

Perform Hierarchical clustering

Harini G

```
#NAME: Harini G

#install.packages("tidyverse")
#install.packages("factoextra")
library(tidyverse) # data manipulation

## Warning: package 'tidyverse' was built under R version 4.0.5

## -- Attaching packages ----- tidyverse
1.3.0 --

## v ggplot2 3.3.3      v purrr  0.3.4
## v tibble  3.0.3      v dplyr  1.0.5
## v tidyr   1.1.1      v stringr 1.4.0
## v readr   1.4.0      v forcats 0.5.0

## Warning: package 'dplyr' was built under R version 4.0.5

## -- Conflicts -----
tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(cluster) # clustering algorithms
library(factoextra) # clustering visualization

## Warning: package 'factoextra' was built under R version 4.0.5

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

library(dendextend) # for comparing two dendrograms

## Warning: package 'dendextend' was built under R version 4.0.5

##
## -----
## Welcome to dendextend version 1.14.0
## 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
## Or contact: <tal.galili@gmail.com>
##
## To suppress this message use:
```

```

suppressPackageStartupMessages(library(dendextend))
## -----

##
## Attaching package: 'dendextend'

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

head(mtcars)

##              mpg  cyl  disp  hp  drat    wt   qsec vs  am gear carb
## Mazda RX4      21.0   6  160 110  3.90  2.620 16.46  0  1   4    4
## Mazda RX4 Wag  21.0   6  160 110  3.90  2.875 17.02  0  1   4    4
## Datsun 710     22.8   4  108  93  3.85  2.320 18.61  1  1   4    1
## Hornet 4 Drive  21.4   6  258 110  3.08  3.215 19.44  1  0   3    1
## Hornet Sportabout 18.7   8  360 175  3.15  3.440 17.02  0  0   3    2
## Valiant        18.1   6  225 105  2.76  3.460 20.22  1  0   3    1

class(mtcars) #class of the dataset

## [1] "data.frame"

typeof(mtcars) #datatype of the dataset

## [1] "list"

nrow(mtcars) #number of rows/observations in the dataset

## [1] 32

ncol(mtcars) #number of columns/features in the dataset

## [1] 11

dim(mtcars) #Shape/dimension of the dataset

## [1] 32 11

is.na(mtcars) #checking missing value in the dataset

##              mpg   cyl  disp    hp  drat    wt   qsec    vs  am
gear
## Mazda RX4      FALSE FALSE FALSE  FALSE  FALSE  FALSE  FALSE  FALSE  FALSE
FALSE
## Mazda RX4 Wag  FALSE FALSE FALSE  FALSE  FALSE  FALSE  FALSE  FALSE  FALSE
FALSE
## Datsun 710     FALSE FALSE FALSE  FALSE  FALSE  FALSE  FALSE  FALSE  FALSE
FALSE
## Hornet 4 Drive  FALSE FALSE FALSE  FALSE  FALSE  FALSE  FALSE  FALSE  FALSE
FALSE
## Hornet Sportabout FALSE FALSE FALSE  FALSE  FALSE  FALSE  FALSE  FALSE  FALSE

```

FALSE	
## Valiant	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Duster 360	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Merc 240D	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Merc 230	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Merc 280	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Merc 280C	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Merc 450SE	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Merc 450SL	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Merc 450SLC	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Cadillac Fleetwood	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Lincoln Continental	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Chrysler Imperial	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Fiat 128	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Honda Civic	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Toyota Corolla	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Toyota Corona	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Dodge Challenger	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## AMC Javelin	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Camaro Z28	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Pontiac Firebird	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Fiat X1-9	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Porsche 914-2	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Lotus Europa	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Ford Pantera L	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE	
## Ferrari Dino	FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE

```

FALSE
## Maserati Bora      FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE
## Volvo 142E        FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE
FALSE
##                  carb
## Mazda RX4         FALSE
## Mazda RX4 Wag     FALSE
## Datsun 710         FALSE
## Hornet 4 Drive    FALSE
## Hornet Sportabout FALSE
## Valiant           FALSE
## Duster 360        FALSE
## Merc 240D         FALSE
## Merc 230          FALSE
## Merc 280          FALSE
## Merc 280C         FALSE
## Merc 450SE        FALSE
## Merc 450SL        FALSE
## Merc 450SLC       FALSE
## Cadillac Fleetwood FALSE
## Lincoln Continental FALSE
## Chrysler Imperial FALSE
## Fiat 128          FALSE
## Honda Civic       FALSE
## Toyota Corolla    FALSE
## Toyota Corona     FALSE
## Dodge Challenger  FALSE
## AMC Javelin       FALSE
## Camaro Z28        FALSE
## Pontiac Firebird  FALSE
## Fiat X1-9         FALSE
## Porsche 914-2     FALSE
## Lotus Europa      FALSE
## Ford Pantera L    FALSE
## Ferrari Dino      FALSE
## Maserati Bora     FALSE
## Volvo 142E        FALSE

```

```
sum(is.na(mtcars)) #total number of missing values in the dataset
```

```
## [1] 0
```

```
str(mtcars) #Structure of the dataset
```

```

## 'data.frame':   32 obs. of  11 variables:
##  $ mpg : num  21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
##  $ cyl : num   6  6  4  6  8  6  8  4  4  6 ...
##  $ disp: num  160 160 108 258 360 ...
##  $ hp  : num  110 110 93 110 175 105 245 62 95 123 ...
##  $ drat: num   3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...

```

```
## $ wt : num 2.62 2.88 2.32 3.21 3.44 ...
## $ qsec: num 16.5 17 18.6 19.4 17 ...
## $ vs : num 0 0 1 1 0 1 0 1 1 1 ...
## $ am : num 1 1 1 0 0 0 0 0 0 0 ...
## $ gear: num 4 4 4 3 3 3 3 4 4 4 ...
## $ carb: num 4 4 1 1 2 1 4 2 2 4 ...
```

```
summary(mtcars)
```

```
##      mpg          cyl          disp          hp
## Min.   :10.40    Min.   :4.000    Min.   : 71.1    Min.   : 52.0
## 1st Qu.:15.43    1st Qu.:4.000    1st Qu.:120.8    1st Qu.: 96.5
## Median :19.20    Median :6.000    Median :196.3    Median :123.0
## Mean   :20.09    Mean   :6.188    Mean   :230.7    Mean   :146.7
## 3rd Qu.:22.80    3rd Qu.:8.000    3rd Qu.:326.0    3rd Qu.:180.0
## Max.   :33.90    Max.   :8.000    Max.   :472.0    Max.   :335.0
##      drat          wt          qsec          vs
## Min.   :2.760    Min.   :1.513    Min.   :14.50    Min.   :0.0000
## 1st Qu.:3.080    1st Qu.:2.581    1st Qu.:16.89    1st Qu.:0.0000
## Median :3.695    Median :3.325    Median :17.71    Median :0.0000
## Mean   :3.597    Mean   :3.217    Mean   :17.85    Mean   :0.4375
## 3rd Qu.:3.920    3rd Qu.:3.610    3rd Qu.:18.90    3rd Qu.:1.0000
## Max.   :4.930    Max.   :5.424    Max.   :22.90    Max.   :1.0000
##      am          gear          carb
## Min.   :0.0000    Min.   :3.000    Min.   :1.000
## 1st Qu.:0.0000    1st Qu.:3.000    1st Qu.:2.000
## Median :0.0000    Median :4.000    Median :2.000
## Mean   :0.4062    Mean   :3.688    Mean   :2.812
## 3rd Qu.:1.0000    3rd Qu.:4.000    3rd Qu.:4.000
## Max.   :1.0000    Max.   :5.000    Max.   :8.000
```

```
sapply(mtcars, class)#displaying the datatype of each column
```

```
##      mpg      cyl      disp      hp      drat      wt      qsec
vs
## "numeric" "numeric" "numeric" "numeric" "numeric" "numeric" "numeric"
"numeric"
##      am      gear      carb
## "numeric" "numeric" "numeric"
```

```
df <- na.omit(mtcars)#removing missing values if they are present
```

```
df <- scale(df)
```

#It works in a bottom-up manner.

#each object is initially considered as a single-element cluster (leaf).

#At each step of the algorithm, the two clusters that are the most similar are combined into a new bigger cluster (nodes).

#This procedure is iterated until all points are member of just one single big cluster (root).

#The result is a tree which is plotted using dendrogram.

```
d <- dist(df, method = "euclidean")# Dissimilarity matrix
d
```

```
## Mazda RX4 Mazda RX4 Wag Datsun 710 Hornet 4 Drive
## Mazda RX4 Wag 0.4075899
## Datsun 710 3.2430644 3.1763654
## Hornet 4 Drive 4.4013651 4.2633265 3.4371367
## Hornet Sportabout 3.8803542 3.8196912 5.0032747 3.0421632
## Valiant 4.8437395 4.6756447 3.8681280 0.9936969
## Duster 360 4.1895788 4.1749365 5.8959064 4.3395668
## Merc 240D 3.9972560 3.8208496 2.5014249 2.5336229
## Merc 230 4.9177375 4.6700230 3.3122031 3.2698916
## Merc 280 3.1377712 2.9882339 3.2950024 2.9859746
## Merc 280C 3.2928005 3.1170530 3.3443599 2.9705073
## Merc 450SE 3.8563035 3.7329721 5.1667877 3.2468885
## Merc 450SL 3.7264672 3.6140741 5.0139378 3.0963299
## Merc 450SLC 3.8587627 3.7280140 5.0836872 3.1350053
## Cadillac Fleetwood 5.4495167 5.2848094 6.7701575 4.6845022
## Lincoln Continental 5.4799639 5.3127593 6.8196686 4.7868961
## Chrysler Imperial 5.0972757 4.9355861 6.5250751 4.6034009
## Fiat 128 4.0243306 3.9407293 1.7832069 4.1853934
## Honda Civic 4.0533412 4.0507443 2.6458745 5.2450434
## Toyota Corolla 4.3445000 4.2722835 2.2120337 4.5513874
## Toyota Corona 4.3303364 4.2137101 2.5743286 2.1359358
## Dodge Challenger 4.1089579 4.0529524 5.1794615 3.1033485
## AMC Javelin 3.7602309 3.6846550 4.8499144 2.9346504
## Camaro Z28 4.1191859 4.1051443 5.9277646 4.6740373
## Pontiac Firebird 4.1721471 4.0882070 5.2718706 3.1815173
## Fiat X1-9 3.6110861 3.5658305 1.0646650 3.9829642
## Porsche 914-2 2.5948592 2.6591243 2.9668981 5.3633226
## Lotus Europa 3.5593956 3.6429111 2.3504814 4.8654619
## Ford Pantera L 3.6239136 3.7004541 5.6724603 6.2182516
## Ferrari Dino 2.2173337 2.3107383 4.6711840 5.7255126
## Maserati Bora 4.9757567 5.0070559 7.2893098 7.4540115
## Volvo 142E 2.9056273 2.7880036 0.9799181 3.5836208
## Hornet Sportabout Valiant Duster 360 Merc 240D Merc
230
## Mazda RX4 Wag
## Datsun 710
## Hornet 4 Drive
## Hornet Sportabout
## Valiant 3.3988179
## Duster 360 1.8907332 4.5959496
## Merc 240D 4.6074904 2.9114013 5.6208325
## Merc 230 5.3601882 3.4090956 6.3074270 1.7677287
## Merc 280 3.7603055 3.3860393 4.1159325 2.3467260
3.1736664
## Merc 280C 3.8404534 3.2902323 4.2017096 2.3314789
2.9644001
```

## Merc 450SE 5.2829002	1.2157212 3.3739087 1.7195474 4.6100696
## Merc 450SL 5.1385834	1.0581618 3.2662012 1.7483591 4.4789673
## Merc 450SLC 5.0939035	1.2772271 3.2103732 1.8280213 4.5289442
## Cadillac Fleetwood 6.4719931	2.8840440 4.6250214 2.4964362 6.0571351
## Lincoln Continental 6.5342344	2.9684612 4.7409039 2.4967056 6.1200901
## Chrysler Imperial 6.3195516	2.6262007 4.7149888 2.1314298 5.8475038
## Fiat 128 3.4760961	5.8776367 4.7310845 6.9265472 2.8477338
## Honda Civic 4.2793565	6.5054903 5.8852758 7.2973276 3.7432868
## Toyota Corolla 3.6385061	6.2330226 5.1096664 7.2823316 3.1877164
## Toyota Corona 2.3602221	4.4095161 2.5379207 5.4586640 1.8187717
## Dodge Challenger 5.5840526	1.0370480 3.2050198 2.1565615 4.7681748
## AMC Javelin 5.1708093	0.8360206 3.1377397 2.0925400 4.4434196
## Camaro Z28 6.3746076	2.3538211 5.0148655 1.0554323 5.7074187
## Pontiac Firebird 5.5292039	0.5475333 3.5076094 1.9842791 4.7856750
## Fiat X1-9 3.4679554	5.6028144 4.4964851 6.5754940 2.7365898
## Porsche 914-2 4.9107991	5.5333206 5.9048663 6.1678160 4.0696390
## Lotus Europa 4.6690572	5.8906362 5.4091213 6.5673285 3.6797302
## Ford Pantera L 7.0037774	4.5704575 6.6597531 3.9756842 6.3426475
## Ferrari Dino 6.0347907	4.8737371 6.0393960 4.5054188 5.2004290
## Maserati Bora 8.1527622	5.7962447 7.6982401 4.5080940 7.6028670
## Volvo 142E 3.1932948	4.9104611 4.0221723 5.5902795 2.4866260
##	Merc 280 Merc 280C Merc 450SE Merc 450SL Merc 450SLC
## Mazda RX4 Wag	
## Datsun 710	
## Hornet 4 Drive	
## Hornet Sportabout	
## Valiant	
## Duster 360	
## Merc 240D	

## Merc 230					
## Merc 280					
## Merc 280C	0.4082884				
## Merc 450SE	3.5036558	3.5446708			
## Merc 450SL	3.4252318	3.4663403	0.3944266		
## Merc 450SLC	3.4613499	3.4571735	0.4901305	0.4172832	
## Cadillac Fleetwood	4.8481443	4.8076108	2.3752167	2.6191633	2.4458852
## Lincoln Continental	4.8652211	4.8310582	2.4147746	2.6820009	2.5174149
## Chrysler Imperial	4.5331791	4.5497518	2.1301591	2.3833766	2.3303195
## Fiat 128	4.1044898	4.1942047	6.1027421	5.9176565	6.0522758
## Honda Civic	4.3521695	4.4596794	6.7429222	6.5608536	6.6876038
## Toyota Corolla	4.4495050	4.5273217	6.4894198	6.2849281	6.4199329
## Toyota Corona	3.0091888	2.9591531	4.5230838	4.3467458	4.3683285
## Dodge Challenger	3.9899083	4.0420607	1.2054395	1.1528213	1.2246720
## AMC Javelin	3.5713821	3.6040168	1.0549265	0.9430857	0.9626504
## Camaro Z28	4.0282873	4.1261450	2.1735238	2.2625903	2.3169202
## Pontiac Firebird	3.9725287	4.0518437	1.3039453	1.2653565	1.4562704
## Fiat X1-9	3.7810906	3.8551580	5.8249673	5.6428137	5.7474798
## Porsche 914-2	4.2059850	4.3568869	5.7510348	5.6099124	5.7482820
## Lotus Europa	4.1265650	4.3109976	6.1480650	5.9587741	6.1321085
## Ford Pantera L	4.7638360	4.9017138	4.6618197	4.6520853	4.7410691
## Ferrari Dino	3.8958003	4.0539793	4.6304461	4.5549676	4.6718041
## Maserati Bora	5.7612264	5.8672946	5.4444617	5.4646796	5.5262308
## Volvo 142E	2.7801612	2.8200779	4.9479927	4.8294785	4.8806503
##	Cadillac Fleetwood Lincoln Continental Chrysler				
Imperial					
## Mazda RX4 Wag					
## Datsun 710					
## Hornet 4 Drive					
## Hornet Sportabout					
## Valiant					
## Duster 360					
## Merc 240D					
## Merc 230					
## Merc 280					
## Merc 280C					
## Merc 450SE					
## Merc 450SL					
## Merc 450SLC					
## Cadillac Fleetwood					
## Lincoln Continental	0.2956825				
## Chrysler Imperial	1.0635310		0.9080748		
## Fiat 128	7.7973439		7.8458703		
7.4347239					
## Honda Civic	8.3928547		8.4234897		
7.9702830					
## Toyota Corolla	8.2005281		8.2527181		
7.8280569					
## Toyota Corona	6.1484614		6.2160567		
5.9697480					

## Dodge Challenger	2.8372165	2.9549058		
2.8683778				
## AMC Javelin	2.8988638	2.9910214		
2.8183456				
## Camaro Z28	2.8150659	2.7412463		
2.2659922				
## Pontiac Firebird	2.5701272	2.6555931		
2.3043685				
## Fiat X1-9	7.5268299	7.5797064		
7.2243403				
## Porsche 914-2	7.3428177	7.3625764		
6.9369982				
## Lotus Europa	7.8314188	7.8778088		
7.4655971				
## Ford Pantera L	5.2908702	5.2360533		
4.8165342				
## Ferrari Dino	5.8052611	5.8069129		
5.4553770				
## Maserati Bora	5.6040962	5.5359080		
5.2173124				
## Volvo 142E	6.4042902	6.4299677		
6.1099484				
##				
	Fiat 128	Honda Civic	Toyota Corolla	Toyota Corona
## Mazda RX4 Wag				
## Datsun 710				
## Hornet 4 Drive				
## Hornet Sportabout				
## Valiant				
## Duster 360				
## Merc 240D				
## Merc 230				
## Merc 280				
## Merc 280C				
## Merc 450SE				
## Merc 450SL				
## Merc 450SLC				
## Cadillac Fleetwood				
## Lincoln Continental				
## Chrysler Imperial				
## Fiat 128				
## Honda Civic	1.9243376			
## Toyota Corolla	0.5757917	1.7799297		
## Toyota Corona	3.1795988	3.9646286	3.4386884	
## Dodge Challenger	6.1897572	6.9208270	6.5812246	4.5624114
## AMC Javelin	5.8508670	6.4507155	6.2193563	4.1915369
## Camaro Z28	6.9503062	7.1027162	7.2970631	5.5640650
## Pontiac Firebird	6.1149641	6.7946533	6.4857014	4.6855121
## Fiat X1-9	0.9440123	1.8356814	1.2631917	2.8771692
## Porsche 914-2	3.2195133	2.9932093	3.4109616	4.6534714
## Lotus Europa	2.3907048	2.8622875	2.5923660	4.2445533

## Ford Pantera L	6.6166138	6.4835328	6.9120796	6.7110287
## Ferrari Dino	5.4662206	5.4859770	5.7728984	5.8353128
## Maserati Bora	8.1807755	8.1867266	8.4801671	8.0606070
## Volvo 142E	2.1944946	2.6127269	2.6006079	2.7534994
##	Dodge Challenger AMC Javelin Camaro Z28 Pontiac			
Firebird				
## Mazda RX4 Wag				
## Datsun 710				
## Hornet 4 Drive				
## Hornet Sportabout				
## Valiant				
## Duster 360				
## Merc 240D				
## Merc 230				
## Merc 280				
## Merc 280C				
## Merc 450SE				
## Merc 450SL				
## Merc 450SLC				
## Cadillac Fleetwood				
## Lincoln Continental				
## Chrysler Imperial				
## Fiat 128				
## Honda Civic				
## Toyota Corolla				
## Toyota Corona				
## Dodge Challenger				
## AMC Javelin	0.7827694			
## Camaro Z28	2.7782606	2.4813171		
## Pontiac Firebird	1.1942533	1.1771569	2.4529311	
## Fiat X1-9	5.8525045	5.4949181	6.5789332	
5.8831504				
## Porsche 914-2	5.8698053	5.4885985	6.0073280	
5.8049597				
## Lotus Europa	6.1425433	5.9214212	6.6478423	
6.1638574				
## Ford Pantera L	5.0078711	4.7275036	3.6306118	
4.7026976				
## Ferrari Dino	5.0080706	4.8308278	4.5053590	
5.0961648				
## Maserati Bora	6.0178217	5.9396759	4.4569171	
5.8678663				
## Volvo 142E	5.1510061	4.7542900	5.5099681	
5.1557398				
##	Fiat X1-9 Porsche 914-2 Lotus Europa Ford Pantera L			
## Mazda RX4 Wag				
## Datsun 710				
## Hornet 4 Drive				
## Hornet Sportabout				
## Valiant				

```

## Duster 360
## Merc 240D
## Merc 230
## Merc 280
## Merc 280C
## Merc 450SE
## Merc 450SL
## Merc 450SLC
## Cadillac Fleetwood
## Lincoln Continental
## Chrysler Imperial
## Fiat 128
## Honda Civic
## Toyota Corolla
## Toyota Corona
## Dodge Challenger
## AMC Javelin
## Camaro Z28
## Pontiac Firebird
## Fiat X1-9
## Porsche 914-2      2.9043943
## Lotus Europa      2.1786546      2.5613776
## Ford Pantera L    6.2258629      4.6929233      5.5540990
## Ferrari Dino      5.1088276      3.6305579      4.2187875      3.0648409
## Maserati Bora     7.8702812      6.5745699      6.9847187      3.0287549
## Volvo 142E       1.6207836      2.8881825      2.6646220      5.2160045
##
## Ferrari Dino Maserati Bora
## Mazda RX4 Wag
## Datsun 710
## Hornet 4 Drive
## Hornet Sportabout
## Valiant
## Duster 360
## Merc 240D
## Merc 230
## Merc 280
## Merc 280C
## Merc 450SE
## Merc 450SL
## Merc 450SLC
## Cadillac Fleetwood
## Lincoln Continental
## Chrysler Imperial
## Fiat 128
## Honda Civic
## Toyota Corolla
## Toyota Corona
## Dodge Challenger
## AMC Javelin
## Camaro Z28

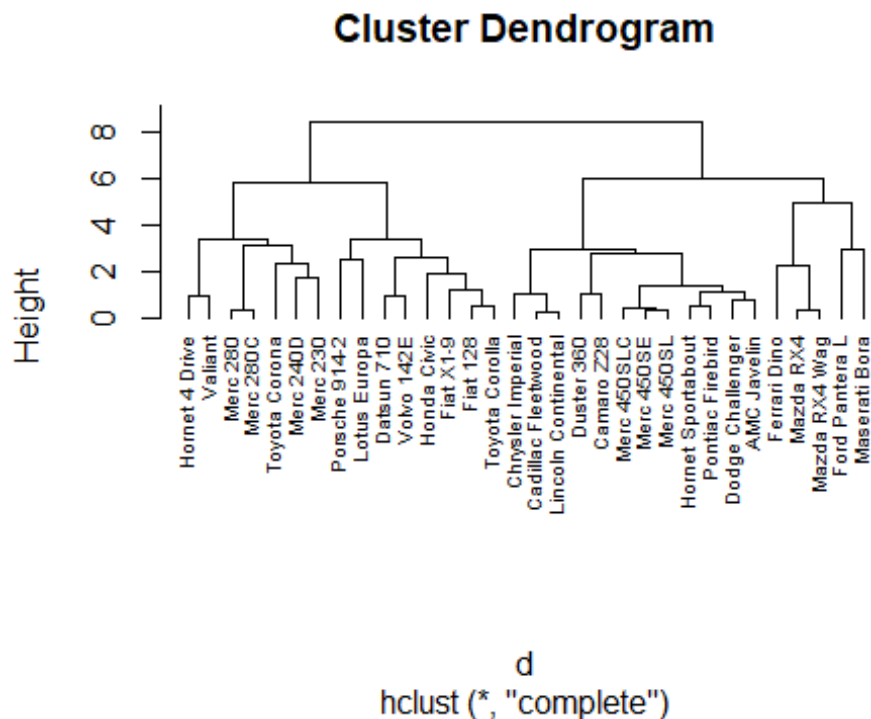
```

```
## Pontiac Firebird
## Fiat X1-9
## Porsche 914-2
## Lotus Europa
## Ford Pantera L
## Ferrari Dino
## Maserati Bora          3.3719604
## Volvo 142E            4.2489749      6.7446456

hc1 <- hclust(d, method = "complete") # Hierarchical clustering using
Complete Linkage
hc1

##
## Call:
## hclust(d = d, method = "complete")
##
## Cluster method      : complete
## Distance            : euclidean
## Number of objects: 32

plot(hc1, cex = 0.6, hang = -1) # Plot the obtained dendrogram
```



```
# methods to assess
m <- c( "average", "single", "complete", "ward")
names(m) <- c( "average", "single", "complete", "ward")
```

```
# function to compute coefficient
```

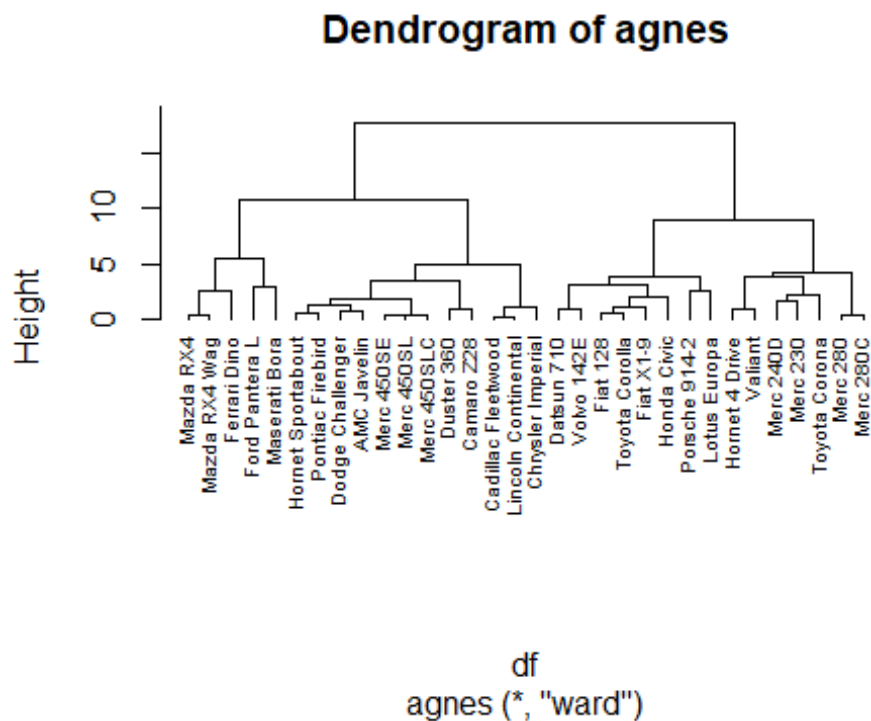
```
ac <- function(x) {  
  agnes(df, method = x)$ac  
}
```

```
map_dbl(m, ac)
```

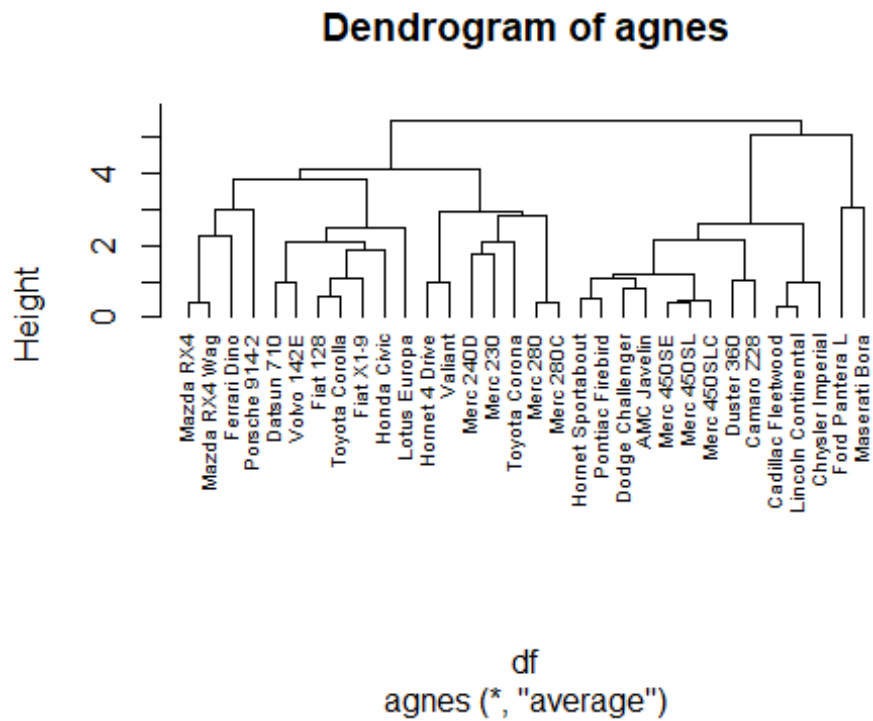
```
## average single complete ward  
## 0.7898219 0.6400643 0.8636151 0.9338742
```

```
#Wards method identifies the strongest clustering structure of the four  
methods
```

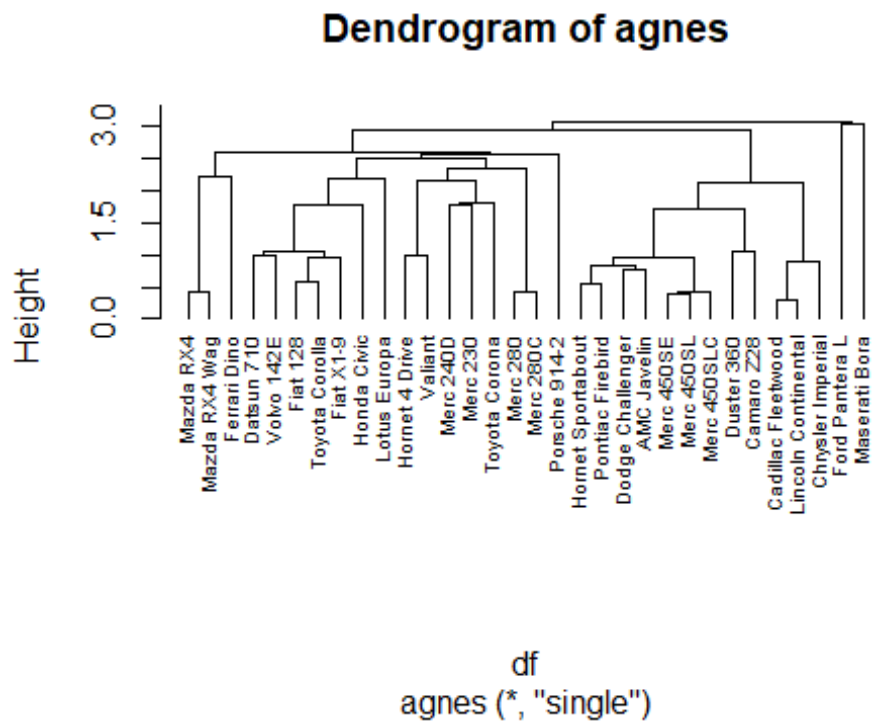
```
hc3 <- agnes(df, method = "ward")  
pltree(hc3, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```



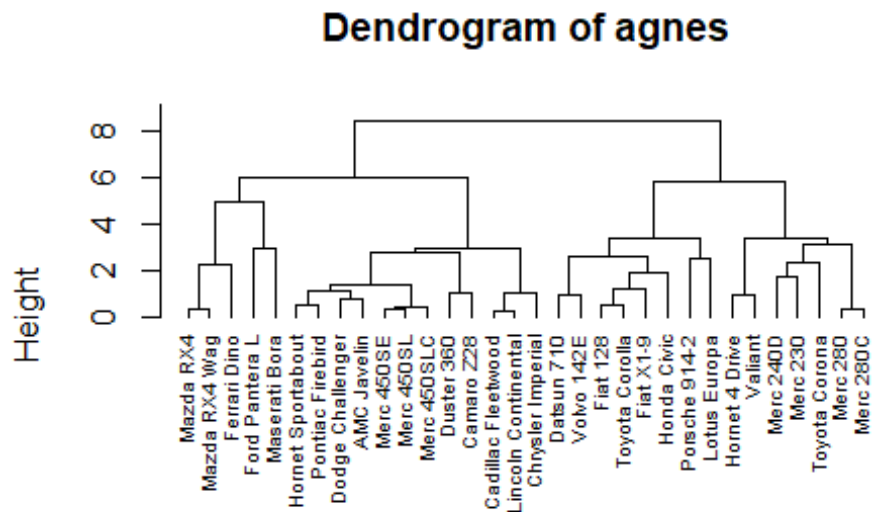
```
hc3 <- agnes(df, method = "average")  
pltree(hc3, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```



```
hc3 <- agnes(df, method = "single")
pltree(hc3, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```

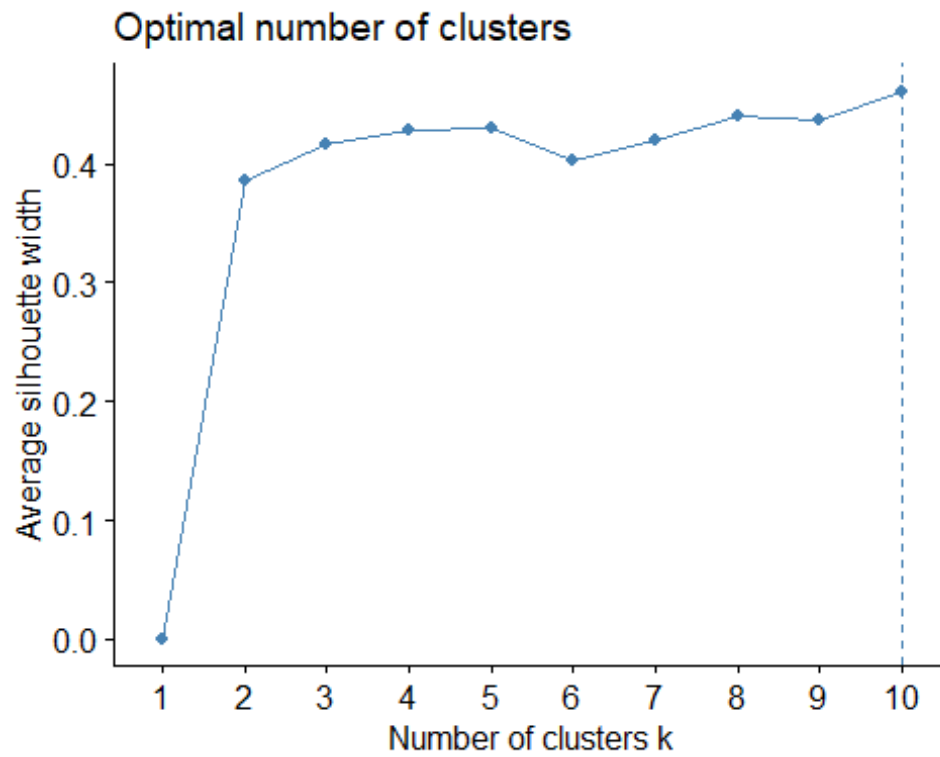


```
hc3 <- agnes(df, method = "complete")
pltree(hc3, cex = 0.6, hang = -1, main = "Dendrogram of agnes")
```



df
agnes (*, "complete")

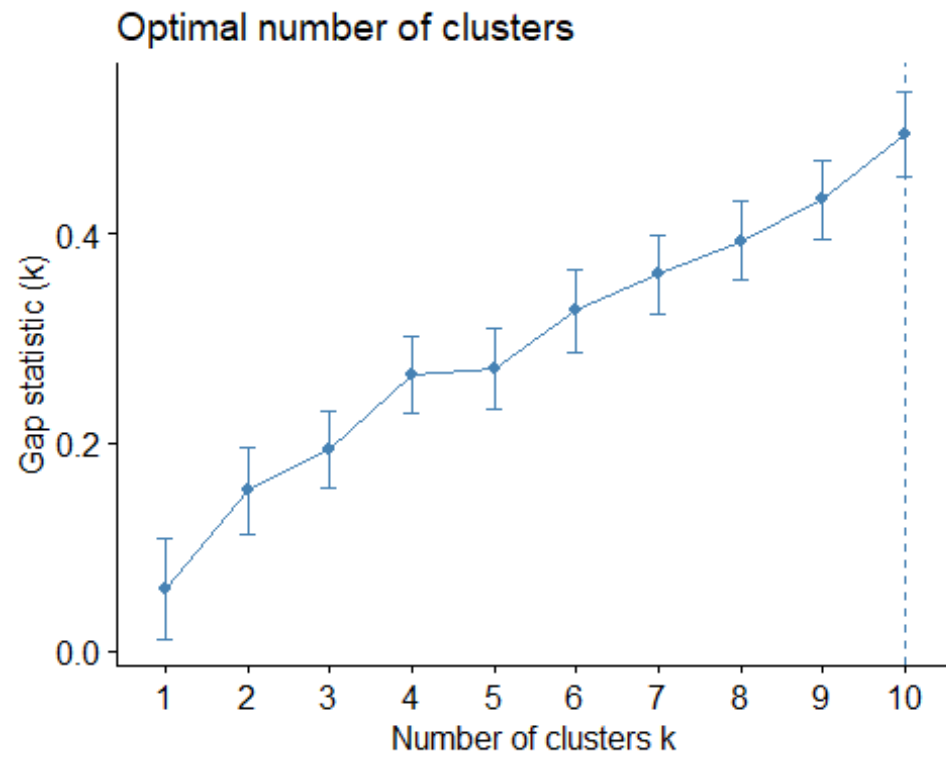
```
#Average Silhouette Method
fviz_nbclust(df, FUN = hcut, method = "silhouette")
```



#the optimal number of clusters required is 10, by using silhouette methos

#Gap Statistic Method

```
gap_stat <- clusGap(df, FUN = hcut, nstart = 25, K.max = 10, B = 50)  
fviz_gap_stat(gap_stat)
```

#the optimal number of clusters required is 10, by using Gap Statistics method