## **Social Network Project**

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#### Introduction

Social Network data analysis is now an increasingly popular subject. With the emergency and fast development of various social media platforms, human connectivity and interaction are reaching a new level. People can connect with each via different means, not only between friends, but also friends of friends, and sometimes event strangers or celeberties, whom we don't normally interact with. Besides such informational social connectivity, network analysis is also essential in medical field. The current Covid-19 Pandemic proved the importance of social network data analysis. The surge of the network data reflects technological, economical, social, political and biological growth.

```
## Load packages
library("igraph")

##
## Attaching package: 'igraph'

## The following objects are masked from 'package:stats':

##
## decompose, spectrum

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

##
## union
```

#### **Data Overview**

Using the github link provided in class, I choose the Game of Throne Dataset. As a fan of the series, I know how complicated the social network is among all the characters. It is interesting to apply social network analysis knowledge on this fantastic story line. The nodes csv contains 107 different characters, and the edges csv contains 353 edges. All the links are weighted relationships between those characters, which were calculated based on how many times two characters' names appeared within 15 words of one another in the novel.Based on the book/TV show settings, all the characters were scattered geographically across the land, and each area has a dominant or ruling family or noble household.

```
setwd("~/Desktop/Social Network/project/datasets")
got_nodes<-read.csv("got-nodes.csv")
got_links<-read.csv("got-edges.csv")</pre>
```

```
## Examine the data
head(got_nodes)
##
          Ιd
               Label
## 1
       Aemon
               Aemon
## 2
       Grenn
               Grenn
## 3 Samwell Samwell
## 4
      Aerys
              Aerys
## 5
       Jaime
               Jaime
## 6 Robert Robert
head(got_links)
##
     Source Target Weight
## 1 Aemon
              Grenn
                         5
## 2 Aemon Samwell
                        31
## 3 Aerys
                        18
              Jaime
## 4 Aerys Robert
                         6
## 5 Aerys
            Tyrion
                         5
## 6 Aerys
            Tywin
                         8
```

### **Analysis**

#### Network

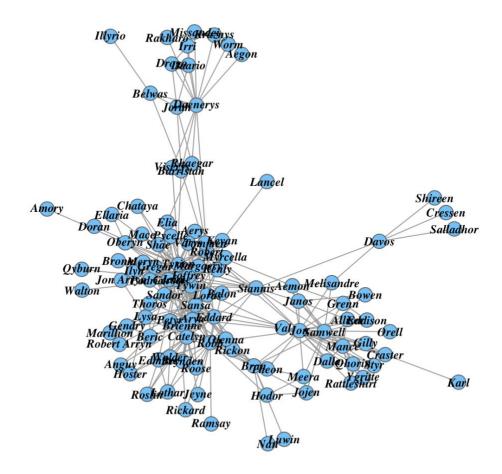
First, I converted nodes and edges to form network.

```
got_net <- graph_from_data_frame(d=got_links, vertices=got_nodes, directed=T)</pre>
# Examine the resulting object:
class(got_net)
## [1] "igraph"
got_net
## IGRAPH 8e08c61 DN-- 107 352 --
## + attr: name (v/c), Label (v/c), Weight (e/n)
## + edges from 8e08c61 (vertex names):
## [1] Aemon ->Grenn
                           Aemon ->Samwell
                                              Aerys ->Jaime
                                                                 Aerys
>Robert
                                              Alliser->Mance
## [5] Aerys
              ->Tyrion
                                  ->Tywin
                                                                 Amory
                           Aerys
>Oberyn
## [9] Arya
              ->Anguy
                           Arya
                                  ->Beric
                                              Arya
                                                     ->Bran
                                                                 Arya
>Brynden
## [13] Arya
               ->Cersei
                           Arya
                                  ->Gendry
                                              Arya
                                                     ->Gregor
                                                                 Arya
>Jaime
## [17] Arya
               ->Joffrey
                           Arya
                                              Arya
                                                     ->Rickon
                                  ->Jon
                                                                 Arya
>Robert
## [21] Arya
                                  ->Sandor
                                                     ->Thoros
              ->Roose
                           Arya
                                             Arya
                                                                 Arya
```

```
>Tyrion
## [25] Balon ->Loras
                          Belwas ->Barristan Belwas ->Illyrio
                                                               Beric -
>Anguy
## [29] Beric ->Gendry
                          Beric ->Thoros
                                             Bran
                                                    ->Hodor
                                                               Bran
>Jojen
## + ... omitted several edges
# We can access the nodes, edges, and their attributes:
head(E(got net)) #edge
## + 6/352 edges from 8e08c61 (vertex names):
## [1] Aemon->Grenn Aemon->Samwell Aerys->Jaime Aerys->Robert Aerys-
>Tyrion
## [6] Aerys->Tywin
head(V(got_net)) #vertex
## + 6/107 vertices, named, from 8e08c61:
## [1] Aemon Grenn Samwell Aerys Jaime Robert
```

### **Network Graph**

Since there are many characters and links, it is better to simplify the graph by removing the multiple connections and self loops.



The plot itself is not very helpful at demonstrating the network relations. There are many nodes and connections. Therefore, other means of measuring social network is implemented. ## Layouts Since the plot above is still very complicated, different layout structures might help to demostrate the network more clearly.

```
par(mfrow=c(1,2))
# random Layout
plot(got_net, edge.arrow.size=0.05,vertex.color="lightskyblue",
vertex.frame.color="#777777",vertex.frame.color="white",vertex.label.color="black",
```

```
vertex.label.cex=0.8,vertex.size=7,vertex.label.font=4,layout=layout_randomly
(got_net))
## sphere Layout
plot(got_net, edge.arrow.size=0.05,vertex.color="lightskyblue",
vertex.frame.color="#777777",vertex.frame.color="white",vertex.label.color="black",

vertex.label.cex=0.8,vertex.size=7,vertex.label.font=4,layout=layout_on_sphere(got_net))
```





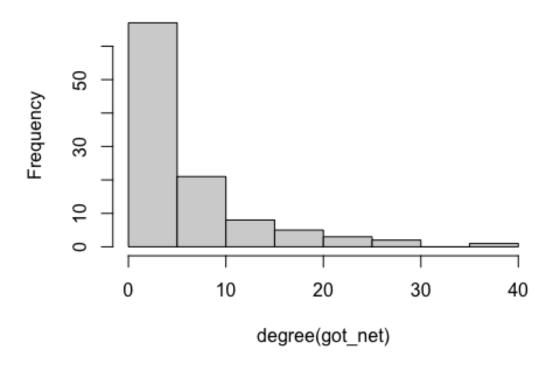
From comparing original plot and different layouts, it seems sphere layout fits this network better. However, the network is still too complicated.

### Cut off by weight

Since Weight is calculated based on how many times two characters' names appeared within 15 words of one another in the novel, most of the characters were mentioned quite often, according to the historgram below. On average, they were mentioned 12 times; therefore, we can try to further simplify the graph by deleting edges with few weights.

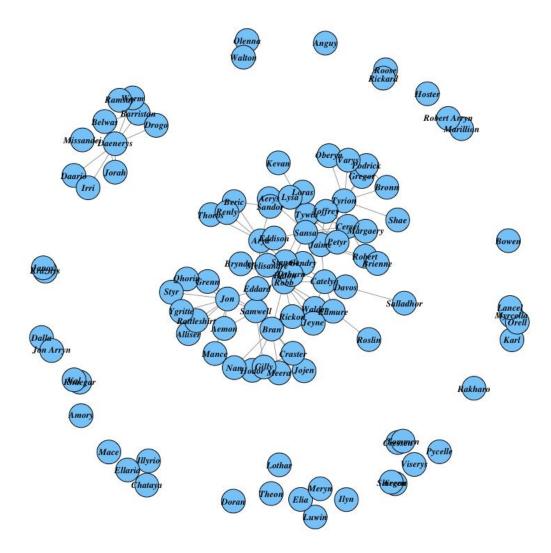
```
hist(degree(got net))
```

# Histogram of degree(got\_net)



```
cut.off<-mean(got_links$Weight)
#net.sp<-delete_edges(got_net, E(got_net)[Weight<cut.off])
#plot(net.sp,edge.arrow.size=0.05,vertex.color="lightskyblue",,vertex.label.c
olor="black",

#vertex.label.cex=0.8,vertex.size=10,vertex.label.font=4,layout=layout_with_k</pre>
```



From the plot above, it is clear that certain people are in the center of the Game of Thrones network. The rest of the characters revolve around the central cluster.

### **Network Description**

The network visualization is not particularly helpful since there are so many nodes and edges entangled together. Mathematically, there is a way of describing a node's characteristics of the network. A common goal in network analysis is to identify "central"

nodes in a network. The central nodes may be passing important information, or are connected to many other nodes, or in vital positions where removing them would change the network structure.

### **Degree Centrality**

Degree is simplest of the methods, it measures the number of connections between a node has to all other nodes. This measure shows the number of direct connections each node has.

<pre>sort(degree(got_net, v=V(got_net)))</pre>							
##	Amory	Illyrio	Karl	Aegon	Kraznys		
Rakharo		4	4	4	4		
## 1	1	1	1	1	1		
##	Worm	Cressen	Salladhor	Qyburn	0rell		
Lancel							
## 1	1	1	1	1	1		
##	Ramsay	Shireen	Doran	Walton	Anguy		
Luwin	_						
##	1	1	1	1	2		
2 ##	Nan	Jevne	Missandei	Bowen	Styr	Jon	
Arryn	NGII	Jeyne	HISSANACI	Dowell	Jey.	3011	
##	2	2	2	2	2		
2 ##	Olenna	Ellaria	Rickard	Chataya	Alliser		
Hoster	OTEIIIIa	LIIai Ia	NICKAIU	Chacaya	ATTISEI		
##	2	2	2	2	3		
3 ##	Viserys	Eddison	Dalla	Manillian	Robert Arryn		
## Mace	viserys	Eddison	Dalla	Martition	Robert Arryll		
##	3	3	3	3	3		
3		_		_			
## Hodor	Grenn	Aerys	Gendry	Roose	Belwas		
##	4	4	4	4	4		
4							
## Daario	Jojen	Theon	Bronn	Roslin	Pycelle		
##	4	4	4	4	4		
4							
## Do++los	Drogo	Irri	Gilly	Myrcella	Melisandre		
Rattles ##	nirt 4	4	4	4	4		
4	-	-	_	_	_		
##	Val	Ygritte	Aemon	Thoros	Meera		
Podrick							

## 5	4	4	5	5	5	
## Tommen	Lothar	Elia	Shae	Craster	Davos	
## 5	5	5	5	5	5	
## Ilyn	Qhorin	Beric	Rickon	Balon	Barristan	
## 6	5	6	6	6	6	
## Brienne	Jorah	Rhaegar	Janos	Kevan	0beryn	
## 7	6	6	6	6	7	
## Walder	Petyr	Meryn	Varys	Margaery	Brynden	
##	7	7	7	7	8	
## Gregor	Edmure	Renly	Loras	Lysa	Mance	
## 12	8	8	9	10	12	
## Samwell	Eddard	Sandor	Bran	Stannis	Daenerys	
## 15	12	13	14	14	14	
## Tywin	Robert	Joffrey	Catelyn	Arya	Cersei	
## 22	18	18	18	19	20	
## ##	Jaime 24	Robb 25	Jon 26	Sansa 26	Tyrion 36	

From the output shown above, Tyrion has the largest degree, indicating he has the most ties with the rest of the characters.

#### **Closeness Centrality**

Closeness centrality is an evaluation of the proximity of a node to all other nodes in a network, not only the nodes to which it is directly connected. It is calculated based on average geodesic distance. Also, closeness centraliy measures how easily other vertices can be reached from it or vice versa.

```
sort(closeness(got_net, mode="all", weight=NA, normalized=F))
##
        Illyrio
                        Karl
                                  Cressen
                                             Salladhor
                                                            Shireen
Amory
## 0.002100840
                0.002358491 0.002358491 0.002358491 0.002358491
0.002469136
         Aegon
                     Kraznys
                                  Rakharo
                                                  Worm
                                                          Missandei
```

Daario ## 0.002531646	0.002531646	0.002531646	0.002531646	0.002538071	
0.002551020 ## Drogo	Irri	Lancel	Belwas	Jorah	
Luwin	11.1.1	Lancer	Deiwas	Joran	
## 0.002551020 0.002808989	0.002551020	0.002604167	0.002695418	0.002717391	
	Bowen	Roslin	Jojen	Hoster	
Qyburn ## 0.002808989	0.002840909	0.002949853	0.002958580	0.002967359	
0.003067485					
## Orell Anguy	Walton	Styr	Rattleshirt	Ygritte	
## 0.003067485	0.003067485	0.003076923	0.003095975	0.003095975	
0.003105590 ## Eddison	Ramsay	Grenn	Rickard	Gilly	
Qhorin ## 0.003105590	0 003105590	0 003115265	0 003115265	0 003125000	
0.003125000					
## Craster Jeyne	Davos	Gendry	Dalla	Lothar	
## 0.003134796	0.003134796	0.003144654	0.003154574	0.003164557	
0.003164557 ## Beric	Hodor	Meera	Doran	Olenna	
Chataya	0.002404000	0.003404000	0.003305430	0 002245424	
## 0.003194888 0.003215434	0.003194888	0.003194888	0.003205128	0.003215434	
## Ellaria Arryn	Alliser	Marillion	Robert Arryn	Bronn	Jon
## 0.003225806	0.003236246	0.003246753	0.003246753	0.003257329	
0.003257329 ## Mace	Tommen	Pycelle	Tlvn	0beryn	
Varys		•	_	•	
## 0.003289474 0.003355705	0.003300330	0.003311258	0.003322259	0.003333333	
## Rhaegar	Mance	Roose	Thoros	Meryn	
Daenerys ## 0.003355705	0.003378378	0.003378378	0.003412969	0.003412969	
0.003448276 ## Melisandre	Walder	Elia	Edmure	Shae	
Viserys	Watuei	LIIa			
## 0.003460208 0.003508772	0.003484321	0.003484321	0.003496503	0.003508772	
## Myrcella	Theon	Val	Brynden	Margaery	
Podrick ## 0.003508772	0.003521127	0.003521127	0.003546099	0.003546099	
0.003558719					
## Rickon Aemon	Barristan	Kevan	Samwell	Brienne	
## 0.003571429	0.003584229	0.003584229	0.003623188	0.003663004	

0.0	03676471					
##	Loras	Aerys	Lysa	Petyr	Janos	
Gre	gor					
##	0.003676471	0.003703704	0.003703704	0.003731343	0.003731343	
0.0	03802281					
##	Balon	Renly	Sandor	Bran	Joffrey	
Cat	elyn					
##	0.003831418	0.003921569	0.003937008	0.003968254	0.004149378	
0.0	04166667					
##	Cersei	Eddard	Tywin	Jaime	Jon	
Stannis						
##	0.004184100	0.004347826	0.004424779	0.004524887	0.004524887	
0.004524887						
##	Arya	Robb	Robert	Sansa	Tyrion	
##	0.004587156	0.004608295	0.004716981	0.004807692	0.004830918	

As seen above, the range of closeness centralities are very small. For all the nodes, their closeness centralities are not far different from each other as well. The resultes indicate that all the nodes are closely connected with each other. Among those nodes, Illyrio has the smallest value of 0.0021 meaning he is in a very central position, and able to reach everybody quickly.

#### **Betweenness Centrality**

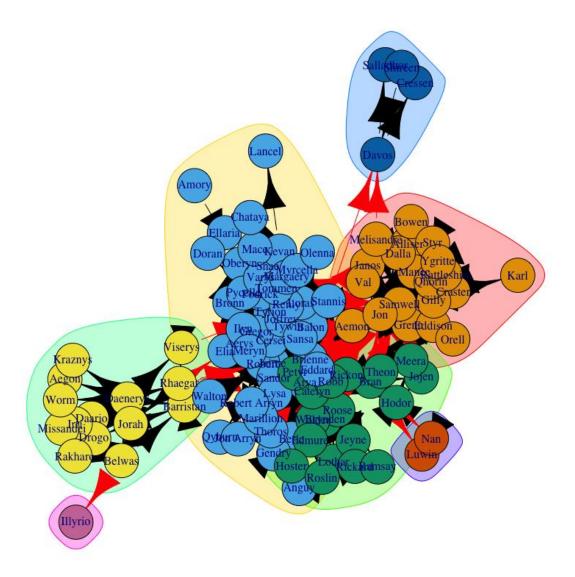
Betweenness centrality measures the number of times a node lies on the shortest path between other nodes. It shows which nodes are "bridges" between nodes in a network. This is calculated by identifying all the shortest paths and counting number of times each node falls on one.

<pre>sort(betweenness(got_net),decreasing=T)</pre>							
##	Tyrion	Samwell	Stannis	Robert	Mance		
Jai ##	me 332.9746032	244.6357143	226.2047619	208.6230159	138.6666667		
119.9956349							
##	Sandor	Jon	Janos	Aemon	Davos		
Lys		111 20000	00 6500000	64 5076100	F.4. 0000000		
## 50.	114.3333333 6166667	111.2666667	90.6500000	64.5976190	54.0000000		
##	Tywin	Gregor	Renly	Cersei	Craster		
Sansa							
##	50.4714286	48.5166667	42.0166667	38.1746032	35.0000000		
32. ##	8428571 Joffrey	Bran	Loras	Viserys	Edmure		
Rob	•	D. dil	20. 43	v13c. y3	Lamare		
##	31.6166667	30.9000000	28.4000000	26.8333333	22.2500000		
19.8718254							
## Wal	Kevan der	Beric	Arya	Varys	Jorah		
##	17.7000000	16.0833333	15.5900794	11.0000000	10.0000000		

9.9166667	5 .	<b>-</b> •		01	
•	Brynden	Jojen	Meera	Oberyn	
Catelyn ## 9.9166667	9.2500000	8.0000000	8.0000000	7.5000000	
6.5678571	3.2300000	8.000000	8.0000000	7.3000000	
## Melisandre	Belwas	Val	Brienne	Hoster	
Balon	Beings		D. 20c		
## 4.5000000	4.0000000	4.0000000	3.9166667	3.6666667	
3.0000000					
## Daario	Lothar	Shae	Podrick	Tommen	
Irri					
## 3.0000000	1.6666667	1.4500000	1.2500000	1.1500000	
1.0000000					
## Rickon	Hodor	Bronn	Meryn	Roose	
Myrcella	0 022222	0 500000	0 500000	0 222222	
## 0.9166667	0.8333333	0.5000000	0.5000000	0.3333333	
0.3333333 ## Grenn	Aonyo	Alliser	Amory	Angus	
Gendry	Aerys	ATTISEL	Alliory	Anguy	
## 0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	
0.0000000	0.000000	0.000000	0.000000	0.000000	
## Thoros	Barristan	Illyrio	Luwin	Nan	
Theon		,			
## 0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	
0.0000000					
## Jeyne	Petyr	Roslin	Elia	Ilyn	
Pycelle					
## 0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	
0.0000000	_	_	_		
## Karl	Drogo	Daenerys	Aegon	Kraznys	
Missandei ## 0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	
0.0000000	0.000000	0.000000	0.000000	0.000000	
## Rakharo	Worm	Cressen	Salladhor	Eddard	
Eddison	WOTIII	Ci CSSCII	Sattaanor	Ladara	
## 0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	
0.0000000					
## Gilly	Qyburn	Bowen	Margaery	Dalla	
Orell					
## 0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	
0.0000000					
_	Rattleshirt	Styr	Ygritte	Jon Arryn	
Lancel					
	0.0000000	0.0000000	0.0000000	0.0000000	
0.0000000 ## 0lanna	Manillia	Dobort Arres	F11	M	
## Olenna Rickard	mar11110n	Robert Arryn	Ellaria	Mace	
## 0.0000000	0.0000000	0.0000000	0.0000000	0.0000000	
0.0000000	0.000000	0.000000	0.000000	0.000000	
0.000000					

# **Community Detection**

```
library(anocva)
## Warning: package 'anocva' was built under R version 4.0.2
groups<-cluster_walktrap(got_net)
mem<-membership(groups)
table(mem)
## mem
## 1 2 3 4 5 6 7
## 20 47 19 14 4 2 1
plot(groups,got_net)</pre>
```



### **Results**

According to the centrality analysis, Tyrion is shown to be an important node in the Game of Throne social network from the result output. Illyrio has the small closeness centrality measure. As shown in the above plot and layout of the network, nodes and edges are closely entangled in the network and it is hard to clearly identify clusters. The data itself did not specify household or geographical locations those characters belong to. Therefore, community detection is useful in this dataset. As shown above, there are 7 membership

groups. Group 2 has the most people. From the plot above, group 2 is in the center, connecting to different outer clusters.

#### **Conclusion**

This result suggests that Tyrion has the most links to other people. At the same time, Tyrion lies most times on the shortest path between other nodes. On the other hand, Illyrio's small closeness measure indicates he is close to other nodes. Yet, since this is a highly and tightly connected network, all the characters are closely linked. As seen in the above output, all the nodes' scores are very small and not that much different from each other It is not surprising that Tyrion has the highest degree and betweenness centrality score. In the original story line, he is knowledgeable person with extensive information of the realm, of the people, and of the culture, history, etc. As shown in the centrality measures, Tyrion is a particularly important node who could influence and perhaps has the authority over the network flow. Since this dataset does not contain any information other than names and links between these names, community detection is helpful. Ideally groups of nodes with some similarity among them are categorized within a community. Clusters of tightly connected groups of nodes are grouped in the same community. According to the plot, 7 communities are detected. Theses nodes are grouped by proximity from other nodes. For instance, nodes in the red circle would not be in the same community with nodes in the yellow circle. With Tyrion being the most connected character, all the characters in the Game of Thrones series are divided into 7 communities. Without prior Game of Thrones knowledge, it is beneficial to apply community detection techniques and see how communities are determined within the network. Additionally, the different degree algorithms can easily show the connectivity for a single person within a network. The degree and betweenness centrality identification can further enable one's understanding of the extensive connectivity a person has. On the other hand, closeness centrality may not be an optimal measure. In a highly connected network, all nodes will have similar score and this phenomenon is also seen in the present analysis. It may be useful to using closeness to find influencers in a single cluster, rather than the whole network. Another limitation is that this dataset does not contain any geographical or household information. It is hard to compare the result from community detection with the real community.