Cluster Analysis I

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1 Preliminaries

In this section, the RStudio workspace and console panes are cleared of old output, variables, and other miscellaneous debris. Packages are loaded and any required data files are retrieved.

```
options(replace.assign = TRUE, width = 65, digits = 4, scipen = 4, fig.width = 4,
    fig.height = 4)

# Clear the workspace and console.
rm(list = ls(all = TRUE))
cat("\f")

# Turn off showing of significance asterisks.
options(show.signif.stars = F)
# Set the contrast option; important for ANOVAs.
options(contrasts = c("contr.sum", "contr.poly"))
how_long <- Sys.time()
set.seed(123)
library(knitr)</pre>
```

```
library(psych)
library(ggplot2)
##
## Attaching package: 'ggplot2'
## The following objects are masked from 'package:psych':
##
      %+%, alpha
library(MASS)
library(sciplot)
library(ggplot2)
library(vegan)
## Warning: package 'vegan' was built under R version 3.5.1
## Loading required package: permute
## Warning: package 'permute' was built under R version 3.5.1
## Loading required package: lattice
## This is vegan 2.5-2
library(smacof)
```

```
## Warning: package 'smacof' was built under R version 3.5.1
## Loading required package: plotrix
## Attaching package: 'plotrix'
## The following object is masked from 'package:psych':
##
##
      rescale
##
## Attaching package: 'smacof'
## The following object is masked from 'package:base':
##
##
      transform
library(ape)
library(ade4)
## Warning: package 'ade4' was built under R version 3.5.1
library(scatterplot3d)
library(cluster)
library(factoextra)
## Warning: package 'factoextra' was built under R version 3.5.1
## Welcome! Related Books: 'Practical Guide To Cluster Analysis in R' at https://goo.gl/13EFCZ
library(ggdendro)
## Warning: package 'ggdendro' was built under R version 3.5.1
```

2 Simple Case

Let's begin with a very simple problem: 5 cases that are arranged in two-dimensional space for which Euclidean distance is easily visualized. To emphasize the distances a bit we can use squared Euclidean distance.

```
Data <- matrix(c(2, 3, 4, 6, 7, 2, 2, 5, 4, 6), nrow = 5, ncol = 2,
   byrow = FALSE)
row.names(Data) <- c("Object 1", "Object 2", "Object 3", "Object 4",</pre>
    "Object 5")
Data
           [,1] [,2]
## Object 1
            2
            3
## Object 2
                   2
## Object 3
            4 5
## Object 4
            6 4
## Object 5
            7 6
Data_Dist <- (dist(Data, method = "euclidean"))^2</pre>
Data_Dist
           Object 1 Object 2 Object 3 Object 4
## Object 2
                  1
## Object 3
                 13
                          10
## Object 4
                 20
                           13
                                    5
## Object 5
                                             5
                 41
                           32
                                    10
```

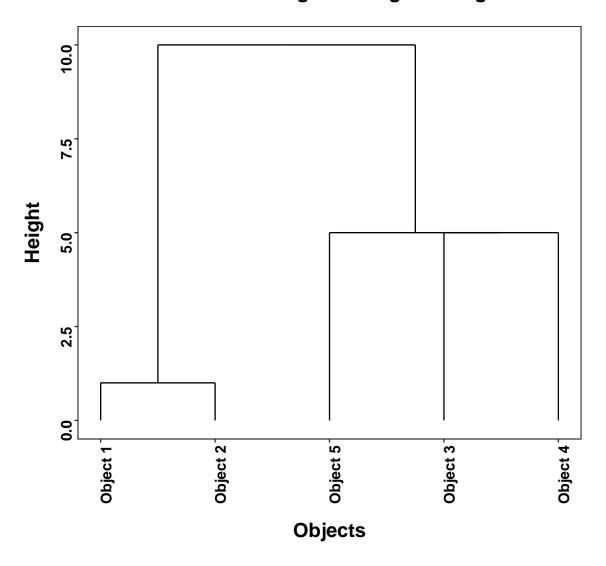
2.1 Single Linkage

The dissimilarity between two clusters (A and B) is the minimum of all possible distances between the cases in Cluster A and the cases in Cluster B.

```
hc_1 <- hclust(Data_Dist, method = "single")</pre>
hc_1$merge
        [,1] [,2]
## [1,]
        -1
              -2
## [2,]
        -3
              -4
## [3,]
        -5
              2
## [4,]
        1
hc_1$height
## [1] 1 5 5 10
```

```
ggdendrogram(hc_1, theme_dendro = FALSE, size = 4) + xlab("Objects") +
   ylab("Height") + theme(text = element_text(size = 14, family = "sans",
   color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
   size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
   size = 12, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
   0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
   15, 0, 0), size = 16, angle = 90), axis.line.x = element_blank(),
   axis.line.y = element_blank(), plot.title = element_text(size = 16,
        face = "bold", margin = margin(0, 0, 20, 0), hjust = 0.5),
   panel.background = element_rect(fill = "white", linetype = 1,
        color = "black"), panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
   plot.background = element_rect(fill = "white"), plot.margin = unit(c(1,
        1, 1, 1), "cm"), legend.position = "bottom", legend.title = element_blank()) +
   ggtitle("Cluster Dendogram: Single Linkage")
```

Cluster Dendogram: Single Linkage

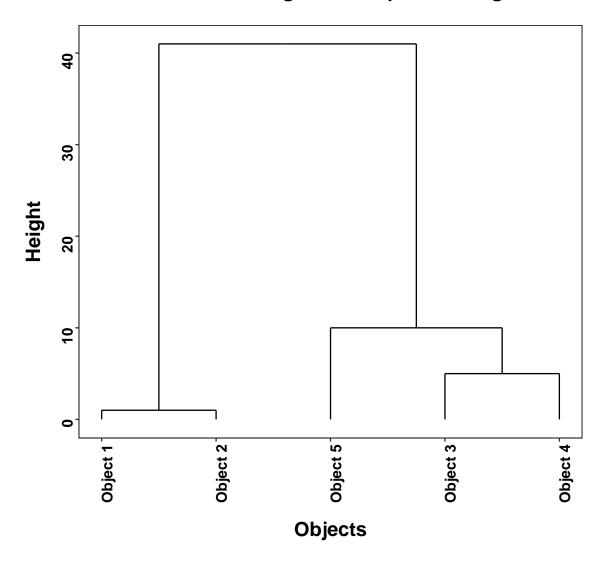


2.2 Complete Linkage

The dissimilarity between two clusters (A and B) is the maximum of all possible distances between the cases in Cluster A and the cases in Cluster B.

```
hc_2 <- hclust(Data_Dist, method = "complete")</pre>
hc_2$merge
        [,1] [,2]
## [1,]
        -1
              -2
## [2,]
        -3
              -4
## [3,]
        -5
              2
## [4,]
        1
hc_2$height
## [1] 1 5 10 41
```

Cluster Dendogram: Complete Linkage

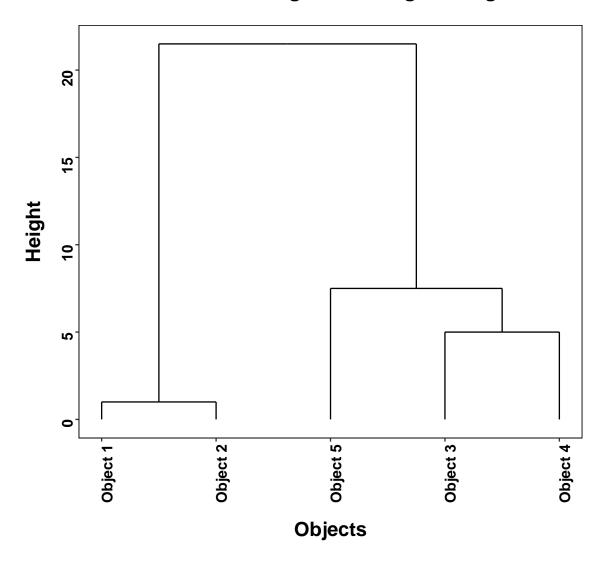


2.3 Average Linkage

The dissimilarity between two clusters (A and B) is the average of all possible distances between the cases in Cluster A and the cases in Cluster B.

```
hc_3 <- hclust(Data_Dist, method = "average")</pre>
hc_3$merge
        [,1] [,2]
## [1,]
        -1
              -2
## [2,]
        -3
              -4
## [3,]
        -5
             2
## [4,]
        1
hc_3$height
## [1] 1.0 5.0 7.5 21.5
```

Cluster Dendogram: Average Linkage

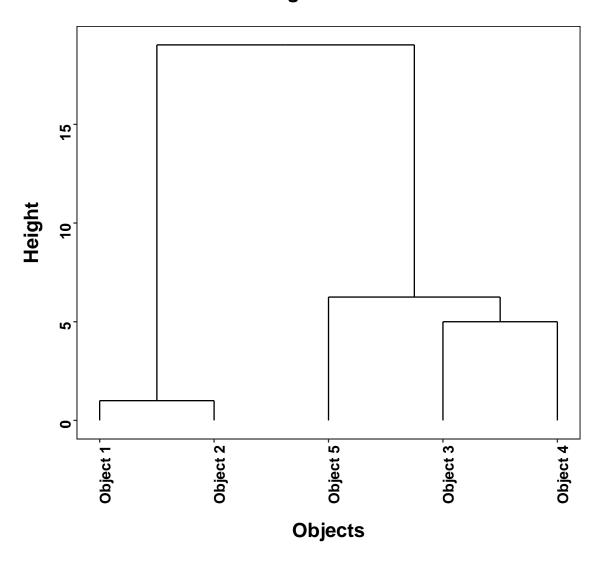


2.4 Centroid Method

The dissimilarity between two clusters (A and B) is the distance between the centroid for the cases in Cluster A and the centroid for the cases in Cluster B.

```
hc_4 <- hclust(Data_Dist, method = "centroid")</pre>
hc_4$merge
        [,1] [,2]
## [1,]
        -1
              -2
## [2,]
        -3
              -4
        -5
## [3,]
             2
## [4,]
        1
hc_4$height
## [1] 1.00 5.00 6.25 19.03
```

Cluster Dendogram: Centroid Method



2.5 Ward's Method

The dissimilarity between two clusters (A and B) is the loss of information from joining the clusters, measured by the increase in error sum of squares.

The sum of squares for a cluster is the sum of squared deviations of each case from the centroid for the cluster. The error sum of squares is the total of these for all clusters. The two clusters among all possible combinations that have the minimum increase in error sum of squares are joined.

Two versions are available. The Ward D method should be chosen if squared Euclidean distances are used. the Ward D2 method will produce the traditional Ward solution starting from Euclidean distances.

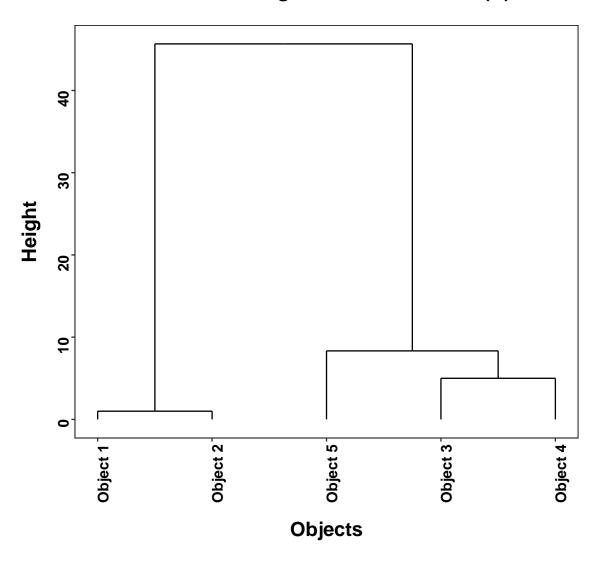
```
hc_5 <- hclust(Data_Dist, method = "ward.D")
hc_5$merge

## [,1] [,2]
## [1,] -1 -2
## [2,] -3 -4
## [3,] -5 2
## [4,] 1 3

hc_5$height

## [1] 1.000 5.000 8.333 45.667
```

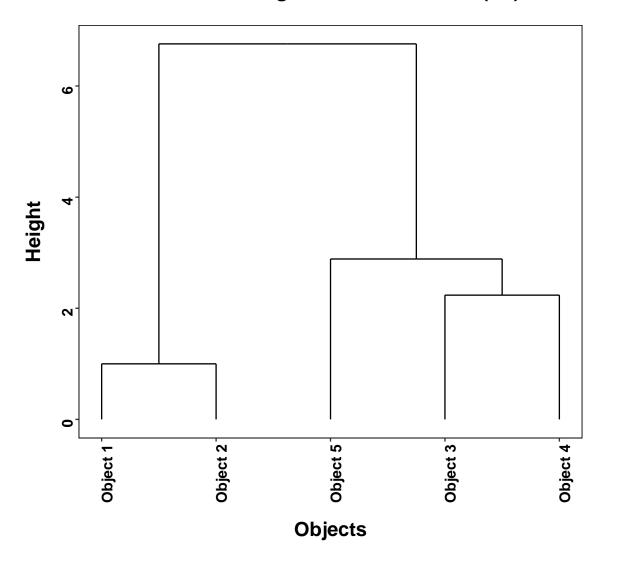
Cluster Dendogram: Ward's Method (D)



```
Data_Dist <- (dist(Data, method = "euclidean"))
hc_5 <- hclust(Data_Dist, method = "ward.D2")
hc_5$merge

## [,1] [,2]
## [1,] -1 -2
## [2,] -3 -4
## [3,] -5 2
## [4,] 1 3
hc_5$height
## [1] 1.000 2.236 2.887 6.758</pre>
```

Cluster Dendogram: Ward's Method (D2)



3 Iris Data

An important question is how well the different clustering methods can recover a group structure when it is known in advance. That can lend insight into the ability of the methods to identify any group structure when that structure is not known in advance.

A classic test data set was introduced by R. A. Fisher (1936): three species of iris, varying in their petal length, petal width, sepal length, and sepal width.

```
# Get the drug use data from the working directory.
setwd("C:\\Courses\\Psychology 516\\PowerPoint\\2018")
Iris <- read.table("iris.csv", sep = ",", header = TRUE)</pre>
Iris <- as.data.frame(Iris)</pre>
Iris$Species[Iris$Species == "1"] <- "Setosa"</pre>
Iris$Species[Iris$Species == "2"] <- "Versicolor"</pre>
Iris$Species[Iris$Species == "3"] <- "Virginica"</pre>
Iris
##
        Sepal_Length Sepal_Width Petal_Length Petal_Width
                                                                   Species
## 1
                   50
                                33
                                                              2
                                                                    Setosa
                                               14
## 2
                   64
                                28
                                                            22
                                                                Virginica
## 3
                   65
                                28
                                               46
                                                            15 Versicolor
## 4
                   67
                                31
                                               56
                                                            24
                                                                 Virginica
## 5
                   63
                                28
                                               51
                                                                 Virginica
## 6
                   46
                                34
                                               14
                                                             3
                                                                    Setosa
## 7
                                               51
                                                            23
                   69
                                31
                                                                Virginica
## 8
                   62
                                22
                                               45
                                                            15 Versicolor
## 9
                   59
                                32
                                               48
                                                            18 Versicolor
## 10
                                                             2
                                                                    Setosa
                   46
                                36
                                               10
## 11
                                30
                                               46
                                                            14 Versicolor
## 19
                   60
                                27
                                               51
                                                            16 Versicolor
## 13
                                30
                                                                Virginica
                                25
                                               39
## 14
                   56
                                                            11 Versicolor
## 15
                                30
                                               55
                   65
                                                                Virginica
## 16
                   58
                                27
                                               51
                                                            19
                                                                 Virginica
## 17
                   68
                                32
                                                                 Virginica
## 18
                   51
                                33
                                               17
                                                             5
                                                                    Setosa
## 19
                   57
                                28
                                               45
                                                            13 Versicolor
## 20
                   62
                                34
                                               54
                                                            23
                                                                Virginica
## 21
                   77
                                38
                                               67
                                                            22
                                                                Virginica
## 22
                   63
                                33
                                               47
                                                            16 Versicolor
## 23
                   67
                                33
                                               57
                                                                 Virginica
## 24
                   76
                                30
                                               66
                                                            21
                                                                 Virginica
## 25
                   49
                                25
                                               45
                                                            17
                                                                 Virginica
                                                             2
## 26
                   55
                                35
                                               13
                                                                    Setosa
## 27
                   67
                                30
                                               52
                                                            23
                                                                Virginica
## 28
                   70
                                32
                                               47
                                                            14 Versicolor
## 29
                   64
                                32
                                               45
                                                            15 Versicolor
## 30
                                28
                                               40
                                                            13 Versicolor
                   61
## 31
                                                             2
                   48
                                31
                                               16
                                                                    Setosa
## 32
                   59
                                30
                                               51
                                                                Virginica
## 33
                   55
                                24
                                               38
                                                            11 Versicolor
## 34
                                                                Virginica
```

	35	64	32	53	23	Virginica
##	36	52	34	14	2	Setosa
##	37	49	36	14	1	Setosa
##	38	54	30	45		Versicolor
##	39	79	38	64	20	Virginica
##	40	44	32	13	2	Setosa
##	41	67	33	57	21	Virginica
##	42	50	35	16	6	Setosa
##	43	58	26	40	12	Versicolor
##	44	44	30	13	2	Setosa
##	45	77	28	67	20	Virginica
##	46	63	27	49	18	Virginica
##	47	47	32	16	2	Setosa
##	48	55	26	44		Versicolor
##	49	50	23	33		Versicolor
; ;; ; #	50	72	32	60	18	Virginica
+# ##						
	51	48	30	14	3	Setosa
##	52	51	38	16	2	Setosa
##	53	61	30	49	18	Virginica
##	54	48	34	19	2	Setosa
#	55	50	30	16	2	Setosa
##	56	50	32	12	2	Setosa
##	57	61	26	56	14	Virginica
##	58	64	28	56	21	Virginica
##	59	43	30	11	1	Setosa
##	60	58	40	12	2	Setosa
##	61	51	38	19	4	Setosa
##	62	67	31	44	14	Versicolor
##	63	62	28	48	18	Virginica
##	64	49	30	14	2	Setosa
##	65	51	35	14	2	Setosa
##	66	56	30	45	15	Versicolor
##	67	58	27	41		Versicolor
##	68	50	34	16	4	Setosa
; ;; ‡#	69	46	32	14	2	Setosa
; #	70	60	29	45		Versicolor
	71	57	26	35		Versicolor
	72	57	44	15	4	Setosa
+# ‡#	73	50	36	14		
					2	Setosa
	74	77	30	61	23	Virginica
##	75 76	63	34	56	24	Virginica
##	76	58	27	51	19	Virginica
##	77	57	29	42		Versicolor
##	78	72	30	58	16	Virginica
##	79	54	34	15	4	Setosa
##	80	52	41	15	1	Setosa
##	81	71	30	59	21	Virginica
##	82	64	31	55	18	Virginica
##	83	60	30	48	18	Virginica
##	84	63	29	56	18	Virginica
##	85	49	24	33		Versicolor
##	86	56	27	42		Versicolor
##		57	30	42		Versicolor
	88	55	42	14	2	
11						20024

##	89	49	31	15	2	Setosa
##	90	77	26	69	23	Virginica
##	91	60	22	50	15	Virginica
##	92	54	39	17	4	Setosa
##	93	66	29	46	13	Versicolor
##	94	52	27	39	14	Versicolor
##	95	60	34	45	16	Versicolor
##	96	50	34	15	2	Setosa
##	97	44	29	14	2	Setosa
##	98	50	20	35	10	Versicolor
##	99	55	24	37	10	Versicolor
##	100	58	27	39	12	Versicolor
##	101	47	32	13	2	Setosa
##	102	46	31	15	2	Setosa
##	103	69	32	57	23	Virginica
##	104	62	29	43		Versicolor
##	105	74	28	61	19	Virginica
##	106	59	30	42		Versicolor
##	107	51	34	15	2	Setosa
##	108	50	35	13	3	Setosa
	109	56	28	49	20	Virginica
	110	60	22	40		Versicolor
	111	73	29	63	18	Virginica
	112	67	25	58	18	Virginica
	113	49	31	15	1	Setosa
				47		Versicolor
	114	67	31			
	115	63	23	44		Versicolor
	116	54	37	15	2	Setosa
	117	56	30	41		Versicolor
##	118	63	25	49		Versicolor
##	119	61	28	47		Versicolor
##	120	64	29	43		Versicolor
##	121	51	25	30		Versicolor
##	122	57	28	41		Versicolor
##	123	65	30	58	22	Virginica
	124	69	31	54	21	Virginica
	125	54	39	13	4	Setosa
	126	51	35	14	3	Setosa
	127	72	36	61	25	Virginica
	128	65	32	51	20	Virginica
	129	61	29	47		Versicolor
##	130	56	29	36		Versicolor
##	131	69	31	49	15	Versicolor
##	132	64	27	53	19	Virginica
##	133	68	30	55	21	Virginica
##	134	55	25	40	13	Versicolor
##	135	48	34	16	2	Setosa
##	136	48	30	14	1	Setosa
##	137	45	23	13	3	Setosa
	138	57	25	50	20	Virginica
	139	57	38	17	3	Setosa
	140	51	38	15	3	Setosa
	141	55	23	40	13	Versicolor
	142	66	30	44		Versicolor
					_	

##	143	68	28	48	14	Versicolor
##	144	54	34	17	2	Setosa
##	145	51	37	15	4	Setosa
##	146	52	35	15	2	Setosa
##	147	58	28	51	24	Virginica
##	148	67	30	50	17	Versicolor
##	149	63	33	60	25	Virginica
##	150	53	37	15	2	Setosa

3.1 Descriptives and ANOVA

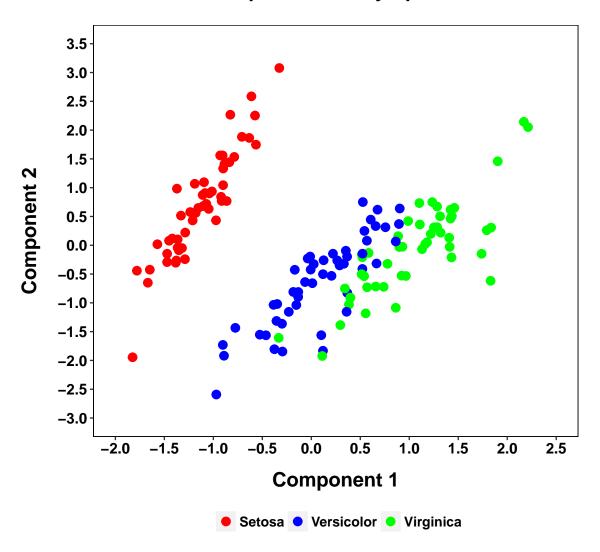
The basic features of the data can be seen in descriptive information and analyses of variance. A principal component analysis also provides a convenient way to capture and display most of the important variation in the data.

```
describeBy(Iris[, c(1:4)], group = Iris$Species, digits = 2)
## Descriptive statistics by group
## group: Setosa
            vars n mean sd median trimmed mad min max
## Sepal_Length 1 50 50.06 3.52 50 50.02 2.97 43 58
              2 50 34.28 3.79
                                34 34.15 3.71 23 44
## Sepal_Width
                              15
                                   14.60 1.48 10 19
## Petal_Length 3 50 14.62 1.74
## Petal_Width 4 50 2.46 1.05
                              2 2.38 0.00 1 6
##
     range skew kurtosis se
## Sepal_Length 15 0.11
                       -0.45 0.50
## Sepal_Width
              21 0.04
                        0.60 0.54
## Petal_Length 9 0.10
                        0.65 0.25
## Petal_Width
               5 1.18 1.26 0.15
## -----
## group: Versicolor
## vars n mean sd median trimmed mad min max
## Sepal_Length 1 50 59.36 5.16 59.0 59.38 5.19 49 70
              2 50 27.70 3.14 28.0 27.80 2.97 20 34
## Sepal_Width
## Petal_Length 3 50 42.60 4.70 43.5 42.92 5.19 30 51
## Petal_Width
              4 50 13.26 1.98 13.0 13.25 2.22 10 18
    range skew kurtosis se
## Sepal_Length 21 0.10 -0.69 0.73
## Sepal_Width 14 -0.34
                       -0.55 0.44
## Petal_Length 21 -0.57 -0.19 0.66
## Petal_Width 8 -0.03 -0.59 0.28
## -----
## group: Virginica
     vars n mean sd median trimmed mad min max
## Sepal_Length 1 50 65.88 6.36 65.0 65.72 5.93 49 79
              2 50 29.74 3.22 30.0 29.62 2.97 22 38
## Sepal_Width
## Petal_Length 3 50 55.52 5.52 55.50 55.10 6.67 45 69
## Petal_Width
            4 50 20.26 2.75 20.0 20.32 2.97 14 25
            range skew kurtosis se
## Sepal_Length 30 0.11 -0.20 0.90
              16 0.34
## Sepal_Width
                         0.38 0.46
## Petal_Length 24 0.52 -0.37 0.78
## Petal_Width
              11 -0.12 -0.75 0.39
# Check if the species are different in their sepal and petal
# measurements.
anova(aov(Iris$Sepal_Length ~ as.factor(Species), data = Iris))
## Analysis of Variance Table
## Response: Iris$Sepal_Length
                   Df Sum Sq Mean Sq F value Pr(>F)
## as.factor(Species) 2 6321 3161 119 <2e-16
## Residuals 147 3896 27
```

```
anova(aov(Iris$Sepal_Width ~ as.factor(Species), data = Iris))
## Analysis of Variance Table
##
## Response: Iris$Sepal_Width
                      Df Sum Sq Mean Sq F value Pr(>F)
## as.factor(Species) 2 1134 567
                                           49.2 <2e-16
## Residuals
                     147 1696
                                    12
anova(aov(Iris$Petal_Length ~ as.factor(Species), data = Iris))
## Analysis of Variance Table
## Response: Iris$Petal_Length
                      Df Sum Sq Mean Sq F value Pr(>F)
## as.factor(Species) 2 43710
                                  21855
                                          1180 <2e-16
## Residuals
                     147
                          2722
anova(aov(Iris$Petal_Width ~ as.factor(Species), data = Iris))
## Analysis of Variance Table
##
## Response: Iris$Petal_Width
                      Df Sum Sq Mean Sq F value Pr(>F)
## as.factor(Species) 2 8041 4021 960 <2e-16
                     147
## Residuals
                            616
# Use PCA to show potential clustering along two dimensions.
PCA <- principal(Iris[, 1:4], nfactors = 2, rotate = "varimax", scores = TRUE)
PCA
## Principal Components Analysis
## Call: principal(r = Iris[, 1:4], nfactors = 2, rotate = "varimax",
      scores = TRUE)
## Standardized loadings (pattern matrix) based upon correlation matrix
                 RC1 RC2 h2 u2 com
## Sepal_Length 0.96 0.05 0.92 0.0774 1.0
## Sepal_Width -0.14 0.98 0.99 0.0091 1.0
## Petal_Length 0.94 -0.30 0.98 0.0163 1.2
## Petal_Width 0.93 -0.26 0.94 0.0647 1.2
##
##
                         RC1 RC2
## SS loadings
                        2.70 1.13
## Proportion Var
                       0.68 0.28
## Cumulative Var
                        0.68 0.96
## Proportion Explained 0.71 0.29
## Cumulative Proportion 0.71 1.00
## Mean item complexity = 1.1
## Test of the hypothesis that 2 components are sufficient.
## The root mean square of the residuals (RMSR) is 0.03
## with the empirical chi square 1.72 with prob < NA
##
## Fit based upon off diagonal values = 1
Iris <- cbind(Iris, PCA$scores)</pre>
```

```
ggplot(Iris, aes(x = RC1, y = RC2, color = factor(Species))) + geom_point(shape = 19,
    size = 3) + scale_color_manual(values = c("red", "blue", "green")) +
    scale_y = continuous(breaks = c(seq(-3, 3.5, 0.5))) + scale_x = continuous(breaks = c(seq(-2, 3.5, 0.5)))
    (2.5, 0.5)) + coord_cartesian(xlim = c(-2, 2.5), ylim = c(-3,
    3.5)) + xlab("Component 1") + ylab("Component 2") + theme(text = element_text(size = 14,
    family = "sans", color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
    size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
    size = 12, face = "bold", angle = 0), axis.title.x = element_text(margin = margin(15,
    0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
    15, 0, 0), size = 16), axis.line.x = element_blank(), axis.line.y = element_blank(),
    plot.title = element_text(size = 16, face = "bold", margin = margin(0,
        0, 20, 0), hjust = 0.5), panel.background = element_rect(fill = "white",
        linetype = 1, color = "black"), panel.grid.major = element_blank(),
    panel.grid.minor = element_blank(), plot.background = element_rect(fill = "white"),
    plot.margin = unit(c(1, 1, 1, 1), "cm"), legend.position = "bottom",
    legend.title = element_blank()) + ggtitle("Component Plot by Species")
```

Component Plot by Species



3.2 Distance Calculation

```
# Use Euclidean distance for subsequent clustering.
Iris_Dist <- dist(Iris[, 1:4], method = "euclidean")</pre>
```

3.3 Clustering Methods

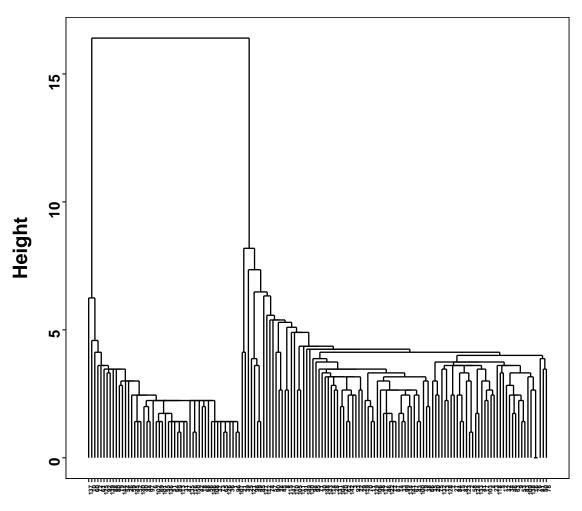
Each of the clustering methods can be applied to the iris data. Given the known structure of the data, the ability to recover the three species will help identify clustering methods that may be particularly useful.

3.3.1 Single Lingage

```
hc_1 <- hclust(Iris_Dist, method = "single")</pre>
clustnumber <- cutree(hc_1, k = 3)</pre>
Iris_Class <- as.data.frame(cbind(clustnumber, Iris$Species))</pre>
names(Iris_Class) <- c("Cluster", "Species")</pre>
table(Iris_Class$Species)
##
       Setosa Versicolor Virginica
##
##
         50 50
table(Iris_Class$Cluster)
##
## 1 2 3
## 50 98 2
table(Iris_Class$Species, Iris_Class$Cluster)
                 1 2 3
##
##
   Setosa
              50 0 0
    Versicolor 0 50 0
##
   Virginica 0 48 2
```

```
ggdendrogram(hc_1, theme_dendro = FALSE) + xlab("Iris Objects") +
   ylab("Height") + theme(text = element_text(size = 14, family = "sans",
   color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
   size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
   size = 5, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
   0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
   15, 0, 0), size = 16, angle = 90), axis.line.x = element_blank(),
   axis.line.y = element_blank(), plot.title = element_text(size = 16,
        face = "bold", margin = margin(0, 0, 20, 0), hjust = 0.5),
   panel.background = element_rect(fill = "white", linetype = 1,
        color = "black"), panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
   plot.background = element_rect(fill = "white"), plot.margin = unit(c(1,
        1, 1, 1), "cm"), legend.position = "bottom", legend.title = element_blank()) +
   ggtitle("Iris Cluster Dendogram: Single Linkage")
```

Iris Cluster Dendogram: Single Linkage



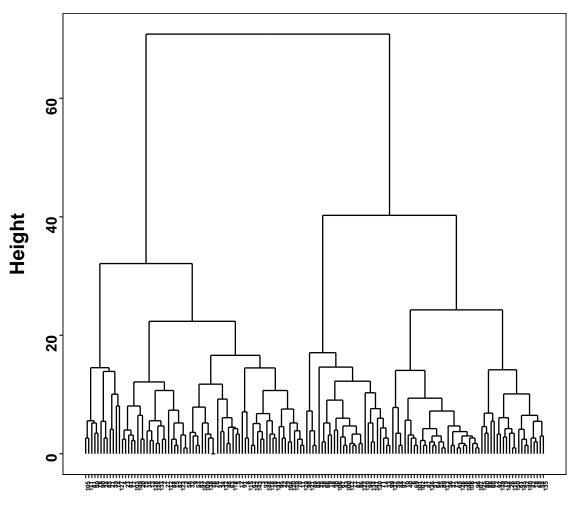
Iris Objects

3.3.2 Complete Lingage

```
hc_2 <- hclust(Iris_Dist, method = "complete")</pre>
clustnumber <- cutree(hc_2, k = 3)</pre>
Iris_Class <- as.data.frame(cbind(clustnumber, Iris$Species))</pre>
names(Iris_Class) <- c("Cluster", "Species")</pre>
table(Iris_Class$Species)
##
##
       Setosa Versicolor Virginica
          50
                 50
table(Iris_Class$Cluster)
##
## 1 2 3
## 50 72 28
table(Iris_Class$Species, Iris_Class$Cluster)
##
##
                 1 2 3
##
                50 0 0
   Setosa
   Versicolor 0 23 27
##
     Virginica 0 49 1
```

```
ggdendrogram(hc_2, theme_dendro = FALSE) + xlab("Iris Objects") +
   ylab("Height") + theme(text = element_text(size = 14, family = "sans",
   color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
   size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
   size = 5, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
   0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
   15, 0, 0), size = 16, angle = 90), axis.line.x = element_blank(),
   axis.line.y = element_blank(), plot.title = element_text(size = 16,
        face = "bold", margin = margin(0, 0, 20, 0), hjust = 0.5),
   panel.background = element_rect(fill = "white", linetype = 1,
        color = "black"), panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
   plot.background = element_rect(fill = "white"), plot.margin = unit(c(1,
        1, 1, 1), "cm"), legend.position = "bottom", legend.title = element_blank()) +
   ggtitle("Iris Cluster Dendogram: Complete Linkage")
```

Iris Cluster Dendogram: Complete Linkage



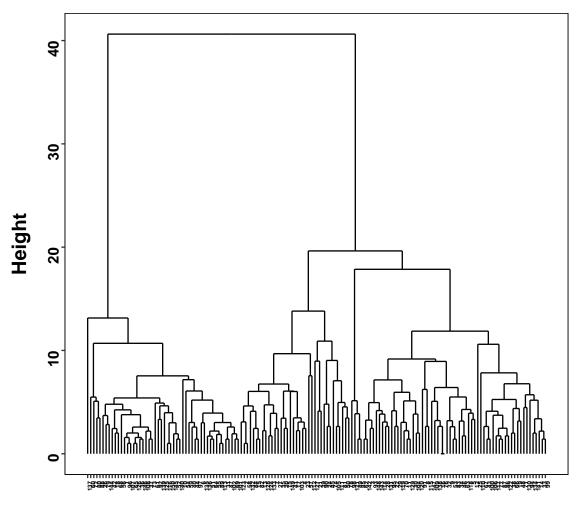
Iris Objects

3.3.3 Average Linkage

```
hc_3 <- hclust(Iris_Dist, method = "average")</pre>
clustnumber <- cutree(hc_3, k = 3)</pre>
Iris_Class <- as.data.frame(cbind(clustnumber, Iris$Species))</pre>
names(Iris_Class) <- c("Cluster", "Species")</pre>
table(Iris_Class$Species)
##
##
       Setosa Versicolor Virginica
          50
                  50
table(Iris_Class$Cluster)
##
## 1 2 3
## 50 36 64
table(Iris_Class$Species, Iris_Class$Cluster)
##
##
                 1 2 3
##
                50 0 0
   Setosa
   Versicolor 0 0 50
     Virginica 0 36 14
```

```
ggdendrogram(hc_3, theme_dendro = FALSE) + xlab("Iris Objects") +
   ylab("Height") + theme(text = element_text(size = 14, family = "sans",
   color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
   size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
   size = 5, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
   0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
   15, 0, 0), size = 16, angle = 90), axis.line.x = element_blank(),
   axis.line.y = element_blank(), plot.title = element_text(size = 16,
        face = "bold", margin = margin(0, 0, 20, 0), hjust = 0.5),
   panel.background = element_rect(fill = "white", linetype = 1,
        color = "black"), panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
   plot.background = element_rect(fill = "white"), plot.margin = unit(c(1,
        1, 1, 1), "cm"), legend.position = "bottom", legend.title = element_blank()) +
   ggtitle("Iris Cluster Dendogram: Average Linkage")
```

Iris Cluster Dendogram: Average Linkage



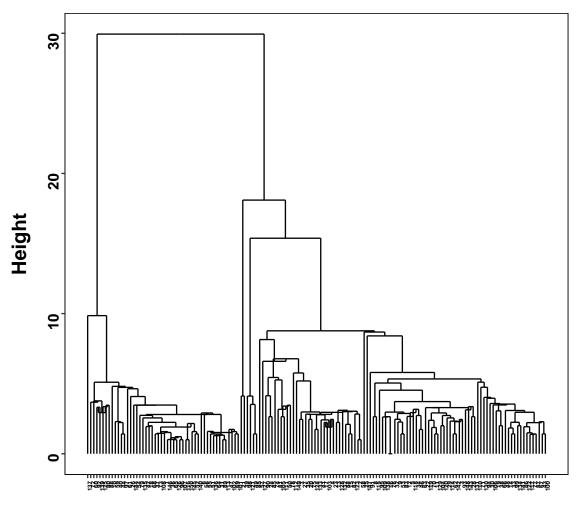
Iris Objects

3.3.4 Centroid Method

```
hc_4 <- hclust(Iris_Dist, method = "centroid")</pre>
clustnumber <- cutree(hc_4, k = 3)</pre>
Iris_Class <- as.data.frame(cbind(clustnumber, Iris$Species))</pre>
names(Iris_Class) <- c("Cluster", "Species")</pre>
table(Iris_Class$Species)
##
##
       Setosa Versicolor Virginica
         50
                 50
table(Iris_Class$Cluster)
##
## 1 2 3
## 50 98 2
table(Iris_Class$Species, Iris_Class$Cluster)
##
##
                 1 2 3
##
                50 0 0
   Setosa
   Versicolor 0 50 0
##
     Virginica 0 48 2
```

```
ggdendrogram(hc_4, theme_dendro = FALSE) + xlab("Iris Objects") +
   ylab("Height") + theme(text = element_text(size = 14, family = "sans",
   color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
   size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
   size = 5, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
   0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
   15, 0, 0), size = 16, angle = 90), axis.line.x = element_blank(),
   axis.line.y = element_blank(), plot.title = element_text(size = 16,
        face = "bold", margin = margin(0, 0, 20, 0), hjust = 0.5),
   panel.background = element_rect(fill = "white", linetype = 1,
        color = "black"), panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
   plot.background = element_rect(fill = "white"), plot.margin = unit(c(1,
        1, 1, 1), "cm"), legend.position = "bottom", legend.title = element_blank()) +
   ggtitle("Iris Cluster Dendogram: Centroid Method")
```

Iris Cluster Dendogram: Centroid Method



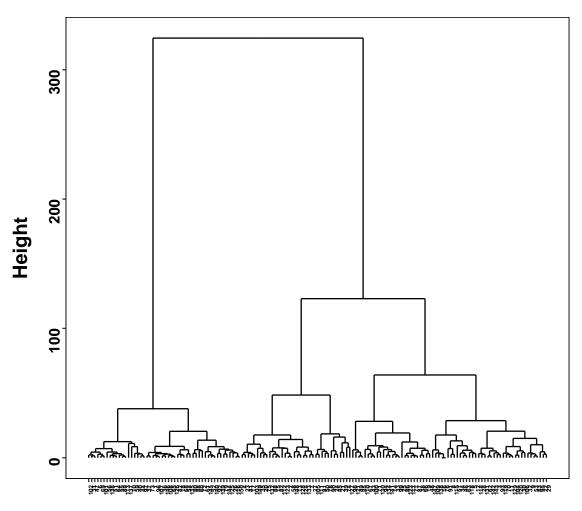
Iris Objects

3.3.5 Ward's Method (D2)

```
hc_5 <- hclust(Iris_Dist, method = "ward.D2")</pre>
clustnumber <- cutree(hc_5, k = 3)</pre>
Iris_Class <- as.data.frame(cbind(clustnumber, Iris$Species))</pre>
names(Iris_Class) <- c("Cluster", "Species")</pre>
table(Iris_Class$Species)
##
##
       Setosa Versicolor Virginica
          50
                 50
table(Iris_Class$Cluster)
##
## 1 2 3
## 50 36 64
table(Iris_Class$Species, Iris_Class$Cluster)
##
##
                 1 2 3
##
                50 0 0
   Setosa
   Versicolor 0 1 49
##
     Virginica 0 35 15
```

```
ggdendrogram(hc_5, theme_dendro = FALSE) + xlab("Iris Objects") +
   ylab("Height") + theme(text = element_text(size = 14, family = "sans",
   color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
   size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
   size = 5, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
   0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
   15, 0, 0), size = 16, angle = 90), axis.line.x = element_blank(),
   axis.line.y = element_blank(), plot.title = element_text(size = 16,
        face = "bold", margin = margin(0, 0, 20, 0), hjust = 0.5),
   panel.background = element_rect(fill = "white", linetype = 1,
        color = "black"), panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
   plot.background = element_rect(fill = "white"), plot.margin = unit(c(1,
        1, 1, 1), "cm"), legend.position = "bottom", legend.title = element_blank()) +
   ggtitle("Iris Cluster Dendogram: Ward's Method (D2)")
```

Iris Cluster Dendogram: Ward's Method (D2)



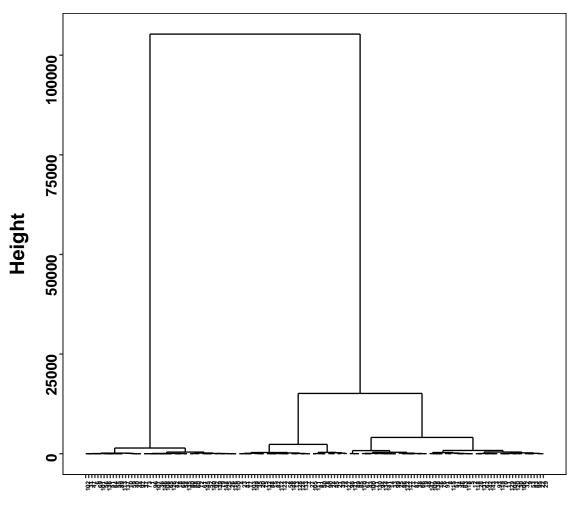
Iris Objects

3.3.6 Ward's Method (D)

```
Iris_Dist <- dist(Iris[, 1:4], method = "euclidean")^2</pre>
hc_5 <- hclust(Iris_Dist, method = "ward.D")</pre>
clustnumber <- cutree(hc_5, k = 3)</pre>
Iris_Class <- as.data.frame(cbind(clustnumber, Iris$Species))</pre>
names(Iris_Class) <- c("Cluster", "Species")</pre>
table(Iris_Class$Species)
##
       Setosa Versicolor Virginica
                      50
table(Iris_Class$Cluster)
##
## 1 2 3
## 50 36 64
table(Iris_Class$Species, Iris_Class$Cluster)
##
                 1 2 3
##
               50 0 0
##
     Setosa
##
    Versicolor 0 1 49
## Virginica 0 35 15
```

```
ggdendrogram(hc_5, theme_dendro = FALSE) + xlab("Iris Objects") +
   ylab("Height") + theme(text = element_text(size = 14, family = "sans",
   color = "black", face = "bold"), axis.text.y = element_text(colour = "black",
   size = 12, face = "bold"), axis.text.x = element_text(colour = "black",
   size = 5, face = "bold", angle = 90), axis.title.x = element_text(margin = margin(15,
   0, 0, 0), size = 16), axis.title.y = element_text(margin = margin(0,
   15, 0, 0), size = 16, angle = 90), axis.line.x = element_blank(),
   axis.line.y = element_blank(), plot.title = element_text(size = 16,
        face = "bold", margin = margin(0, 0, 20, 0), hjust = 0.5),
   panel.background = element_rect(fill = "white", linetype = 1,
        color = "black"), panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
   plot.background = element_rect(fill = "white"), plot.margin = unit(c(1,
        1, 1, 1), "cm"), legend.position = "bottom", legend.title = element_blank()) +
   ggtitle("Iris Cluster Dendogram: Ward's Method (D)")
```

Iris Cluster Dendogram: Ward's Method (D)



Iris Objects