

Lab 7 Data Visualization SNA

Aryan Vigyat 20BCE1452

2023-02-13

Attaching the Libraries

```
library(sna)
```

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

```
## Loading required package: statnet.common
```

```
## Warning: package 'statnet.common' was built under R version 4.2.2
```

```
##  
## Attaching package: 'statnet.common'
```

```
## The following objects are masked from 'package:base':  
##  
## attr, order
```

```
## Loading required package: network
```

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

```
##  
## 'network' 1.18.1 (2023-01-24), part of the Statnet Project  
## * 'news(package="network")' for changes since last version  
## * 'citation("network")' for citation information  
## * 'https://statnet.org' for help, support, and other information
```

```
## sna: Tools for Social Network Analysis  
## Version 2.7-1 created on 2023-01-24.  
## copyright (c) 2005, Carter T. Butts, University of California-Irvine  
## For citation information, type citation("sna").  
## Type help(package="sna") to get started.
```

```
library(igraph)
```

```
##  
## Attaching package: 'igraph'
```

```
## The following objects are masked from 'package:sna':  
##  
##   betweenness, bonpow, closeness, components, degree, dyad.census,  
##   evcent, hierarchy, is.connected, neighborhood, triad.census
```

```
## The following objects are masked from 'package:network':  
##  
##   %c%, %s%, add.edges, add.vertices, delete.edges, delete.vertices,  
##   get.edge.attribute, get.edges, get.vertex.attribute, is.bipartite,  
##   is.directed, list.edge.attributes, list.vertex.attributes,  
##   set.edge.attribute, set.vertex.attribute
```

```
## The following objects are masked from 'package:stats':  
##  
##   decompose, spectrum
```

```
## The following object is masked from 'package:base':  
##  
##   union
```

1. Import and load the dataset and view information about it using str() function.

```
df <- read.csv("~/socialnetworkdata.csv")  
data <- data.frame(df$first, df$second)  
str(data)
```

```
## 'data.frame':   290 obs. of  2 variables:  
## $ df.first : chr  "AA" "AB" "AF" "DD" ...  
## $ df.second: chr  "DD" "DD" "BA" "DA" ...
```

2. Create the graph network using graph.data.frame() function.

```
graph <- graph.data.frame(data)  
V(graph)
```

```
## + 52/52 vertices, named, from 74589e3:  
## [1] AA AB AF DD CD BA CB CC BC ED AE CA EB BF BB AC DC BD DB CF DF BE EA CE EE  
## [26] EF FF FD GB GC GD AD KA KF LC DA EC FA FB DE FC FE GA GE KB KC KD KE LB LA  
## [51] LD LE
```

```
E(graph)
```

```
## + 290/290 edges from 74589e3 (vertex names):
## [1] AA->DD AB->DD AF->BA DD->DA CD->EC DD->CE CD->FA CD->CC BA->AF CB->CA
## [11] CC->CA CD->CA BC->CA DD->DA ED->AD AE->AC AB->BA CD->EC CA->CC EB->CC
## [21] BF->CE BB->CD AC->AE CC->FB DC->BB BD->CF DB->DA DD->DA DB->DD BC->AF
## [31] CF->DE DF->BF CB->CA BE->CA EA->CA CB->CA CB->CA CC->CA CD->CA BC->CA
## [41] BF->CA CE->CA AC->AD BD->BE AE->DF CB->DF AC->DF AA->DD AA->DD AA->DD
## [51] CD->DD AA->DD EE->DD CD->DD DB->AA AA->FC BE->CC EF->FD CF->FE BB->DD
## [61] CD->DD BA->AB CD->EC BE->EE CE->CC CD->CC ED->CC BB->CC BE->CE DD->CE
## [71] AC->CD ED->CD FF->CD AC->CD DD->CD DD->CD AE->GA AE->GA AE->GA AE->GA
## [81] BA->ED BE->ED EB->ED CD->ED FD->EF FD->EF CD->BB BF->BB BC->BB BB->CF
## [91] AE->AC DD->DA BE->CA BE->CA CB->CA CB->CA CC->CA BE->CC BE->CC DB->DD
## + ... omitted several edges
```

```
igraph::degree(graph)
```

```
## AA AB AF DD CD BA CB CC BC ED AE CA EB BF BB AC DC BD DB CF DF BE EA CE EE EF
## 18 9 23 36 40 26 24 50 21 27 15 62 7 12 23 27 2 4 8 12 23 20 8 10 6 8
## FF FD GB GC GD AD KA KF LC DA EC FA FB DE FC FE GA GE KB KC KD KE LB LA LD LE
## 1 8 1 1 1 9 3 3 1 7 3 1 1 2 1 2 5 1 1 1 1 1 1 1 1
```

```
igraph::betweenness(graph)
```

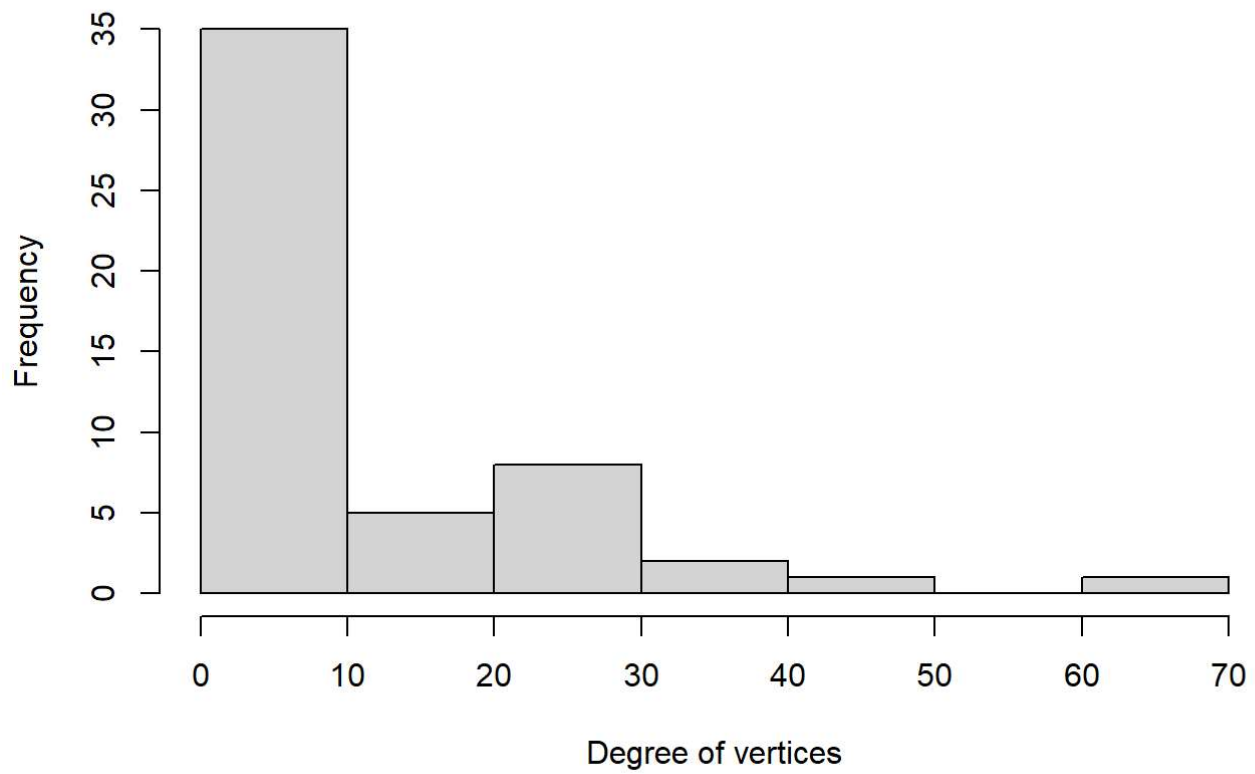
	AA	AB	AF	DD	CD	BA	CB
##	89.196239	4.564734	109.762225	162.990640	375.992047	184.496219	36.237779
	CC	BC	ED	AE	CA	EB	BF
##	225.269107	94.821412	343.016093	105.368424	292.157605	0.000000	28.446981
	BB	AC	DC	BD	DB	CF	DF
##	244.963181	172.427422	0.000000	5.966115	0.000000	181.814414	42.768995
	BE	EA	CE	EE	EF	FF	FD
##	55.964872	183.604022	11.943347	4.002137	0.000000	0.000000	33.000000
	GB	GC	GD	AD	KA	KF	LC
##	0.000000	0.000000	0.000000	48.996857	0.000000	17.229134	0.000000
	DA	EC	FA	FB	DE	FC	FE
##	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	GA	GE	KB	KC	KD	KE	LB
##	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	LA	LD	LE				
##	0.000000	0.000000	0.000000				

```
V(graph)$label <- V(graph)$name
V(graph)$degree <- igraph::degree(graph)
```

3. Plot the Histogram graph For Node Degrees.

```
hist(V(graph)$degree,main='Histogram of node degree',xlab='Degree of vertices',ylab='Frequency')
```

Histogram of node degree



4. Plot the Network Graph Diagram.

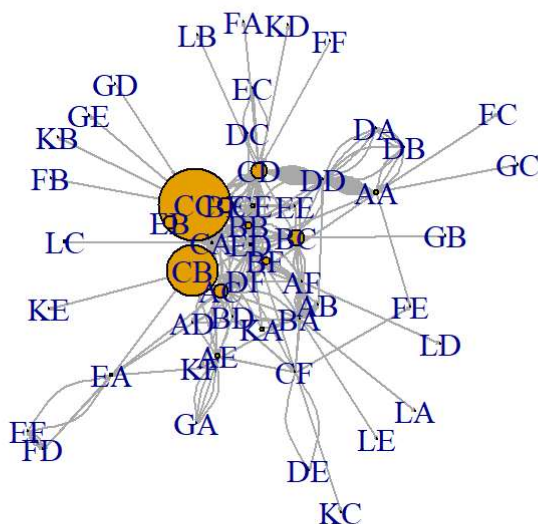
```
plot(graph,vertex.label.cex=0.7,vertex.size=2,edge.arrow.size=0.1)
```

```
as <- authority.score(graph)$vector
as
```

##	AA	AB	AF	DD	CD	BA
##	4.412541e-03	3.968841e-03	8.375004e-02	1.409698e-01	7.894656e-02	2.146204e-02
##	CB	CC	BC	ED	AE	CA
##	3.660444e-03	1.121941e-01	6.326094e-03	4.799095e-02	1.524499e-02	1.000000e+00
##	EB	BF	BB	AC	DC	BD
##	5.863552e-17	6.072030e-03	8.323026e-02	6.262068e-02	1.832360e-17	2.486679e-04
##	DB	CF	DF	BE	EA	CE
##	7.329440e-17	8.146610e-03	1.925650e-01	1.095155e-01	7.776767e-03	2.437300e-02
##	EE	EF	FF	FD	GB	GC
##	2.419075e-02	6.390996e-05	9.161800e-18	2.071608e-03	9.161800e-18	9.161800e-18
##	GD	AD	KA	KF	LC	DA
##	9.161800e-18	1.806051e-02	2.565304e-17	2.156359e-03	9.161800e-18	3.583779e-03
##	EC	FA	FB	DE	FC	FE
##	2.468362e-02	8.227873e-03	3.500976e-02	4.973358e-04	2.815768e-03	3.064436e-03
##	GA	GE	KB	KC	KD	KE
##	1.391587e-02	3.500976e-02	3.500976e-02	2.486679e-04	8.227873e-03	2.501759e-02
##	LB	LA	LD	LE		
##	8.227873e-03	1.133971e-03	1.656466e-04	1.133971e-03		

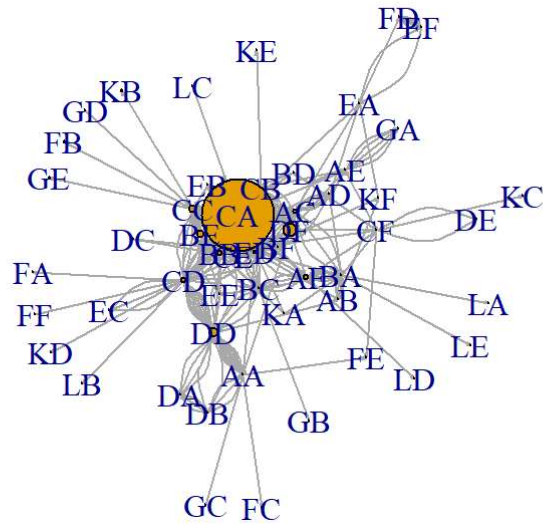
```
plot(graph,vertex.size=hs*30,main='Hubs',edge.arrow.size=0.1,layout=layout.fruchterman.reingold)
```

Hubs



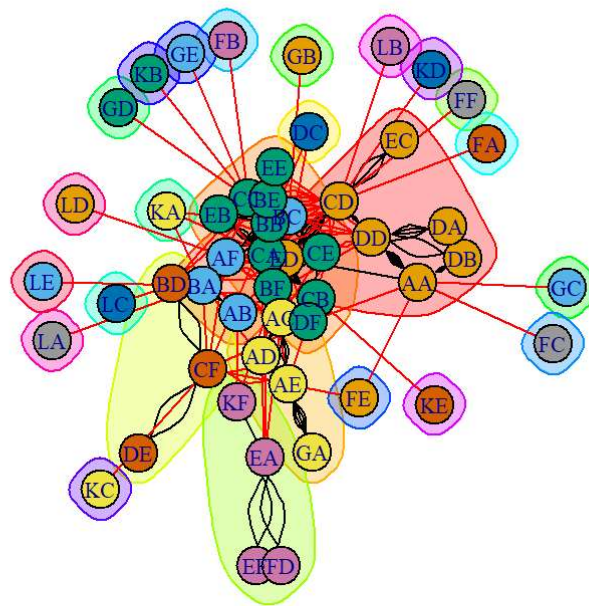
```
plot(graph,vertex.size=as*30,main='Authorities',edge.arrow.size=0.1,layout=layout.fruchterman.reingold)
```

Authorities



6. Find out the communities present in the network.

```
network <- graph.data.frame(data,directed=F)
cnetwork <- cluster_edge_betweenness(network)
plot(cnetwork,network,vertex.label.cex=0.7)
```



(vii) Conclusion

#In social network research, the SNA and igraph packages in R were utilized to analyze a data set with network data involving different nodes. By computing the hub and authority scores of each node, a network diagram was created. The identification of communities in the network revealed that nodes within the same group were highly connected, while nodes between different groups were only weakly connected.