.average <- hclust(distance, method="single")
# hc.average <- hclust(distance, method="average")
# par(mfrow=c(1,3)) #show plots in 1 row, 3 columns
> plot(hc.complete, main="Complete Linkage",xlab="",ylab="",cex=0.9)
# plot(hc.single, main="Single Linkage",xlab="",ylab="",cex=0.9)
# plot(hc.average, main="Average Linkage",xlab="",ylab="",cex=0.9)
```

The resulting plot is shown in Fig. 2.



Figure 2: The dendrogram derived by R