

Experiment 3.1

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Semester: 6

Subject Name: Data Mining Lab

Subject Code: 20CSP- 376

Aim: To perform hierarchical clustering using R programming.

Objective: To Learn about hierarchical Clustering.

Theory: Hierarchical cluster analysis (also known as hierarchical clustering) is a clustering technique where clusters have a hierarchy or a predetermined order. Hierarchical clustering can be represented by a tree- like structure called a Dendrogram.

There are two types of hierarchical clustering:

- Agglomerative hierarchical clustering: This is a bottom-up approach where each data point starts in its own cluster and as one moves up the hierarchy, similar pairs of clusters are merged.
- Divisive hierarchical clustering: This is a top-down approach where all data points start in one cluster and as one moves down the hierarchy, clusters are split recursively.

Script/Code/Steps:

```
install.packages("dplyr")
library(dplyr)
distance_mat <- dist(mtcars, method = 'euclidean')
distance_mat
set.seed(240)
Hierar_cl <- hclust(distance_mat, method = "average")
Hierar_cl
plot(Hierar_cl)
abline(h = 10, col = "blue")
fit <- cutree(Hierar_cl, k = 3 )
fit
table(fit)
rect.hclust(Hierar_cl, k = 3, border =
"blue")
```

```

1 install.packages("dplyr")
2 library(dplyr)
3 distance_mat <- dist(mtcars, method = 'euclidean')
4 distance_mat
5 set.seed(240)
6 Hierar_cl <- hclust(distance_mat, method = "average")
7 Hierar_cl
8 plot(Hierar_cl)
9 abline(h = 10, col = "green")
10 fit <- cutree(Hierar_cl, k = 3 )
11 fit
12 table(fit)
13 rect.hclust(Hierar_cl, k = 3, border = "green")
14 |
15

```

14:1 (Top Level) R Script

Output Screenshots:

Console

Terminal

Background Jobs

R 4.2.2 · ~ /

> library(dplyr)

> distance_mat <- dist(mtcars, method = 'euclidean')

> distance_mat

Mazda RX4 Wag	Mazda RX4	Mazda RX4 Wag	Datsun 710	Hornet 4 Drive	Hornet Sportabout	
	0.6153251					
Datsun 710	54.9086059	54.8915169				
Hornet 4 Drive	98.1125212	98.0958939	150.9935191			
Hornet Sportabout	210.3374396	210.3358546	265.0831615	121.0297564		
Valiant	65.4717710	65.4392224	117.7547018	33.5508692	152.1241352	
Duster 360	241.4076490	241.4088680	294.4790230	169.4299647	70.1767262	
Merc 240D	50.1532711	50.1146059	49.6584796	121.2739722	241.5069657	
Merc 230	25.4683117	25.3284509	33.1803843	118.2433145	233.4924012	
Merc 280	15.3641921	15.2956865	66.9363534	91.4224033	199.3344960	
Merc 280C	15.6724727	15.5837744	67.0261397	91.4612914	199.3406564	
Merc 450SE	135.4307018	135.4254826	189.1954941	72.4964325	84.3888482	
Merc 450SL	135.4014424	135.3960351	189.1631745	72.4313532	84.3683999	
Merc 450SLC	135.4794674	135.4723157	189.2345426	72.5718466	84.4332423	
Cadillac Fleetwood	326.3395903	326.3355070	381.0926242	234.4403876	116.2804201	
Lincoln Continental	318.0469808	318.0429333	372.8012090	227.9726091	108.0624299	
Chrysler Imperial	304.7203408	304.7169175	359.3014906	218.1548299	97.2049146	
Fiat 128	93.2679950	93.2530993	40.9933763	184.9689734	302.0377212	
Honda Civic	102.8307567	102.8238713	52.7704607	191.5518700	310.0324645	
Toyota Corolla	100.6040368	100.5887588	47.6535017	192.6714187	309.5581776	
Toyota Corona	42.3075233	42.2659224	12.9654743	138.5304725	252.3331988	
Dodge Challenger	163.1150750	163.1134210	217.7795805	72.4403915	48.9838851	
AMC Javelin	149.6047203	149.6014522	204.3188913	61.3601899	61.4274240	
Camaro Z28	233.2228758	233.2248748	286.0049209	163.6632641	70.9665308	
Pontiac Firebird	248.6780270	248.6762035	303.3583889	156.2240346	40.0052475	
Fiat X1-9	92.5048389	92.4940020	39.8815148	184.4471198	301.5669483	
Porsche 914-2	44.4033659	44.4073589	13.1357109	139.1579524	254.1452553	
Lotus Europa	65.7328377	65.7362635	25.0948550	163.2367437	272.3582423	
Ford Pantera L	245.4247064	245.4293785	297.2940489	180.1140339	89.5934049	
Ferrari Dino	66.7661029	66.7764167	90.2415509	130.5523007	215.0673853	
Maserati Bora	265.6454248	265.6491465	309.7718171	229.3419352	170.7094473	
Volvo 142E	39.1894029	39.1626037	20.6939436	137.0363299	248.0063378	
	Valiant	Duster 360	Merc 240D	Merc 230	Merc 280	Merc 280C
Mazda RX4 Wag						
Datsun 710						
Hornet 4 Drive						
Hornet Sportabout						
Valiant						

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Camaro Z28	100.3046106	105.6062618							
Pontiac Firebird	85.8075196	99.2836114	86.2665759						
Fiat X1-9	253.6624046	240.5266823	325.1490914	339.1396182					
Porsche 914-2	206.6452569	193.3080584	276.8924414	292.1646488	48.3775209				
Lotus Europa	226.5004836	212.7568765	287.6179004	311.3862342	49.8406880				
Ford Pantera L	118.7516779	123.3832044	19.3589023	101.7389686	336.7018783				
Ferrari Dino	174.9280395	161.1060307	216.7489910	255.0570519	127.8210813				
Maserati Bora	185.9059273	185.1553411	102.5946154	188.3240020	349.1199576				
Volvo 142E	201.3682522	187.6978440	266.5277736	286.7497823	60.4120429				
Porsche 914-2 Lotus Europa Ford Pantera L 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	33.7678653								
Ford Pantera L	288.5852993	297.5376920							
Ferrari Dino	87.9105966	80.4553451	224.4587490						
Maserati Bora	303.9222549	303.2796468	86.9383253	223.5342175					
Volvo 142E	18.7555858	27.8104457	277.4803312	70.4751034	289.1157363				

```
> set.seed(240)
> Hierar_cl <- hclust(distance_mat, method = "average")
> Hierar_cl

Call:
hclust(d = distance_mat, method = "average")

Cluster method   : average
Distance         : euclidean
Number of objects: 32

> plot(Hierar_cl)
> abline(h = 10, col = "green")
> fit <- cutree(Hierar_cl, k = 3)
> fit
```

Mazda RX4	Mazda RX4 Wag	Datsun 710	Hornet 4 Drive
1	1	1	2
Hornet Sportabout	Valiant	Duster 360	Merc 240D
2	2	2	1
Merc 230	Merc 280	Merc 280C	Merc 450SE
1	1	1	2
Merc 450SL	Merc 450SLC	Cadillac Fleetwood	Lincoln Continental
2	2	2	2
Chrysler Imperial	Fiat 128	Honda Civic	Toyota Corolla
2	1	1	1
Toyota Corona	Dodge Challenger	AMC Javelin	Camaro Z28
1	2	2	2
Pontiac Firebird	Fiat X1-9	Porsche 914-2	Lotus Europa
2	1	1	1
Ford Pantera L	Ferrari Dino	Maserati Bora	Volvo 142E
2	1	3	1

```
> table(fit)
fit
1 2 3
16 15 1
> rect.hclust(Hierar_cl, k = 3, border = "green")
```

Environment	History	Connections	Tutorial
<div> <div>Import Dataset</div> <div>377 MiB</div> </div> <div> <div>R</div> <div>Global Environment</div> </div>			
Data			
Groceries	Formal class transactions		
Hierar_cl	List of 7		
Mushroom	Large transactions (8124 elements, 1.3 MB)		
rules	Formal class rules		
Values			
distance_mat	'dist' num [1:496] 0.615 54.909 98.113 210.337 65.472 ...		
fit	Named int [1:32] 1 1 1 2 2 2 2 1 1 1 ...		

Output of Cluster Dendrogram:

