

# Networks: Project 2

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## 1 Problem Statement

Select an undirected labeled network with at least 50 nodes and average degree at least 4. Determine the graphlet spectra for its nodes. On their basis construct a dissimilarity between nodes and use it to cluster the nodes.

## 2 Network

The data is the network of American football games between Division IA colleges during regular season Fall 2000, as compiled by M. Girvan and M. Newman (Girvan and Newman (2002)). To quote

The network we look at is a representation of the schedule of Division I games for the 2000 season: vertices in the graph represent teams (identified by their college names) and edges represent regular-season games between the two teams they connect. What makes this network interesting is that it incorporates a known community structure. The teams are divided into conferences containing around 8–12 teams each. Games are more frequent between members of the same conference than between members of different conferences, with teams playing an average of about seven intraconference games and four interconference games in the 2000 season.

Let's check if the clustering based on the graphlets add more information.

### 2.1 Basic Network Characteristics

Property	Value
Vertices	115
Edges	613
Directed	No

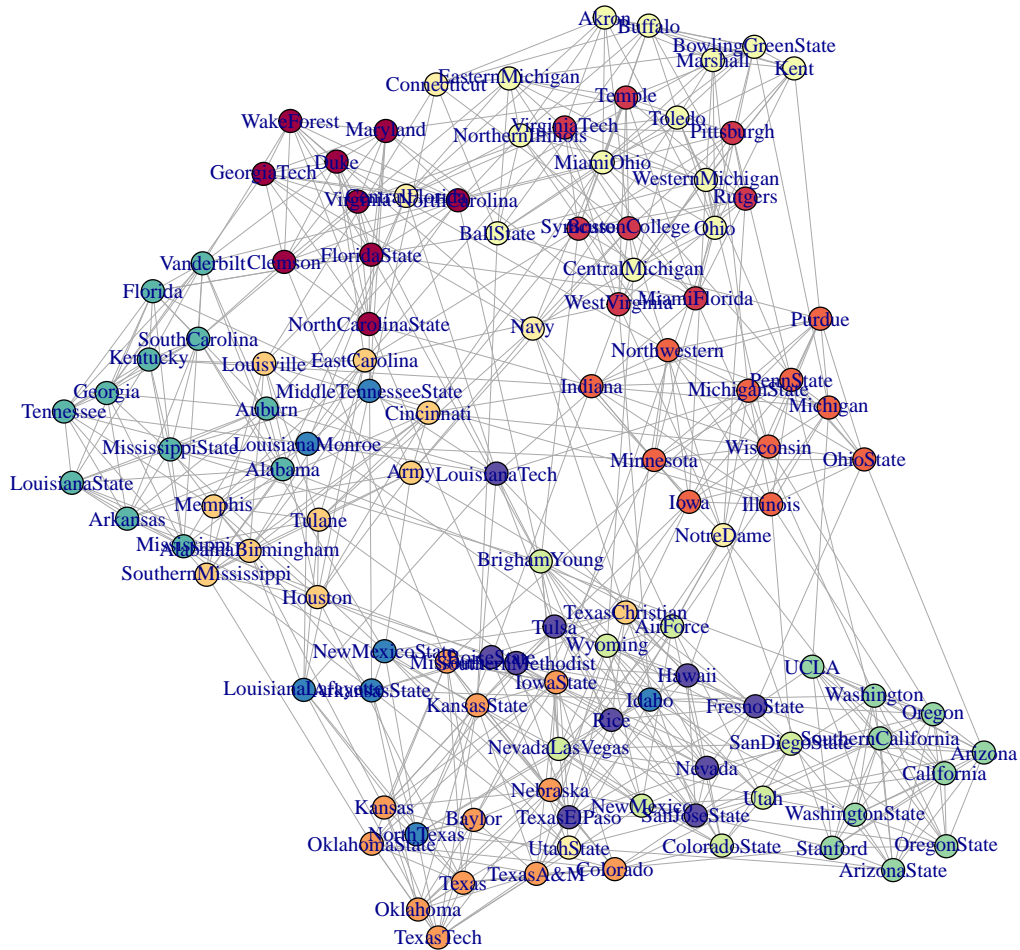
Property	Value
Weighted	No
Average degree	10.7
Diameter	4
Acyclic	No
Edge density	0.0935163996948894
Average Path Length	2.50816170861937
Transitivity (global)	0.407239819004525

The network coloured according to divisions each club belongs to.

```
pal = colorRampPalette(brewer.pal(11,"Spectral"))(length(unique(V(net)$value)))
node_colors <- pal[V(net)$value + 1]
V(net)$color <- node_colors

coords_fr = layout.fruchterman.reingold(net, weights=E(net)$weight)

# igraph plot options
igraph.options(vertex.size=8, edge.width=0.75)
plot(net, layout=coords_fr, vertex.size = 5, vertex.color=V(net)$color)
```



### 3 Analysis

The orca package is used to find associations of nodes with the 4-node orbits.

#### 3.1 Getting Orbits

```
# convert to the 'edgelist' format required by orca
net_edgelist <- as.data.frame(as_edgelist(net))
net_edgelist <- transform(net_edgelist, V1 = as.integer(net_edgelist$V1), V2 = as.integer(net_edgelist$V2))
```

Printing the graphlet spectra.

```
# convert to the 'edgelist' format required by orca
net_orbits      <- count4(net_edgelist)
net_orbits_scaled <- scale(net_orbits)      # scale the data
```

```
print('Found orbits for the nodes are')
```

```
## [1] "Found orbits for the nodes are"
```

```
print(net_orbits)
```

```
##      00 01 02 03 04 05 06 07 08 09 010 011 012 013 014
## [1,] 12 71 43 23 433 575 116 66 11 105 158 113 13 6 35
## [2,] 12 58 37 29 363 461 98 49 13 81 147 108 2 7 56
## [3,] 12 72 41 25 446 474 145 47 26 52 178 117 44 35 21
## [4,] 12 73 44 22 424 527 145 73 32 68 152 91 30 39 17
## [5,] 11 62 32 23 394 389 111 34 21 70 167 90 11 6 35
## [6,] 12 64 42 24 371 417 100 59 34 58 159 99 30 45 17
## [7,] 12 77 45 21 472 588 162 80 28 79 141 90 31 30 20
## [8,] 12 68 40 26 391 496 131 61 28 81 136 90 30 37 32
## [9,] 11 56 29 26 306 275 87 26 33 46 139 76 40 31 32
## [10,] 11 57 29 26 350 327 76 24 17 75 175 83 17 23 35
## [11,] 10 58 25 20 343 265 109 23 29 47 131 49 34 33 15
## [12,] 10 64 31 14 412 393 129 38 20 75 119 62 12 10 10
## [13,] 10 63 30 15 367 360 120 37 27 71 119 56 17 17 10
## [14,] 11 69 34 21 412 445 125 46 18 82 143 69 33 30 20
## [15,] 10 61 28 17 358 308 118 24 32 61 123 68 27 16 12
## [16,] 12 74 43 23 422 547 123 70 24 99 157 90 31 40 20
## [17,] 11 65 32 23 418 452 117 40 9 92 170 79 5 10 36
## [18,] 11 67 36 19 403 434 135 52 27 60 124 70 29 28 15
## [19,] 11 55 33 22 335 296 84 34 32 43 156 85 17 25 21
## [20,] 11 56 31 24 341 347 72 31 16 73 161 88 10 10 36
## [21,] 11 64 35 20 366 392 110 47 27 63 134 71 30 32 15
## [22,] 11 60 30 25 350 368 96 34 17 76 140 69 29 30 32
## [23,] 11 54 27 28 308 281 81 23 20 54 143 64 38 46 32
## [24,] 11 58 30 25 336 309 82 24 30 68 173 92 19 14 35
## [25,] 10 73 35 10 437 533 144 60 16 124 86 50 3 0 10
## [26,] 11 52 27 28 371 360 94 25 3 63 148 84 3 0 56
## [27,] 10 63 30 15 375 366 124 38 26 74 121 54 14 18 10
## [28,] 11 65 35 20 358 397 120 47 32 66 131 71 29 32 15
## [29,] 9 59 24 12 363 332 100 28 13 92 108 38 3 8 10
## [30,] 11 50 28 27 394 286 88 23 10 26 176 79 9 25 38
## [31,] 11 54 29 26 360 345 76 31 7 66 167 71 7 26 37
## [32,] 11 63 34 21 332 365 98 37 35 77 153 87 21 21 20
## [33,] 11 70 34 21 411 440 133 46 26 84 144 69 32 30 20
## [34,] 10 42 16 29 291 197 63 7 7 47 163 50 2 7 56
## [35,] 11 58 34 21 332 329 85 37 32 58 141 87 23 21 20
## [36,] 11 61 31 24 365 405 108 37 17 84 155 76 8 16 36
## [37,] 8 67 24 4 365 381 141 34 13 113 46 20 8 2 0
## [38,] 11 46 26 29 362 291 79 23 6 34 153 79 2 7 56
## [39,] 11 67 38 17 372 444 111 55 30 84 122 79 22 19 12
## [40,] 11 65 32 23 429 370 132 33 23 45 166 77 34 35 20
## [41,] 11 70 36 19 376 442 128 52 37 93 126 71 29 26 16
## [42,] 10 54 21 24 346 273 76 16 10 78 183 52 5 16 36
## [43,] 7 60 18 3 322 276 139 22 18 91 22 11 13 2 0
```

```

## [44,] 11 66 39 16 383 436 115 62 35 77 124 72 15 21 10
## [45,] 11 67 35 20 390 424 121 37 19 70 138 93 35 18 17
## [46,] 11 50 26 29 349 329 85 23 7 61 153 79 2 7 56
## [47,] 11 55 26 29 377 319 86 16 12 64 185 93 11 0 56
## [48,] 11 71 34 21 408 444 135 46 26 86 150 69 32 30 20
## [49,] 11 68 35 20 394 452 118 47 20 90 146 76 21 22 20
## [50,] 11 53 25 30 371 314 82 21 9 63 188 74 5 14 56
## [51,] 9 56 22 14 368 290 93 14 5 73 114 53 17 6 11
## [52,] 11 54 29 26 330 312 80 31 18 55 132 67 30 34 33
## [53,] 10 63 26 19 365 318 127 26 25 62 137 52 31 26 16
## [54,] 12 61 36 30 408 455 95 40 7 78 182 116 10 8 56
## [55,] 10 53 23 22 303 239 74 16 24 64 166 55 12 26 23
## [56,] 11 62 32 23 371 411 107 35 16 82 143 88 17 7 35
## [57,] 10 59 26 19 350 296 106 26 24 57 127 51 30 28 15
## [58,] 10 60 28 17 390 343 111 34 16 61 127 52 19 18 16
## [59,] 10 81 40 5 491 646 183 84 12 135 43 33 8 2 1
## [60,] 8 58 22 6 377 303 116 25 12 76 67 27 6 3 1
## [61,] 11 66 34 21 413 400 122 46 23 57 150 69 32 30 20
## [62,] 11 70 37 18 399 477 143 55 30 81 125 78 19 12 20
## [63,] 11 67 35 20 358 418 115 43 25 78 128 79 37 28 15
## [64,] 9 68 31 5 426 454 142 53 14 100 52 28 5 2 1
## [65,] 11 68 34 21 423 428 129 46 21 69 150 69 30 30 20
## [66,] 11 66 35 20 365 397 133 47 37 62 133 71 28 32 15
## [67,] 11 66 36 19 398 419 126 50 28 61 143 75 21 24 16
## [68,] 12 58 34 32 403 395 89 23 6 61 193 131 19 9 57
## [69,] 11 58 29 26 341 352 98 31 15 70 138 67 30 34 33
## [70,] 11 76 42 13 456 602 139 72 13 122 104 79 14 4 10
## [71,] 11 66 35 20 357 419 120 48 26 72 127 69 31 33 15
## [72,] 10 57 25 20 342 299 102 24 22 64 144 55 13 18 23
## [73,] 11 66 34 21 380 402 114 44 30 83 146 70 26 34 17
## [74,] 11 56 26 29 384 330 96 16 11 63 184 93 12 0 56
## [75,] 11 68 34 21 380 402 130 45 37 78 148 68 28 35 17
## [76,] 10 59 26 19 372 312 105 24 15 57 143 56 23 24 16
## [77,] 11 56 32 23 321 281 80 27 30 43 151 84 34 39 15
## [78,] 11 58 29 26 358 350 98 32 13 61 144 64 30 37 32
## [79,] 11 48 26 29 292 228 58 19 17 38 152 64 33 49 33
## [80,] 11 56 29 26 350 341 91 26 13 62 167 80 13 23 36
## [81,] 11 85 45 10 428 698 127 90 6 167 82 62 29 11 2
## [82,] 11 70 35 20 395 434 140 48 32 77 136 69 33 33 15
## [83,] 11 93 47 8 488 773 189 101 18 167 59 58 31 4 2
## [84,] 11 51 24 31 373 280 82 13 8 46 198 81 8 15 56
## [85,] 11 66 34 21 352 372 111 40 37 73 144 78 35 30 17
## [86,] 9 61 23 13 381 301 137 23 24 63 107 41 16 10 10
## [87,] 11 66 35 20 429 445 132 51 18 66 147 68 16 26 20
## [88,] 11 68 36 19 397 446 137 52 28 72 122 70 28 28 15
## [89,] 12 60 36 30 421 461 101 46 5 70 183 104 3 14 56
## [90,] 11 52 27 28 340 352 84 25 7 73 148 84 3 0 56
## [91,] 9 62 24 12 370 323 129 24 24 85 103 46 11 4 10
## [92,] 11 67 35 20 420 438 124 43 15 68 158 80 22 26 16
## [93,] 11 72 37 18 451 507 152 53 17 81 129 78 25 18 16
## [94,] 10 55 22 23 445 281 114 7 3 33 186 77 11 1 35
## [95,] 10 45 18 27 348 199 75 8 5 28 182 46 7 28 38
## [96,] 10 58 25 20 324 281 100 23 23 59 137 49 29 33 15
## [97,] 10 59 26 19 324 270 115 26 37 48 125 51 31 28 15

```

```
## [98,] 8 61 22 6 388 317 140 25 14 74 64 27 9 3 1
## [99,] 11 64 35 20 384 400 113 48 26 66 126 69 31 33 15
## [100,] 10 57 26 19 332 302 89 26 23 73 135 57 15 16 21
## [101,] 11 67 33 22 432 401 135 42 24 56 159 68 31 35 20
## [102,] 10 57 24 21 375 335 109 22 8 66 138 63 12 0 35
## [103,] 10 59 26 19 353 298 98 26 24 67 127 52 29 26 16
## [104,] 10 42 17 28 315 209 64 8 4 40 155 56 3 0 56
## [105,] 12 71 42 24 434 569 120 64 12 105 166 108 11 12 36
## [106,] 10 39 15 30 309 174 62 6 4 29 167 44 1 14 56
## [107,] 11 68 33 22 420 430 124 42 14 75 160 68 31 35 20
## [108,] 10 61 26 19 354 313 112 26 25 70 132 53 27 24 17
## [109,] 10 49 19 26 297 181 80 7 20 33 140 50 40 31 32
## [110,] 11 48 26 29 333 288 78 16 10 43 154 93 9 0 56
## [111,] 11 56 27 28 426 355 111 25 8 48 179 84 5 0 56
## [112,] 11 58 29 26 304 322 83 31 27 77 144 66 30 36 32
## [113,] 10 65 32 13 395 372 130 44 32 58 95 56 27 12 8
## [114,] 10 53 25 20 325 233 89 22 25 31 132 51 31 32 15
## [115,] 11 54 25 30 401 333 102 21 4 53 189 74 5 14 56
```

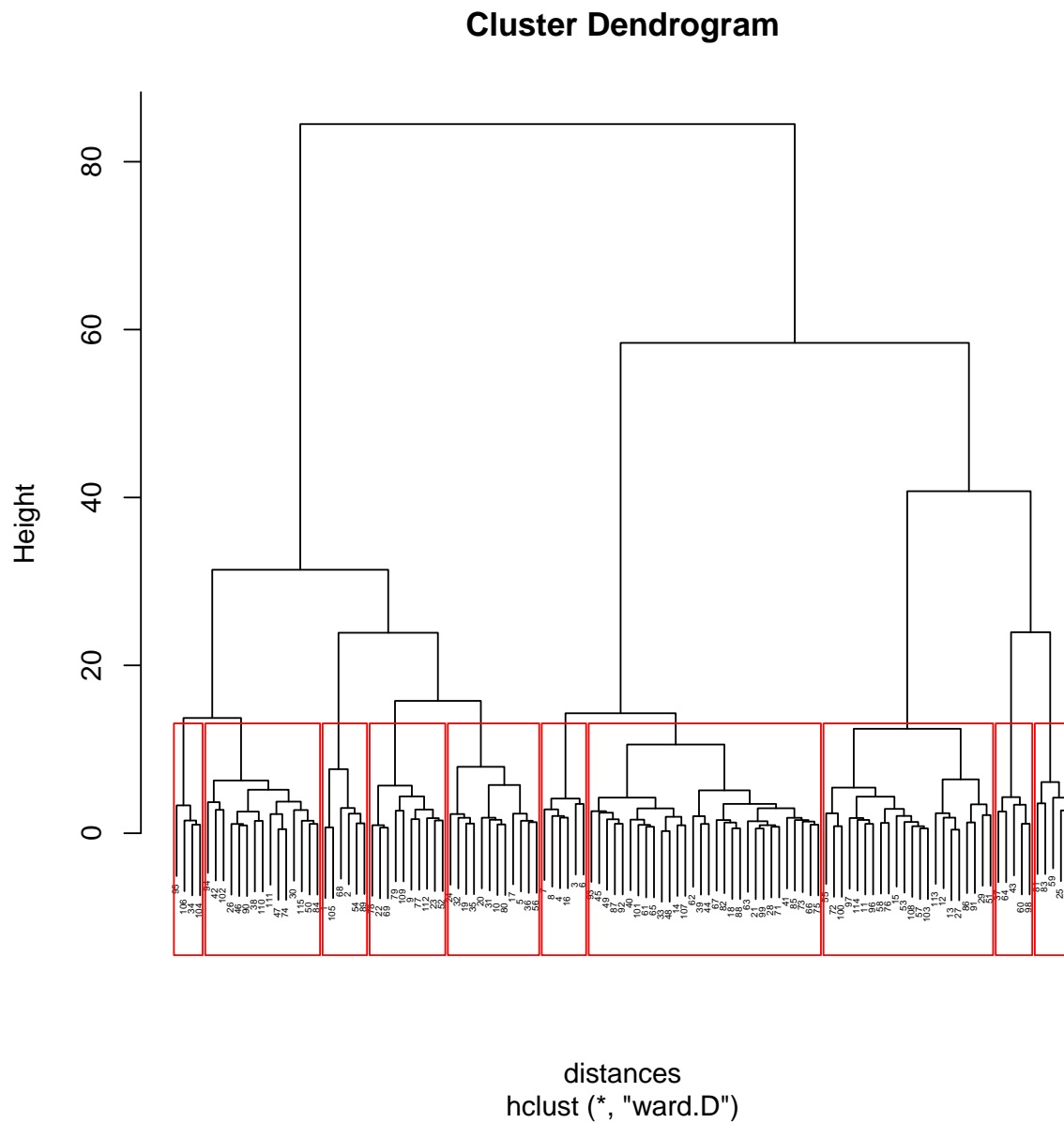
```
#
```

## 3.2 H-Clustering

I use the euclidian distance for hierarchical clustering and made the cut to form ten (10) partitions.

```
# heatmap(net_orbits_scaled, cexRow = 0.3)

# Ward Hierarchical Clustering
distances <- dist(net_orbits_scaled, method = "euclidean") # distance matrix
hclust_fit <- hclust(distances, method="ward.D")
plot(hclust_fit, cex = 0.3) # display dendrogram
k = 10
# draw dendrogram with red borders around the k clusters
rect.hclust(hclust_fit, k = k, border="red")
```



```
clusters2 <- cutree(hclust_fit, k = k) # cut tree into k clusters
```

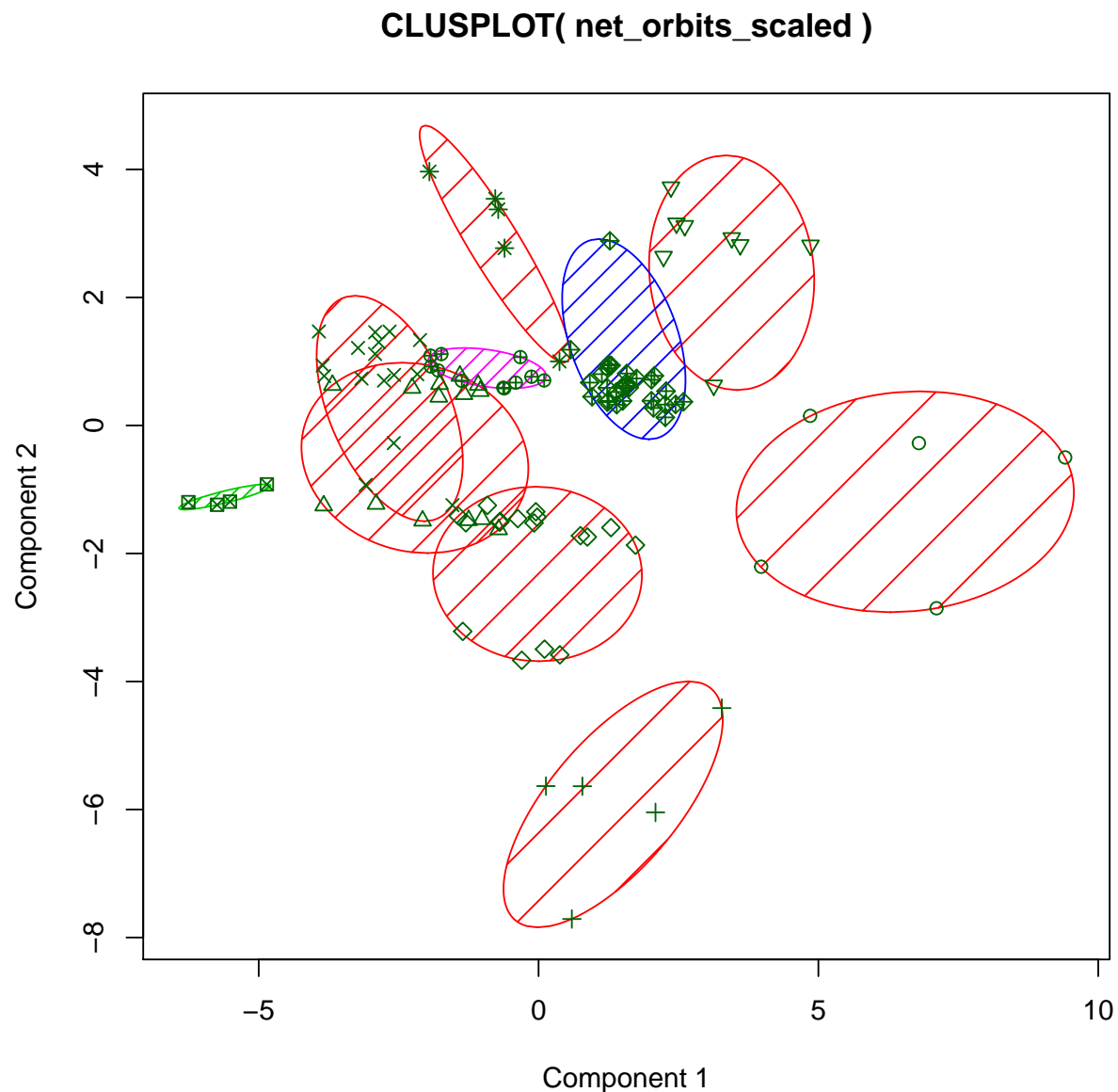
### 3.3 k-Means Clustering

The same number of centers (10) was used for k-Means clustering to check later if results of two different clustering methods will match.

```
k = 10

kmeans_fit <- kmeans(net_orbits_scaled, k)

# Cluster Plot against 1st 2 principal components
library(cluster)
clusplot(net_orbits_scaled, kmeans_fit$cluster, color = TRUE, shade = TRUE, labels = FALSE, lines=0)
```



These two components explain 70.78 % of the point variability.

```
# Centroid Plot against 1st 2 discriminant functions
#library(fpc)
#plotcluster(net_orbits_scaled, kmeans_fit$cluster)
```

### 3.4 Visualization (Hierarchical Clustering)

The network coloured based on the clusters defined during hierarchical clustering.

```
# color the nodes
pal = colorRampPalette(brewer.pal(11,"Spectral"))(length(unique(clusters2)))
node_colors <- pal[clusters2]
```



```
V(net)$color <- node_colors
```

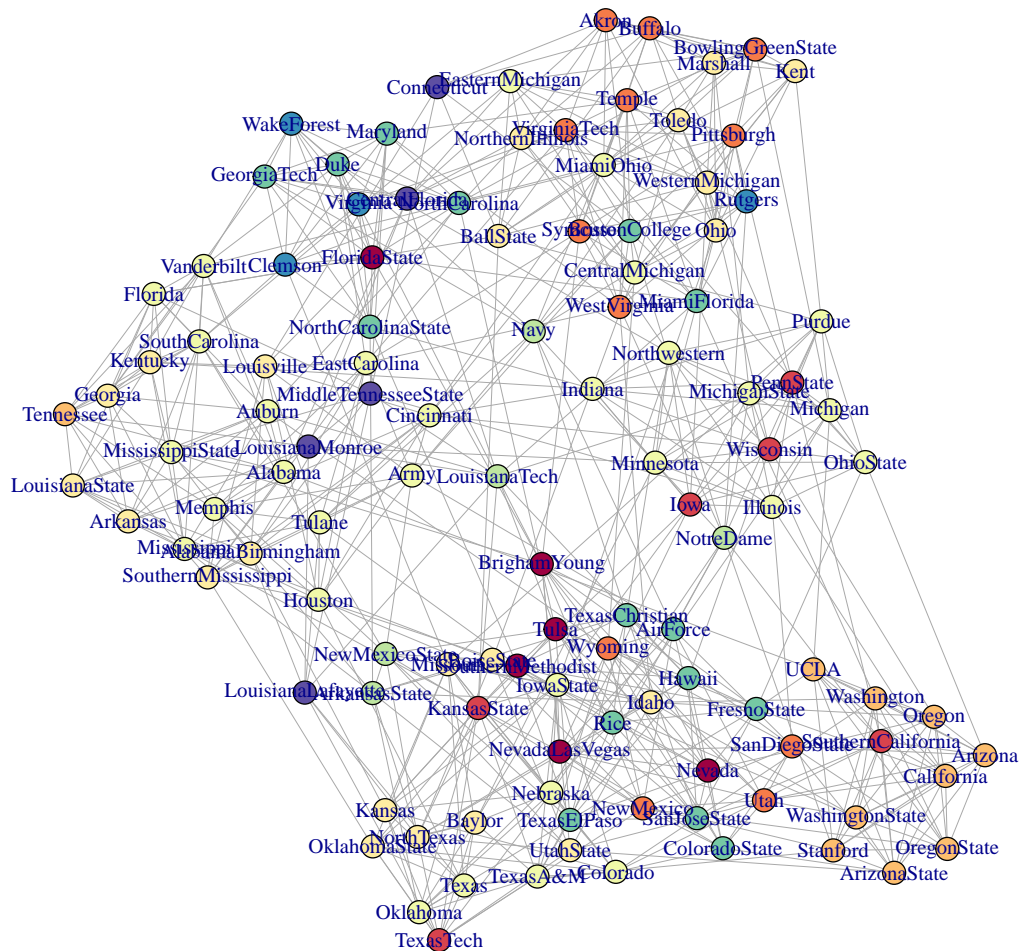
```
#coords_fr = layout.fruchterman.reingold(net, weights=E(net)$weight)
```

```
# igraph plot options
```

```
igraph.options(vertex.size=5, edge.width=0.75)
```

```
# plot network
```

```
plot(net, layout=coords_fr, vertex.color=V(net)$color)
```



### 3.5 Visualization (k-Means)

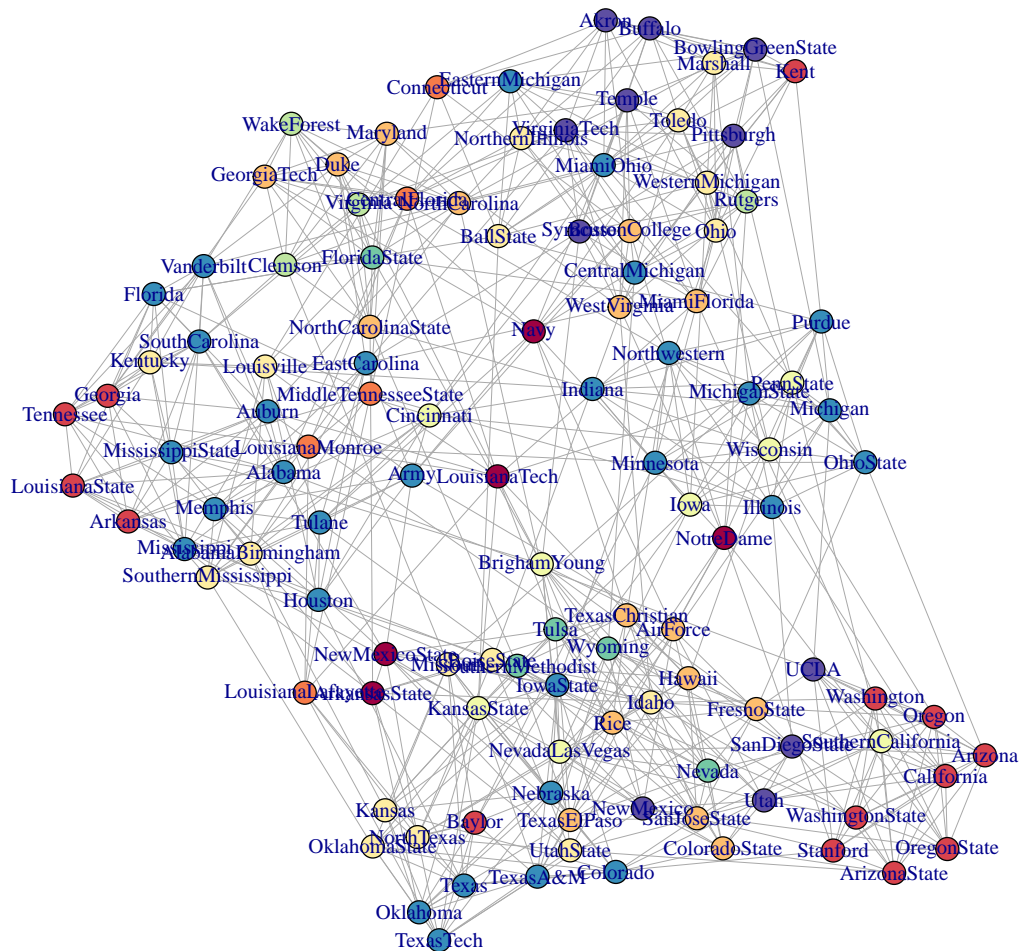
The network coloured based on the clusters defined during k-Means clustering.

```

# color the nodes
pal = colorRampPalette(brewer.pal(11,"Spectral"))(length(unique(kmeans_fit$cluster)))
node_colors <- pal[kmeans_fit$cluster]
V(net)$color <- node_colors

# coords_fr = layout.fruchterman.reingold(net, weights=E(net)$weight)
# igraph plot options
igraph.options(vertex.size=5, edge.width=0.75)
# plot network
plot(net, layout=coords_fr, vertex.color=V(net)$color)

```



## 4 Findings

1. Clustering based on the graphlet spectra helps finding nodes with similar set of associations in a networks (in terms of the graphlets).
2. Different clustering methods show similar results with some differences. To check that one can open the report twice in two different windows and compare the two visualizations of clustering.

## References

Girvan, M., and M. E. J. Newman. 2002. "Community Structure in Social and Biological Networks." *Proceedings of the National Academy of Sciences* 99 (12). National Academy of Sciences:7821–6. <https://doi.org/10.1073/pnas.122653799>.