HW 4 Report

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Introduction

The affiliation network is a network where members are affiliated with one another based on co-membership in a group, or co-participation in some type of event. Incidence matrices, edge lists, plotting and projections are four base components for affiliation networks. The incidence matrix describes how n actors belong to g groups. Unlike adjacency matrix, it has two dimensions: actors and groups. The related affiliation networks are also known as two-mode network. In the graph level, ties connect the actors to groups and no direct ties among actors, nor groups. One of the important attributes is vertex.type which is equal to FALSE if the vertex belongs to actors and equal to TRUE if the vertex belongs to groups. Edge lists can also be translated into an affiliation network, as long as nodes of actors are only connected to nodes of groups. Visual inspection can be helpful to understand affiliation networks. It is often useful to examine the connections among actors only (or groups only) which can be realized by extracting and visualizing the one-mode projections of the two-mode affiliation network.

I Repeat the analysis

a. Preparation for the data

Initially, we need to load the data and represent the original data hwd as h1. Also, general information of h1 will be showed below.

```
data(hwd) # Load the data
h1 <- hwd # represent original data hwd as h1
h1 # show the information of h1
## IGRAPH 9cdab39 UN-B 1365 1600 --
## + attr: name (v/c), type (v/l), year (v/n), IMDBrating (v/n),
## | MPAArating (v/c)
## + edges from 9cdab39 (vertex names):
## [1] Inception
                          --Leonardo DiCaprio
## [2] Inception
                          --Joseph Gordon-Levitt
## [3] Inception
                          --Ellen Page
## [4] Inception
                          --Tom Hardy
## [5] Inception
                          --Ken Watanabe
## [6] Inception
                          --Dileep Rao
## [7] Inception
                          --Cillian Murphy
## + ... omitted several edges
```

Then we will use length(.) and V(h1)\$name function to show the length of the vertex.name which represents the number of vertices.

```
length(V(h1)$name) # show the length of the vertex.name which represents th
e number of vertex
## [1] 1365
```

As showed above, there are too many vertices in this network, so we will only check first 10 and No.155-No.165 node's property by function V(h1)\$name and V(h1)\$type.

From name and type attributes we know that the first 10 nodes all represent movie names.

[1] TRUE TRUE TRUE TRUE TRUE TRUE FALSE FALSE

Summary:

Generally, the summary description of h1 indicates that hwd is indeed a bipartite graph. We can surmise that the ties link each actor to the movie that actor was in. The results also reveal that the network has 1,365 nodes and 1,600 ties. This is a little harder to decipher for an affiliation network but given what we already know we can figure out that there are 160 movie nodes, 1,205 actor nodes, and the 1,600 ties arise from each movie having links to just ten

actors. (There are only 1,205 actors listed because some actors appear in more than one movie.)

Also, the entire network is very large! It is difficult to visualize the whole network so we can show a subgraph of the whole network.

b. A typical subnetwork

There are three well-known movies, with actor Leonardo DiCaprio, including "The Departed", "Gangs of New York" and "The Wolf of Wall Street". And we will analyze this subnetwork.

Firstly, we will use V(h1)\$shape, V(h1)\$type and ifelse function to assign values to the shape attribute of network h1. If the value of type attribute is equal to TRUE, the corresponding value of shape attribute will be "square". Otherwise, the corresponding value of shape attribute will be "circle".

V(h1)\$shape <- ifelse(V(h1)\$type==TRUE, "square", "circle") # Assign values to the shape attribute of network h1. If the value of type attribute is equ al to TRUE, the corresponding value of shape attribute will be "square". Ot ehrwise, the corresponding value of shape attribute will be "circle".

Next, we will use V(h1)\$color, V(h1)\$type and ifelse function to assign values to the color attribute of network h1. If the value of type attribute is equal to TRUE, the corresponding value of color attribute will be "red". Otherwise, the corresponding value of color attribute will be "lightblue".

V(h1)\$color <- **ifelse(V(h1)\$type==**TRUE, "red","lightblue") # Assign values to the color attribute of network h1. If the value of type attribute is equal to TRUE, the corresponding value of shape attribute will be "red". Otherwise, the corresponding value of shape attribute will be "lightblue".

Then we use **E**(h1)[inc(V(h1)[.])] and **subgraph.edges** function to Create a subgraph,h2, of h1. The mechanism contains two steps. Firstly, obtain a vertex sequence which is the sequence of vertex.name attribute matched with the given three movie names.

```
V(h1)[name %in%
c("The Wolf of Wall Street",
    "Gangs of New York",
    "The Departed")] #Obtain a vertex sequence which is the sequence of verte
x.name attribute matched with the given three mobvie names.

## + 3/1365 vertices, named, from 9cdab39:
## [1] The Wolf of Wall Street The Departed Gangs of New York
```

Secondly, select all edges that have at least one incident vertex in the obtained vertex sequence.

```
E(h1)[inc(V(h1)[name %in%
c("The Wolf of Wall Street",
    "Gangs of New York",
```

"The Departed")])] #Select all edges that have at least one incident vert ex in the obtained vertex sequence.

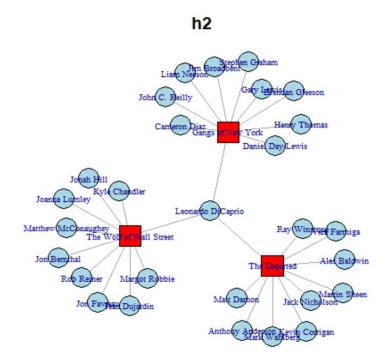
```
## + 30/1600 edges from 9cdab39 (vertex names):
## [1] The Wolf of Wall Street--Leonardo DiCaprio
## [2] The Wolf of Wall Street--Jonah Hill
## [3] The Wolf of Wall Street--Margot Robbie
## [4] The Wolf of Wall Street--Matthew McConaughey
## [5] The Wolf of Wall Street--Kyle Chandler
## [6] The Wolf of Wall Street--Rob Reiner
## [7] The Wolf of Wall Street--Jon Bernthal
## [8] The Wolf of Wall Street--Jon Favreau
## [9] The Wolf of Wall Street--Jean Dujardin
## [10] The Wolf of Wall Street--Joanna Lumley
## + ... omitted several edges
```

Finally, we use **subgraph.edges** function to Create a subgraph,h2, of h1.

h2 <- subgraph.edges(h1, E(h1)[inc(V(h1)[name %in% c("The Wolf of Wall Stre et", "Gangs of New York", "The Departed")])]) # Create a subgraph,h2, of h 1. The mechanism contains two steps. Firstly, obtain a vertex sequence which is the sequence of vertex.name attribute matched with the given three movie names. Secondly, select all edges that have at least one incident vertex in the obtained vertex sequence.

Then, we will plot h2 with layout equal to layout_with_kk.

plot(h2, layout = layout_with_kk, main = "h2") # Plot the network h2 with l
ayout as layout_with_kk pattern.



c. Busy and popular actors analysis

(1) Busy actors

Initially, we analyze the degree first. We use **degree**, **mean**, **V(h1)** and **table** function to show the degree of each vertex representing actor and calculate both frequency and mean of degrees.

```
table(degree(h1,v=V(h1)[type==FALSE])) # Show the degree of each vertex rep
resenting actor and calcaluate the frequency of degrees.
##
## 1 2 3 4 5 6 7 8
## 955 165 47 23 11 2 1 1
```

mean(degree(h1,v=V(h1)[type==FALSE])) # Show the degree of each vertex repr
esenting actor and calcaluate the mean of degrees.

```
## [1] 1.327801
```

From the results, we can know that the vast majority of actors only appeared in one movie, but there were 15 actors who each starred in five or more movies since 1999. Across all the actors, they starred in an average of 1.3 movies. This information can then be used to identify the busiest actors of the past decade and a half.

We create a new vertex attribute named deg by function V(h1)\$deg to save the value of corresponding degree value of the vertex. Also select all vertices which meet both two conditions: representing actor name and deg value being bigger than 4 by function V(h1)[type==FALSE & deg > 4]\$name.

```
V(h1)$deg <- degree(h1) # Create a new vertex attribute named deg to save
the value of corresponding degree value of the vertex.
V(h1)[type==FALSE & deg > 4]$name # Select all vertices which meet both two
conditions: representing actor name and deg value being bigger than 4.
```

```
## [1] "Leonardo DiCaprio" "Emma Watson" "Richard Griffiths"
## [4] "Harry Melling" "Daniel Radcliffe" "Rupert Grint"
## [7] "James Franco" "Ian McKellen" "Martin Freeman"
## [10] "Bradley Cooper" "Christian Bale" "Samuel L. Jackson"
## [13] "Natalie Portman" "Brad Pitt" "Liam Neeson"
```

To get the busy actor, firstly, we create a variable named Actor to save the names of vertices which meet two conditions: representing actor name and deg value being bigger than 4 by function Actor = V(h1)[type==FALSE & deg > 4]\$name. Secondly, we create a variable named Movies to save the degrees of vertices which meet two conditions: representing actor name and deg value being bigger than 4 by function Movies = V(h1)[type==FALSE & deg > 4]\$deg. Finally, combine these two variable by columns to form the data.frame data representing busy actor by function cbind and data.frame.

busy_actor <- data.frame(cbind(Actor = V(h1)[type==FALSE & deg > 4]\$name, Movies = V(h1)[type==FALSE & deg > 4]\$deg)) # Firstly, create a variable named Actor to save the names of vertices which meet two conditions: representing actor name and deg value being bigger than 4. Secondly, create a variable named Movies to save the degrees of vertices which meet two conditions: representing actor name and deg value being bigger than 4. Finally, comb ine these two variable by columns to form the data.frame data representing busy_actor.

Then we use function **order**(busy_actor\$Movies,decreasing=TRUE) to order the busy actor to find the busiest actor.

busy_actor[order(busy_actor\$Movies,decreasing=TRUE),] # Sort the busy_actor by variable movies in a decreasing order.

```
##
                 Actor Movies
## 5
       Daniel Radcliffe
## 11
         Christian Bale
                              7
## 1 Leonardo DiCaprio
                              6
## 2
            Emma Watson
                             6
## 3 Richard Griffiths
                              5
## 4
                             5
          Harry Melling
## 6
           Rupert Grint
                             5
## 7
           James Franco
                             5
## 8
           Ian McKellen
                             5
                              5
## 9
         Martin Freeman
                              5
## 10
         Bradley Cooper
## 12 Samuel L. Jackson
                              5
## 13
        Natalie Portman
                              5
## 14
              Brad Pitt
                             5
## 15
                             5
            Liam Neeson
```

From the results we can tell that Daniel Radcliffe is the busiest actor.

(2) Popular actors

If we want to find the popularist actors, we need to used the vertex attribute IMDBrating. Typically, we need to generate two vertex attributes named totrating and avgrating. To generate the totrating attribute, we need to use <code>for</code>, <code>V(h1)[nei(i)]\$IMDBrating</code>) and <code>sum</code> function. The mechanism is that for vertices representing actors, firstly, select attribute IMDBrating of neighbors of these vertices and calculate the sum of all values of each actor's neighbors' attribute IMDBrating for each actor. Secondly, create a new vertex attribute named totrating to save the calculated sum value of corresponding vertex.

for (i in 161:1365) {V(h1)[i]\$totrating <- sum(V(h1)[nei(i)]\$IMDBrating)}
For vertices representing actors, firstly, select attribute IMDBrating of
neighbors of these vertices and calculate the sum of all values of each</pre>

actor's neighbors' attribute IMDBrating for each actor. Secondly, create a new vertex attribute named totrating to save the calculated sum value of corresponding vertex.

To generate the avgrating attribute, we need to use **for**, V(h1)[nei(i)]\$IMDBrating) and **mean** function. The mechanism is that for vertices representing actors, firstly, select attribute IMDBrating of neighbors of these vertices and calculate the mean of all values of each actor's neighbors' attribute IMDBrating for each actor. Secondly, create a new vertex attribute named avgrating to save the calculated mean value of corresponding vertex.

for (i in 161:1365) {V(h1)[i]\$avgrating <- mean(V(h1)[nei(i)]\$IMDBrating)} # For vertices representing actors, firstly, select attribute IMDBrating of neighbors of these vertices and calculate the mean of all values of each actor's neighbors' attribute IMDBrating for each actor. Secondly, create a new vertex attribute named avgrating to save the calculated mean value of corresponding vertex.

To get the popular actor, firstly, we create a variable named Actor to save the names of vertices which meet two conditions: representing actor name and totrating value being bigger than 40 by function Actor = V(h1)[type==FALSE & totrating > 40]\$name. Secondly, we create a variable named Popularity to save the totrating of vertices which meet two conditions: representing actor name and totrating value being bigger than 40 by function Popularity = V(h1)[type==FALSE &totrating > 40]\$totrating. Finally, combine these two variable by columns to form the data.frame data representing pop_actor by function cbind and data.frame.

pop_actor <- data.frame(cbind(Actor = V(h1)[type==FALSE & totrating > 40]\$n ame,Popularity = V(h1)[type==FALSE &totrating > 40]\$totrating)) # Firstl y, create a variable named Actor to save the names of vertices which meet t wo conditions: representing actor name and totrating value being bigger than 40. Secondly, create a variable named Popularity to save the totrating of vertices which meet two conditions: representing actor name and totrating value being bigger than 40. Finally, combine these two variable by columns to form the data.frame data representing pop_actor.

Then we use function **order**(pop_actor\$Popularity,decreasing=TRUE) to order the pop_actor to find the most popular actor.

pop_actor[order(pop_actor\$Popularity,decreasing=TRUE),] # Sort the pop_act
or by variable Popularity in a decreasing order.

```
## Actor Popularity
## 3 Daniel Radcliffe 60.9
## 4 Christian Bale 55.5
## 1 Leonardo DiCaprio 49.6
```

## 2	Emma Watson	45
## 5	Brad Pitt	40.5

From the results we can tell that Daniel Radcliffe is the most popular actor.

3 Busy actors vs Popular movie

Now we want to know whether busier actors means more popular movies. So we first need to generate three variables: num, avgpop and totpop. Specifically, num saves the corresponding degree of all vertices representing actors; avgpop saves the corresponding avgrating of all vertices representing actors; totpop saves the corresponding totrating of all vertices.

num $\leftarrow V(h1)[type==FALSE]$deg # Create a variable named num to save the corresponding degree of all vertices representing actors.$

avgpop <- V(h1)[type==FALSE]\$avgrating # Create a variable named avgpop to save the corresponding avgrating of all vertices representing actors.

totpop <- V(h1)[type==FALSE]\$totrating # Create a variable named totpop to save the corresponding totrating of all vertices

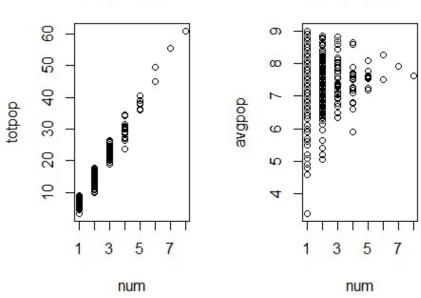
Then we will show the scatter plots of *num vs totpop* and *num vs avgpop*.

plot(num,totpop) # Show a scatter plot with num vs totpop
plot(num,avgpop) # Show a scatter plot with num vs avgpop

scatter plot

Call:

scatter plot

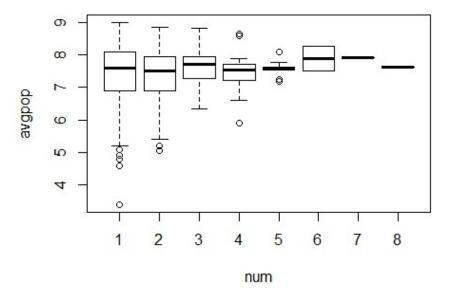


Also, a linear regression and box-plot have been done for this data by function <code>lm(avgpop~num)</code> and <code>boxplot(avgpop~num, main= "Box plot")</code> respectively.

<code>summary(lm(avgpop~num)) # use simple linear regression to analysis dependen t variable avgpop and independent variable num and make a summary of the result</code>

```
## lm(formula = avgpop ~ num)
##
## Residuals:
               1Q Median
##
      Min
                               3Q
                                      Max
## -3.9858 -0.4330 0.1977
                            0.6170
                                   1.6142
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 7.33868
                           0.05440 134.911
                                             <2e-16 ***
               0.04714
                          0.03527
                                    1.337
                                             0.182
## num
## ---
## Signif. codes:
                     '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 '
## Residual standard error: 0.9605 on 1203 degrees of freedom
## Multiple R-squared: 0.001483,
                                    Adjusted R-squared: 0.0006528
## F-statistic: 1.786 on 1 and 1203 DF, p-value: 0.1816
```

Box Plot



Summary:

A simple scatterplot, regression and box plot are examined to see the relationship between number of movies and the average ratings of those movies. The results suggest that there is not a strong relationship between how busy an actor has been and the popularity of their movies. However, the scatterplot also suggests that actors who appear in less popular movies are most likely to appear in only one or two movies.

d. Projections

We use function to h1.pr <- bipartite.projection(h1) project the bipartite graph h1

into two one-mode networks and save them into h1.pr.

```
h1.pr <- bipartite.projection(h1) # Project the bipartite graph h1 into two one-mode networks and save them into h1.pr
```

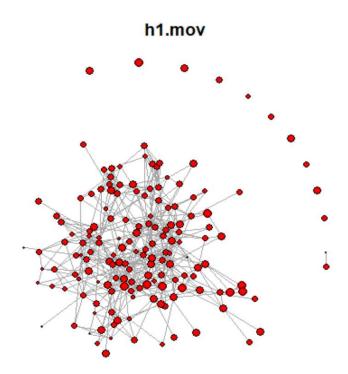
Then we use function h1.act <- h1.pr\$proj1 to save the first projection of h1 into h1.act and function h1.mov <- h1.pr\$proj2 to save the second projection of h1 into h1.mov h1.act <- h1.pr\$proj1 # Save the first projection of h1 into h1.act h1.mov <- h1.pr\$proj2 # Save the second projection of h1 into h1.mov

Now we can see h1.act is an undirected weighted graph for actors and h1.mov is an undirected weighted graph for movies.

```
h1.act # Show h1.act
## IGRAPH d8c301a UNW- 1205 6903 --
## + attr: name (v/c), year (v/n), IMDBrating (v/n), MPAArating
## | (v/c), shape (v/c), color (v/c), deg (v/n), totrating (v/n),
## | avgrating (v/n), weight (e/n)
## + edges from d8c301a (vertex names):
## [1] Leonardo DiCaprio--Joseph Gordon-Levitt
## [2] Leonardo DiCaprio--Ellen Page
## [3] Leonardo DiCaprio--Tom Hardy
## [4] Leonardo DiCaprio--Ken Watanabe
## [5] Leonardo DiCaprio--Dileep Rao
## [6] Leonardo DiCaprio--Cillian Murphy
## + ... omitted several edges
h1.mov # Show h1.mov
## IGRAPH d8c301a UNW- 160 472 --
## + attr: name (v/c), year (v/n), IMDBrating (v/n), MPAArating
## | (v/c), shape (v/c), color (v/c), deg (v/n), totrating (v/n),
## | avgrating (v/n), weight (e/n)
## + edges from d8c301a (vertex names):
## [1] Inception--The Wolf of Wall Street
## [2] Inception--Django Unchained
## [3] Inception--The Departed
## [4] Inception--Gangs of New York
## [5] Inception--Catch Me If You Can
## [6] Inception--The Dark Knight Rises
## + ... omitted several edges
```

Then we can plot h1.mov with vertex color as red, vertex shape as circle, vertex size as the corresponding value of IMBDrating-3 and title as h1.mov.

```
plot(h1.mov,vertex.color="red", # set vertex color as red
  vertex.shape="circle", # set vertex shape as circle
  vertex.size=(V(h1.mov)$IMDBrating)-3, # Set the vertex size by the
```



Next, we use function **graph.density(h1.mov)** to calculate the density of h1.mov and find the result is 0.03710692. We use function **no.clusters(h1.mov)** to calculate the number of clusters of h1.mov and find the result is 12.

```
graph.density(h1.mov) # Calculate the density of h1.mov

## [1] 0.03710692

no.clusters(h1.mov) # Calculate the number of clusters of h1.mov

## [1] 12
```

We also function **clusters**(h1.mov)\$csize to calculate and show the size of each cluster in h1.mov and function **table**(E(h1.mov)\$weight) to calculate and show the frequency of edge attribute named weight

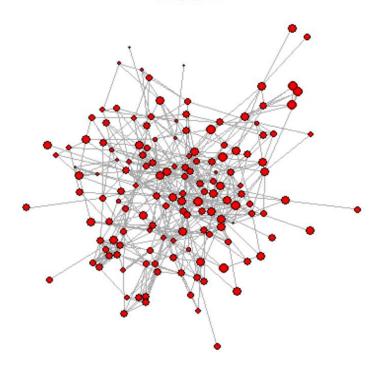
Based on the result of clusters we can generate another subgraph from all nodes in the first cluster of h1.mov by function h2.mov <-induced.subgraph(h1.mov,vids=clusters(h1.mov)\$membership==1).

```
h2.mov <- induced.subgraph(h1.mov,vids=clusters(h1.mov)$membership==1) #
Generate a subgraph by all nodes in the first cluster of h1.mov
```

Then we can plot h2.mov with vertex color as red, vertex shape as circle, vertex size as the corresponding value of IMBDrating-3, edge width as the sqrt of the weight edge attribute and the title as h2.mov.

```
plot(h2.mov,vertex.color="red", # set vertex color as red
edge.width=sqrt(E(h1.mov)$weight), # set edge width as the sqrt of the weight
edge attribute
vertex.shape="circle", # set vertex shape as circle
vertex.size=(V(h2.mov)$IMDBrating)-3, # Set the vertex size by the
corresponding value of IMBDrating-3
vertex.label=NA,, main = "h2.mov")
```

h2.mov



Next, we use function **table(graph.coreness(h2.mov))** to calculate an show the the frequency of coreness of each node in h2.mov and find the largest coreness is 7-cores.

```
table(graph.coreness(h2.mov)) # Calculate the frequency of coreness of eac
h node in h2.mov
## 1 2 3 4 5 6 7
## 11 5 23 65 29 7 8
```

Based on the result of coreness we can generate another subgraph from all nodes all nodes whose coreness is bigger than 4 in h2.mov by function h3.mov <- induce d.subgraph(h2.mov, vids=graph.coreness(h2.mov)>4).

h3.mov <- induced.subgraph(h2.mov, vids=graph.coreness(h2.mov)>4)# Generate a subgraph by all nodes whose coreness is bigger than 4 in h2.mov

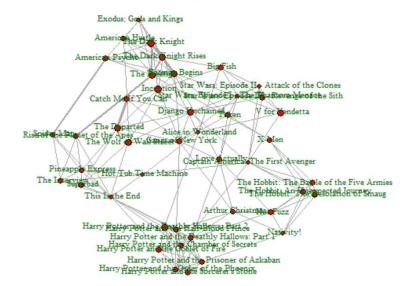
Now we can see h3. mov is an undirected weighted graph for movies.

```
h3.mov #Show h3.mov
## IGRAPH d8eaf7b UNW- 44 158 --
## + attr: name (v/c), year (v/n), IMDBrating (v/n), MPAArating
## | (v/c), shape (v/c), color (v/c), deg (v/n), totrating (v/n),
## | avgrating (v/n), weight (e/n)
## + edges from d8eaf7b (vertex names):
## [1] Alice in Wonderland
                                               --Hot Tub Time Machine
## [2] Hot Tub Time Machine
                                               --The Interview
## [3] The Hobbit: The Battle of the Five Armies--The Hobbit: The Desolatio
n of Smaug
## [4] Inception
                                              --The Wolf of Wall Street
## [5] Exodus: Gods and Kings
                                               --American Hustle
## + ... omitted several edges
```

Then we can plot h3.mov with vertex color as red, vertex shape as circle, vertex size as the corresponding value of IMBDrating-3, edge width as the sqrt of the weight edge attribute, vertex's label as 0.7, the color of vertex's label as darkgreen, the distance of the label from the center of the vertex as 0.3 and the title as h3.mov.

```
plot(h3.mov,vertex.color="red", # set vertex color as red
vertex.shape="circle", # set vertex shape as circle
edge.width=sqrt(E(h1.mov)$weight), # set edge width as the sqrt of the wei
ght edge
vertex.label.cex=0.7, # Set the size of vertex's label as 0.7
vertex.label.color="darkgreen", # Set the color of vertex's label' as dark
green
vertex.label.dist=0.3, #Set the distance of the label from the center of th
e vertex as 0.3
vertex.size=
   (V(h3.mov)$IMDBrating)-3, main = "h3.mov") # Set the vertex size by the
corresponding value of IMBDrating-3
```

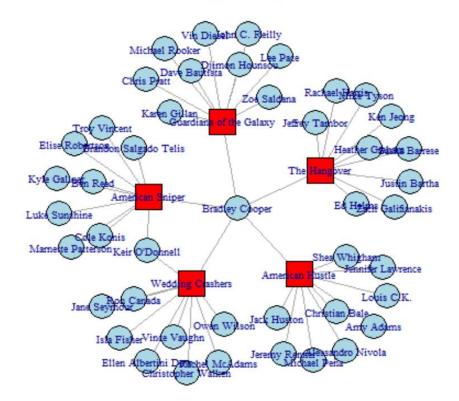
h3.mov



II The similar graph for the actor Bradley Cooper

First, we use function V(h1)[nei(V(h1)\$name == "Bradley Cooper")]\$name to check all movies with Bradley Cooper and find there are five movies related to Bradley Cooper with names as "Guardians of the Galaxy", "American Sniper", "American Hustle" "The Hangover" and "Wedding Crashers". Then we can get the similar plot.

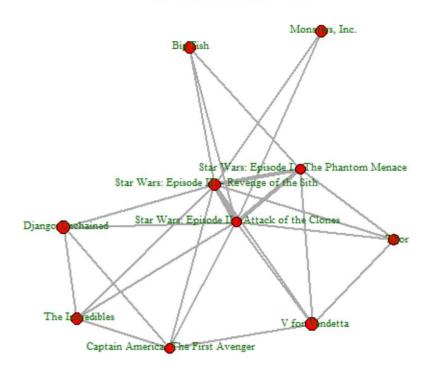
Bradley Cooper



III The similar graph for all the movies related to "Star Wars: Episode III - Revenge of the Sith"

```
First, we use functionH5.mov<-induced.subgraph(h1.mov,vids=V(h1.mov)[
nei(V(h1.mov)$name =="Star Wars: Episode III - Revenge of the Sith" ) |
V(h1.mov)$name == "Star Wars: Episode III - Revenge of the Sith" ]) to find
all movies related to "Star Wars: Episode III - Revenge of the Sith" and then get the plot.
```

All related movies



Appendix

Repeat the analysis for HWD

```
library(UserNetR)
library(igraph)
```

Prepration for the data

```
data(hwd) # Load the data
h1 <- hwd # represent original data hwd as h1
h1 # show the information of h1
length(V(h1)$name) # show the length of the vertex.name which represents the number of vertex

V(h1)$name[1:10] # Show first 10 values of name attribute

V(h1)$type[1:10] # Show first 10 values of type attribute

V(h1)$name[155:165] # Show first No.155-No.165 values of name attribute

V(h1)$type[155:165] # Show first No.155-No.165 values of type attribute</pre>
```

A typical subnetwork

V(h1)\$shape <- ifelse(V(h1)\$type==TRUE, "square","circle") # Assign values
to the shape attribute of network h1. If the value of type attribute is equ
al to TRUE, the corresponding value of shape attribute will be "square". Ot
ehrwise, the corresponding value of shape attribute will be "circle".
V(h1)\$color <- ifelse(V(h1)\$type==TRUE, "red","lightblue") # Assign values
to the color attribute of network h1. If the value of type attribute is equ
al to TRUE, the corresponding value of shape attribute will be "red". Otehr
wise, the corresponding value of shape attribute will be "lightblue".</pre>

h2 <- subgraph.edges(h1, E(h1)[inc(V(h1)[name %in%c("The Wolf of Wall Stree
t", "Gangs of New York", "The Departed")])]) # Create a subgraph, h2, of h
1. The mechanism contains two steps. Fistly, obtain a vertex sequence which
is the sequence of vertex.name attribute matched with the given three mobv
ie names. Secondly, select all edges that have at least one incident vertex
in the obtained vertex sequence.</pre>

V(h1)[name %in%c("The Wolf of Wall Street", "Gangs of New York", "The Depart
ed")] #Obtain a vertex sequence which is the sequence of vertex.name attrib
ute matched with the given three mobvie names.

E(h1)[inc(V(h1)[name %in%c("The Wolf of Wall Street", "Gangs of New York",
 "The Departed")])] #Select all edges that have at least one incident vert
ex in the obtained vertex sequence.

plot(h2, layout = layout_with_kk, vertex.label.cex =0.65, main = "h2") # Pl
ot the network h2 with layout as layout_with_kk pattern.

Density analysis

table(degree(h1,v=V(h1)[type==FALSE])) # Show the degree of each vertex representing actor and calcaluate the frequency of degrees.

mean(degree(h1,v=V(h1)[type==FALSE])) # Show the degree of each vertex representing actor and calcaluate the mean of degrees.

V(h1)\$deg <- degree(h1) # Creat a new vertex attribute named deg to save t he value of corresponding degree value of the vertex.

V(h1)[type==FALSE & deg > 4]\$name # Select all vertices which meet both two
conditions: representing actor name and deg value being bigger than 4.

busy_actor <- data.frame(cbind(Actor = V(h1)[type==FALSE & deg > 4]\$name, Movies = V(h1)[type==FALSE & deg > 4]\$deg)) # Firstly, create a variable named Actor to save the names of vertices which meet two conditions: representing actor name and deg value being bigger than 4. Secondly, create a variable named Movies to save the degrees of vertices which meet two conditions: representing actor name and deg value being bigger than 4. Finally, comb ine these two variable by columns to form the data.frame data representing busy_actor.

busy_actor[order(busy_actor\$Movies,decreasing=TRUE),] # Sort the busy_actor by variable movies in a decreasing order.

for (i in 161:1365) {V(h1)[i]\$totrating <- sum(V(h1)[nei(i)]\$IMDBrating)}
For vertices representing actors, firstly, select attribute IMDBrating of
 neighbors of these vertices and calculate the sum of all values of each ac
 tor's neighbors' attribute IMDBrating for each actor. Secondly, create a ne
 w vertex attribute named totrating to save the calculated sum value of corr
 esponding vertex.</pre>

for (i in 161:1365) {V(h1)[i]\$avgrating <- mean(V(h1)[nei(i)]\$IMDBrating)} # For vertices representing actors, firstly, select attribute IMDBrating of neighbors of these vertices and calculate the mean of all values of each actor's neighbors' attribute IMDBrating for each actor. Secondly, create a new vertex attribute named avgrating to save the calculated mean value of c orresponding vertex.

pop_actor <- data.frame(cbind(Actor = V(h1)[type==FALSE & totrating > 40]\$n
ame,Popularity = V(h1)[type==FALSE &totrating > 40]\$totrating)) # First!

```
y, create a variable named Actor to save the names of vertices which meet t
wo conditions: representing actor name and totrating value being bigger tha
n 40. Secondly, create a variable named Popularity to save the totrating of
 vertices which meet two conditions: representing actor name and totrating
value being bigger than 40. Finally, combine these two variable by columns
to form the data.frame data representing pop_actor.
pop actor[order(pop actor$Popularity,decreasing=TRUE),] # Sort the pop act
or by variable Popularity in a decreasing order.
num <- V(h1)[type==FALSE]$deg # Create a variable named num to save the cor
responding degree of all vertices representing actors.
avgpop <- V(h1)[type==FALSE]$avgrating # Create a variable named avgpop to
save the corresponding avgrating of all vertices representing actors.
totpop <- V(h1)[type==FALSE] $totrating # Create a variable named totpop to
save the corresponding totrating of all vertices
op <- par(mfrow=c(1,2))
plot(num,totpop,main = "scatter plot") # Show a scatter plot with num vs to
plot(num,avgpop,main = "scatter plot") # Show a scatter plot with num vs av
qpop
par(op)
summary(lm(avgpop~num)) # use simple linear regression to analysis dependen
t variable avgpop and independent variable num and make a summary of the re
sult
```

Projection

```
h1.pr <- bipartite.projection(h1) # Project the bipartite graph h1 into two
  one-mode networks and save them into h1.pr
h1.act <- h1.pr$proj1 # Save the first projection of h1 into h1.act
h1.mov <- h1.pr$proj2 # Save the second projection of h1 into h1.mov
h1.act # Show h1.act
h1.mov # Show h1.mov

op <- par(mar = rep(0, 4))
plot(h1.mov,vertex.color="red", # set vertex color as red
  vertex.shape="circle", # set vertex shape as circle
  vertex.size=(V(h1.mov)$IMDBrating)-3, # Set the vertex size by the corre
sponding value of IMBDrating-3
vertex.label=NA, main = "h1.mov") # Plot h1.mov</pre>
```

boxplot(avgpop~num,main="Box Plot") # Show the box plot

```
par(op)
graph.density(h1.mov) # Calculate the density of h1.mov
no.clusters(h1.mov) # Calculate the number of clusters of h1.mov
clusters(h1.mov)$csize # Calculate the size of each cluster in h1.mov
table(E(h1.mov) weight) # Calculate the frequency of edge attribute named w
eight
h2.mov <- induced.subgraph(h1.mov,</pre>
vids=clusters(h1.mov)$membership==1) # Generate a subgraph by all nodes in
the first cluster of h1.mov
plot(h2.mov, vertex.color="red", # set vertex color as red
edge.width=sqrt(E(h1.mov)$weight), # set edge width as the sqrt of the wei
ght edge attribute
vertex.shape="circle", # set vertex shape as circle
vertex.size=(V(h2.mov)$IMDBrating)-3, # Set the vertex size by the corresp
onding value of IMBDrating-3
vertex.label=NA, main = "h2.mov")
# Plot h2.mov
table(graph.coreness(h2.mov)) # Calculate the frequency of coreness of eac
h node in h2.mov
h3.mov <- induced.subgraph(h2.mov,
  vids=graph.coreness(h2.mov)>4)# Generate a subgraph by all nodes whose c
oreness is bigger than 4 in h2.mov
h3.mov #Show h3.mov
plot(h3.mov, vertex.color="red", # set vertex color as red
vertex.shape="circle", # set vertex shape as circle
edge.width=sqrt(E(h1.mov)$weight), # set edge width as the sqrt of the wei
ght edge
vertex.label.cex=0.7, # Set the size of vertex's label as 0.7
vertex.label.color="darkgreen", # Set the color of vertex's label' as dark
green
vertex.label.dist=0.3, #Set the distance of the label from the center of th
e vertex as 0.3
vertex.size=(V(h3.mov)$IMDBrating)-3, # Set the vertex size by the correspo
nding value of IMBDrating-3
main = "h3.mov")
```

Bradley Cooper

V(h1)[nei(V(h1)\$name == "Bradley Cooper")]\$name # Selected all vertices who se neibors' vertex name is equal to "Bradley Cooper" and show all names of thses selected vertices. V(h1)\$shape <- ifelse(V(h1)\$type==TRUE, "square", "circle") # Assign values to the shape attribute of network h1. If the value of type attribute is equ al to TRUE, the corresponding value of shape attribute will be "square". Ot ehrwise, the corresponding value of shape attribute will be "circle". V(h1)\$color <- ifelse(V(h1)\$type==TRUE, "red","lightblue") # Assign values</pre> to the shape attribute of network h1. If the value of type attribute is equ al to TRUE, the corresponding value of shape attribute will be "red". Otehr wise, the corresponding value of shape attribute will be "lightblue". hB <- subgraph.edges(h1, E(h1)[inc(V(h1)[name %in% V(h1)[nei(V(h1)\$name == "Bradley Cooper")]\$name])]) # Creat a subgraph,hB, of h1. The mechanism c ontains two steps. Fistly, obtain a vertex sequence which is the sequence o f vertex.name attribute matched with the given three mobvie names,. Secondl y, sellect all edges that have at least one incident vertex in the obtained vertex sequence. plot(hB, layout = layout_with_kk, main = "Bradley Cooper ", vertex.label.ce x = 0.6) # Plot hB with layout aslayout_with_kk.

"Star Wars: Episode III - Revenge of the Sith

```
H5.mov <-induced.subgraph(h1.mov,vids=V(h1.mov)[nei(V(h1.mov)$name == "Sta
s: Episode III - Revenge of the Sith" ]) #Create a subgraph, H5.mov, from h
1.mov. The mechanism is selecting and using all vertices whose neibors' ver
tex name is equal to "Star Wars: Episode III - Revenge of the Sith"
plot(H5.mov, vertex.color="red", # set vertex color as red
    vertex.shape="circle", # set vertex shape as "circle"
    edge.width=sqrt(E(H5.mov)$weight)*2, # set edge width as the sqrt of
the weight edge
    vertex.label.cex=0.7, # Set the size of vertex's label as 0.7
    vertex.label.color="darkgreen", # Set the color of vertex's label as "
darkgreen"
    vertex.label.dist=0.3, #Set the distance of the label from the center
of the vertex as 0.3
    vertex.size=(V(H5.mov)$IMDBrating), main = "All related movies")# Set
the vertex size by the corresponding value of IMBDrating-3
```