

# Assignment 5

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```
library(factoextra)
```

```
## Loading required package: ggplot2
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
library(dendextend)
```

```
##
## -----
## Welcome to dendextend version 1.16.0
## Type citation('dendextend') for how to cite the package.
##
## Type browseVignettes(package = 'dendextend') for the package vignette.
## The github page is: https://github.com/talgalili/dendextend/
##
## Suggestions and bug-reports can be submitted at: https://github.com/talgalili/dendextend/issues
## You may ask questions at stackoverflow, use the r and dendextend tags:
## https://stackoverflow.com/questions/tagged/dendextend
##
## To suppress this message use: suppressPackageStartupMessages(library(dendextend))
## -----
```

```
##
## Attaching package: 'dendextend'
```

```
## The following object is masked from 'package:stats':
##
## cutree
```

```
library(cluster)
```

```
## Warning: package 'cluster' was built under R version 4.2.2
```

```
library(tidyverse)
```

```
## — Attaching packages
```

```
##
```

```
## tidyverse 1.3.2 —
```

```
## ✓ tibble 3.1.8    ✓ dplyr 1.0.10
```

```
## ✓ tidyr 1.2.1    ✓ stringr 1.4.1
```

```
## ✓ readr 2.1.3    ✓ forcats 0.5.2
```

```
## ✓ purrr 0.3.4
```

```
## — Conflicts ————— tidyverse_conflicts() —
```

```
## ✗ dplyr::filter() masks stats::filter()
```

```
## ✗ dplyr::lag()   masks stats::lag()
```

```
library(knitr)
```

```
cereals<- read.csv("C:/Users/jenru/OneDrive/Documents/Cereals.csv")
```

```
numericaldata = data.frame(cereals[,4:16])
```

```
OmitMissing = na.omit(numericaldata)
```

```
Normalise = scale(OmitMissing)
```

```
distance = dist(Normalise, method = "euclidian")
```

```
clustering_heirarchial = hclust(distance,method = "complete")
```

```
plot(clustering_heirarchial)
```

```
round(clustering_heirarchial$height, 3)
```

```
## [1] 0.143 0.196 0.575 0.698 0.828 0.904 1.003 1.004 1.201 1.203
```

```
## [11] 1.254 1.378 1.408 1.421 1.454 1.463 1.474 1.517 1.608 1.611
```

```
## [21] 1.616 1.625 1.650 1.687 1.692 1.720 1.730 1.795 1.839 1.897
```

```
## [31] 1.919 1.982 2.015 2.046 2.203 2.224 2.339 2.381 2.394 2.522
```

```
## [41] 2.563 2.574 2.579 2.668 2.682 2.734 2.776 2.787 3.229 3.236
```

```
## [51] 3.385 3.451 3.510 3.535 3.717 3.866 3.957 4.005 4.031 4.168
```

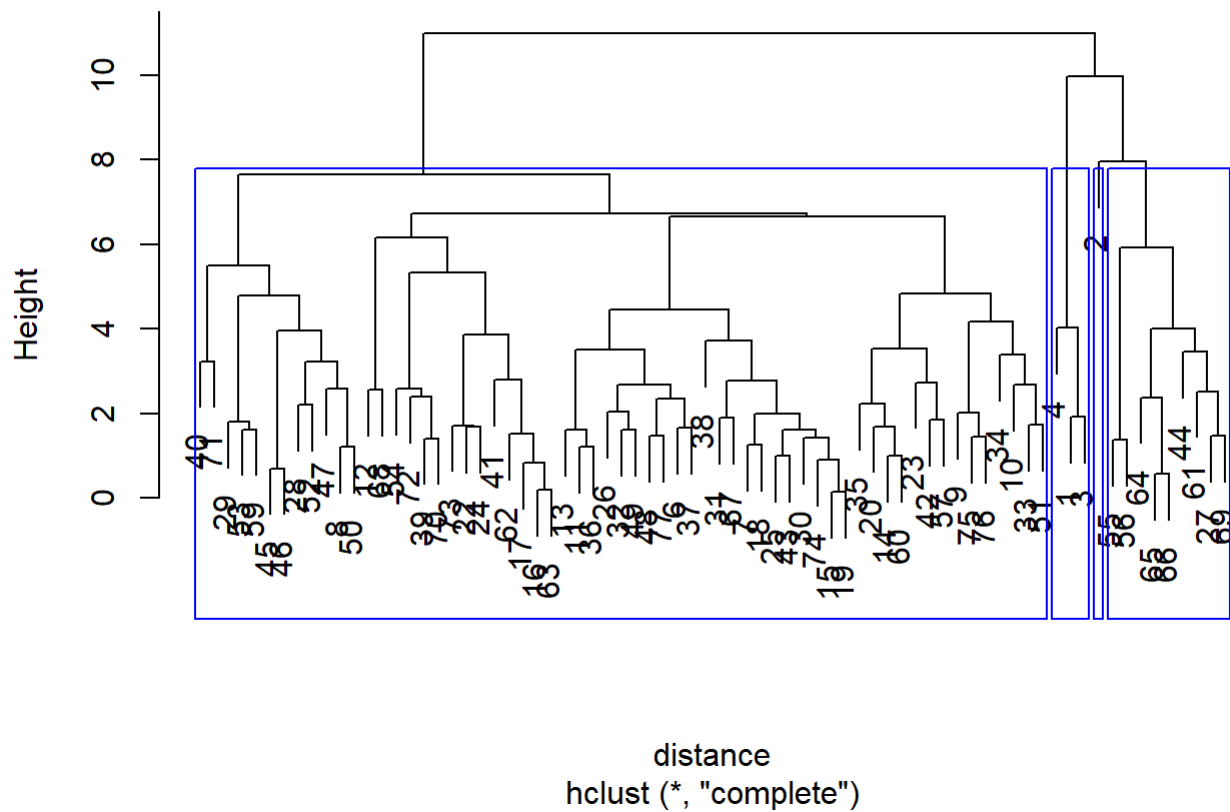
```
## [61] 4.456 4.779 4.839 5.342 5.488 5.920 6.169 6.669 6.731 7.650
```

```
## [71] 7.964 9.979 10.984
```

```
plot(clustering_heirarchial)
```

```
rect.hclust(clustering_heirarchial,k = 4, border = "blue")
```

## Cluster Dendrogram



```
singleClustering = agnes(Normalise, method = "single")
completeClustering = agnes(Normalise, method = "complete")
averageClustering = agnes(Normalise, method = "average")
wardClustering = agnes(Normalise, method = "ward")

print(singleClustering$ac)
```

```
## [1] 0.6067859
```

```
print(completeClustering$ac)
```

```
## [1] 0.8353712
```

```
print(averageClustering$ac)
```

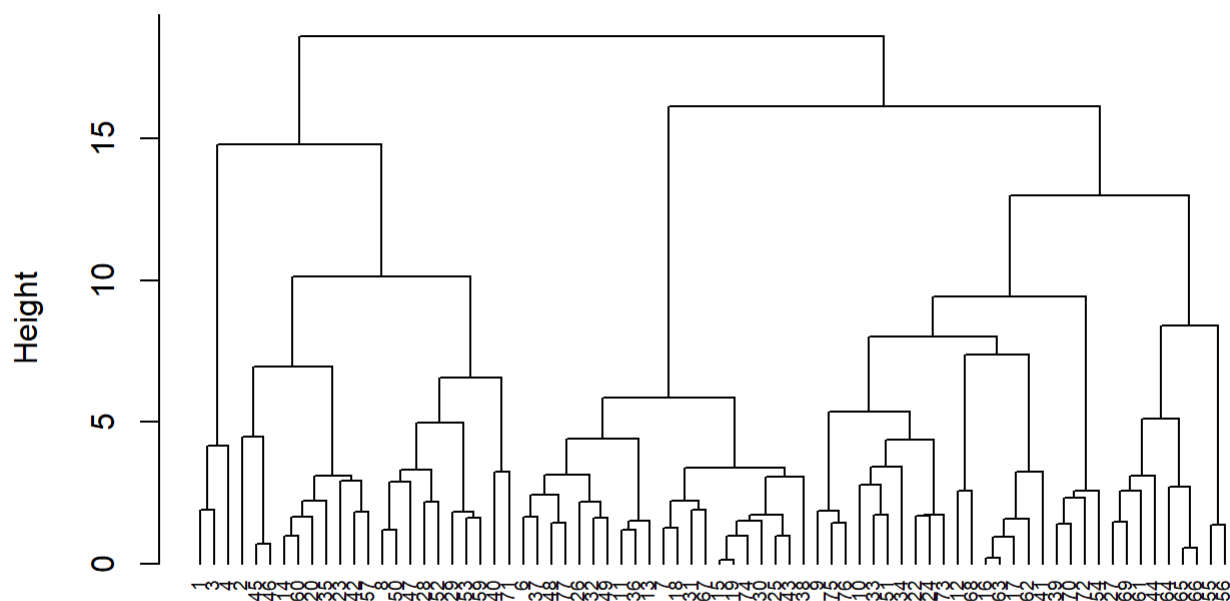
```
## [1] 0.7766075
```

```
print(wardClustering$ac)
```

```
## [1] 0.9046042
```

```
pltree(wardClustering, cex = 0.6, hang = -1, main = "Dendrogram of Agnes")
```

## Dendrogram of Agnes



## Normalise agnes (\*, "ward")

```
#Per the values above, Wards method is the best since the value is: 0.9046042
```

```
HierarchicalClustering1 <- hclust(distance, method = "ward.D2" )
subgrp <- cutree(HierarchicalClustering1, k = 4)
table(subgrp)
```

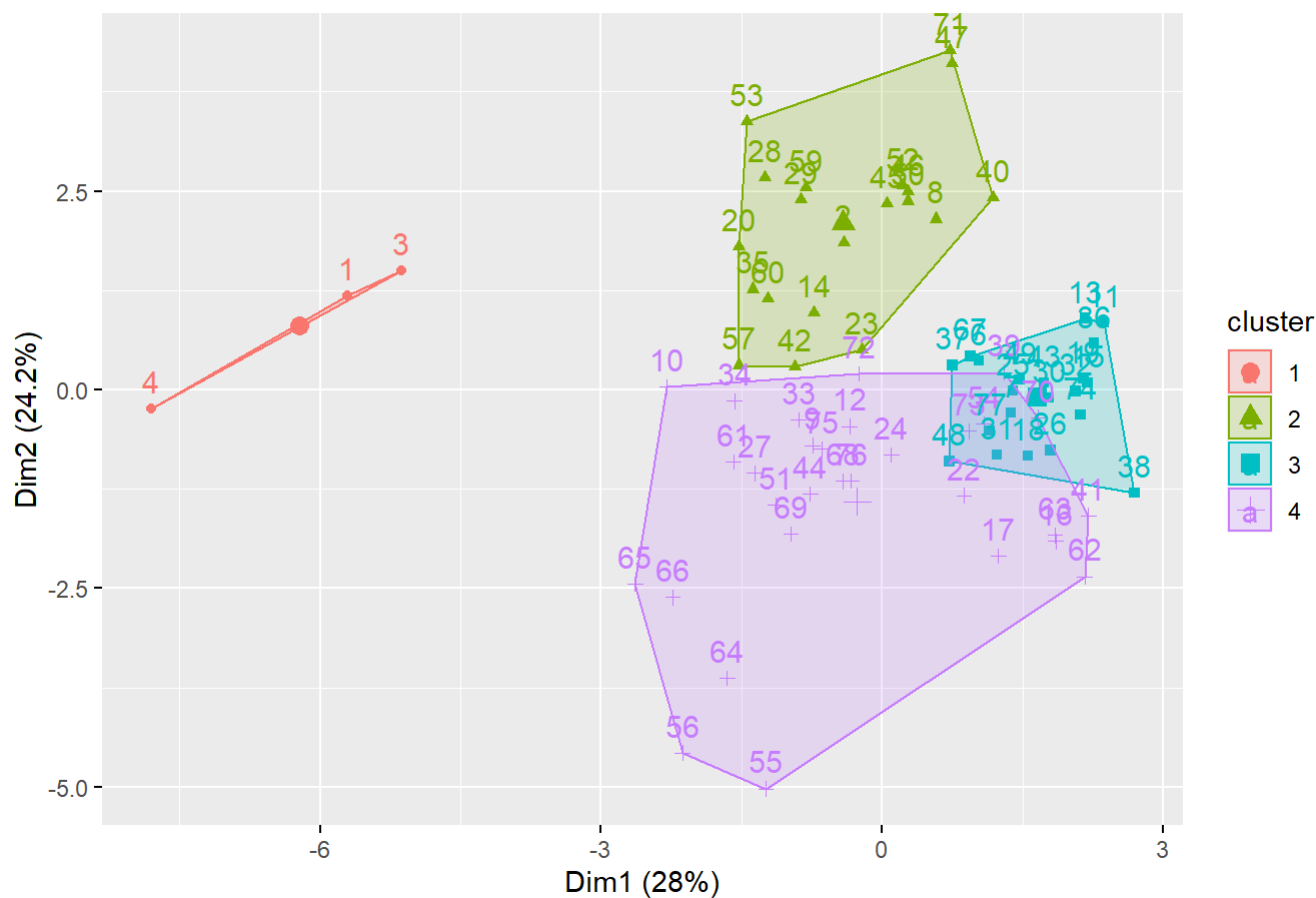
```
## subgrp
## 1 2 3 4
## 3 20 21 30
```

```
cereals <- as.data.frame(cbind(Normalise,subgrp))
```

```
#Visualizing the results on a scatterplot:
```

```
fviz_cluster(list(data = Normalise, cluster = subgrp))
```

## Cluster plot



```
Newcereals = numericaldata
Newcereals_omit = na.omit(Newcereals)
Clust = cbind(Newcereals_omit, subgrp)
Clust[Clust$subgrp==1,]
```

```
## calories protein fat sodium fiber carbo sugars potass vitamins shelf weight
## 1 70 4 1 130 10 5 6 280 25 3 1
## 3 70 4 1 260 9 7 5 320 25 3 1
## 4 50 4 0 140 14 8 0 330 25 3 1

## cups rating subgrp
## 1 0.33 68.40297 1
## 3 0.33 59.42551 1
## 4 0.50 93.70491 1
```

```
Clust[Clust$subgrp==2,]
```

```
## calories protein fat sodium fiber carbo sugars potass vitamins shelf weight
```

```
## 2 120 3 5 15 2.0 8.0 8 135 0 3 1.00
## 8 130 3 2 210 2.0 18.0 8 100 25 3 1.33
## 14 110 3 2 140 2.0 13.0 7 105 25 3 1.00
## 20 110 3 3 140 4.0 10.0 7 160 25 3 1.00
## 23 100 2 1 140 2.0 11.0 10 120 25 3 1.00
## 28 120 3 2 160 5.0 12.0 10 200 25 3 1.25
## 29 120 3 0 240 5.0 14.0 12 190 25 3 1.33
## 35 120 3 3 75 3.0 13.0 4 100 25 3 1.00
## 40 140 3 1 170 2.0 20.0 9 95 100 3 1.30
## 42 100 4 2 150 2.0 12.0 6 95 25 2 1.00
## 45 150 4 3 95 3.0 16.0 11 170 25 3 1.00
## 46 150 4 3 150 3.0 16.0 11 170 25 3 1.00
## 47 160 3 2 150 3.0 17.0 13 160 25 3 1.50
## 50 140 3 2 220 3.0 21.0 7 130 25 3 1.33
## 52 130 3 2 170 1.5 13.5 10 120 25 3 1.25
## 53 120 3 1 200 6.0 11.0 14 260 25 3 1.33
## 57 100 4 1 135 2.0 14.0 6 110 25 3 1.00
## 59 120 3 1 210 5.0 14.0 12 240 25 2 1.33
## 60 100 3 2 140 2.5 10.5 8 140 25 3 1.00
## 71 140 3 1 190 4.0 15.0 14 230 100 3 1.50
```

```
## cups rating subgrp
```

```
## 2 1.00 33.98368 2
## 8 0.75 37.03856 2
## 14 0.50 40.40021 2
## 20 0.50 40.44877 2
## 23 0.75 36.17620 2
## 28 0.67 40.91705 2
## 29 0.67 41.01549 2
## 35 0.33 45.81172 2
## 40 0.75 36.47151 2
## 42 0.67 45.32807 2
## 45 1.00 37.13686 2
## 46 1.00 34.13976 2
## 47 0.67 30.31335 2
## 50 0.67 40.69232 2
## 52 0.50 30.45084 2
## 53 0.67 37.84059 2
## 57 0.50 49.51187 2
## 59 0.75 39.25920 2
## 60 0.50 39.70340 2
## 71 1.00 28.59278 2
```

```
Clust[Clust$subgrp==3,]
```

```
## calories protein fat sodium fiber carbo sugars potass vitamins shelf weight
```

```
## 6 110 2 2 180 1.5 10.5 10 70 25 1 1
## 7 110 2 0 125 1.0 11.0 14 30 25 2 1
## 11 120 1 2 220 0.0 12.0 12 35 25 2 1
## 13 120 1 3 210 0.0 13.0 9 45 25 2 1
## 15 110 1 1 180 0.0 12.0 13 55 25 2 1
## 18 110 1 0 90 1.0 13.0 12 20 25 2 1
## 19 110 1 1 180 0.0 12.0 13 65 25 2 1
## 25 110 2 1 125 1.0 11.0 13 30 25 2 1
## 26 110 1 0 200 1.0 14.0 11 25 25 1 1
## 30 110 1 1 135 0.0 13.0 12 25 25 2 1
## 31 100 2 0 45 0.0 11.0 15 40 25 1 1
## 32 110 1 1 280 0.0 15.0 9 45 25 2 1
## 36 120 1 2 220 1.0 12.0 11 45 25 2 1
## 37 110 3 1 250 1.5 11.5 10 90 25 1 1
## 38 110 1 0 180 0.0 14.0 11 35 25 1 1
## 43 110 2 1 180 0.0 12.0 12 55 25 2 1
## 48 100 2 1 220 2.0 15.0 6 90 25 1 1
## 49 120 2 1 190 0.0 15.0 9 40 25 2 1
## 67 110 2 1 70 1.0 9.0 15 40 25 2 1
## 74 110 1 1 140 0.0 13.0 12 25 25 2 1
## 77 110 2 1 200 1.0 16.0 8 60 25 1 1
```

```
## cups rating subgrp
```

```
## 6 0.75 29.50954 3
## 7 1.00 33.17409 3
## 11 0.75 18.04285 3
## 13 0.75 19.82357 3
## 15 1.00 22.73645 3
## 18 1.00 35.78279 3
## 19 1.00 22.39651 3
## 25 1.00 32.20758 3
## 26 0.75 31.43597 3
## 30 0.75 28.02576 3
## 31 0.88 35.25244 3
## 32 0.75 23.80404 3
## 36 1.00 21.87129 3
## 37 0.75 31.07222 3
## 38 1.33 28.74241 3
## 43 1.00 26.73451 3
## 48 1.00 40.10596 3
## 49 0.67 29.92429 3
## 67 0.75 31.23005 3
## 74 1.00 27.75330 3
## 77 0.75 36.18756 3
```

```
Clust[Clust$subgrp==4,]
```

```
## calories protein fat sodium fiber carbo sugars potass vitamins shelf weight
```

```
## 9 90 2 1 200 4 15 6 125 25 1 1.00
## 10 90 3 0 210 5 13 5 190 25 3 1.00
## 12 110 6 2 290 2 17 1 105 25 1 1.00
## 16 110 2 0 280 0 22 3 25 25 1 1.00
## 17 100 2 0 290 1 21 2 35 25 1 1.00
## 22 110 2 0 220 1 21 3 30 25 3 1.00
## 24 100 2 0 190 1 18 5 80 25 3 1.00
## 27 100 3 0 0 3 14 7 100 25 2 1.00
## 33 100 3 1 140 3 15 5 85 25 3 1.00
## 34 110 3 0 170 3 17 3 90 25 3 1.00
## 39 110 2 1 170 1 17 6 60 100 3 1.00
## 41 110 2 1 260 0 21 3 40 25 2 1.00
## 44 100 4 1 0 0 16 3 95 25 2 1.00
## 51 90 3 0 170 3 18 2 90 25 3 1.00
## 54 100 3 0 320 1 20 3 45 100 3 1.00
## 55 50 1 0 0 0 13 0 15 0 3 0.50
## 56 50 2 0 0 1 10 0 50 0 3 0.50
## 61 90 2 0 0 2 15 6 110 25 3 1.00
## 62 110 1 0 240 0 23 2 30 25 1 1.00
## 63 110 2 0 290 0 22 3 35 25 1 1.00
## 64 80 2 0 0 3 16 0 95 0 1 0.83
## 65 90 3 0 0 4 19 0 140 0 1 1.00
## 66 90 3 0 0 3 20 0 120 0 1 1.00
## 68 110 6 0 230 1 16 3 55 25 1 1.00
## 69 90 2 0 15 3 15 5 90 25 2 1.00
## 70 110 2 1 200 0 21 3 35 100 3 1.00
## 72 100 3 1 200 3 16 3 110 100 3 1.00
## 73 110 2 1 250 0 21 3 60 25 3 1.00
## 75 100 3 1 230 3 17 3 115 25 1 1.00
## 76 100 3 1 200 3 17 3 110 25 1 1.00
```

```
## cups rating subgrp
```

```
## 9 0.67 49.12025 4
## 10 0.67 53.31381 4
## 12 1.25 50.76500 4
## 16 1.00 41.44502 4
## 17 1.00 45.86332 4
## 22 1.00 46.89564 4
## 24 0.75 44.33086 4
## 27 0.80 58.34514 4
## 33 0.88 52.07690 4
## 34 0.25 53.37101 4
## 39 1.00 36.52368 4
## 41 1.50 39.24111 4
## 44 1.00 54.85092 4
## 51 1.00 59.64284 4
## 54 1.00 41.50354 4
## 55 1.00 60.75611 4
## 56 1.00 63.00565 4
## 61 0.50 55.33314 4
## 62 1.13 41.99893 4
## 63 1.00 40.56016 4
```



```
## 64 1.00 68.23588 4
## 65 0.67 74.47295 4
## 66 0.67 72.80179 4
## 68 1.00 53.13132 4
## 69 1.00 59.36399 4
## 70 1.00 38.83975 4
## 72 1.00 46.65884 4
## 73 0.75 39.10617 4
## 75 0.67 49.78744 4
## 76 1.00 51.59219 4
```

*#Calculating mean rating in order for cluster of healthy cereals*

```
mean(Clust[Clust$subgrp==1,"rating"])
```

```
## [1] 73.84446
```

```
mean(Clust[Clust$subgrp==2,"rating"])
```

```
## [1] 38.26161
```

```
mean(Clust[Clust$subgrp==3,"rating"])
```

```
## [1] 28.84825
```

```
mean(Clust[Clust$subgrp==4,"rating"])
```

```
## [1] 51.43111
```

*#As we see through calculating the mean rating, subgroup 1 has the highest. It is recommended to choose subgroup1 as a healthy cluster*