

Affiliation Networks - Chapter 9

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Libraries and data used

Libraries

```
library(UserNetR)
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

```
data(hwd)
```

defining affiliation networks

affiliation as 2-mode networks

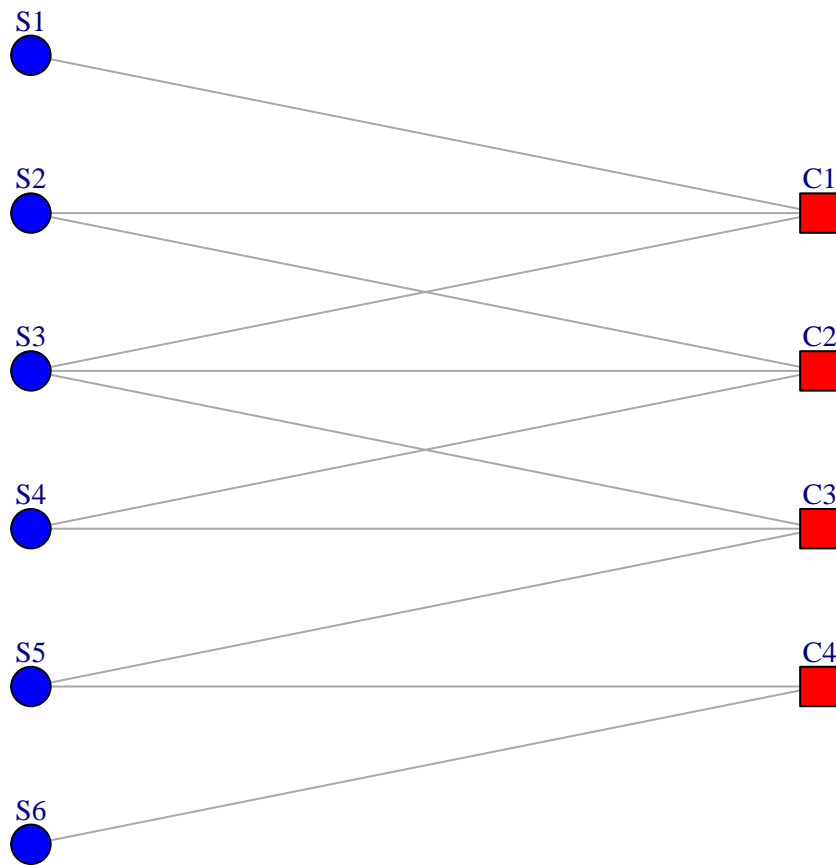
```
C1 <- c(1,1,1,0,0,0)
C2 <- c(0,1,1,1,0,0)
C3 <- c(0,0,1,1,1,0)
C4 <- c(0,0,0,0,1,1)
aff.df <- data.frame(C1,C2,C3,C4)
rownames(aff.df) <- c("S1","S2","S3","S4","S5","S6")
aff.df
```

```
##      C1 C2 C3 C4
## S1   1  0  0  0
## S2   1  1  0  0
## S3   1  1  1  0
## S4   0  1  1  0
## S5   0  0  1  1
## S6   0  0  0  1
```

bipartite graphs

```
bn <- graph.incidence(aff.df)
plt.x <- c(rep(2,6),
           rep(4,4))
plt.y <- c(7:2,
           6:3)
lay <- as.matrix(cbind(plt.x,
                       plt.y))

shapes <- c("circle",
           "square")
colors <- c("blue",
           "red")
plot(bn,
     vertex.color = colors[V(bn)$type + 1],
     vertex.shape = shapes[V(bn)$type + 1],
     vertex.size = 10,
     vertex.label.degree = -pi/2,
     vertex.label.dist = 1.2,
     vertex.label.cex = 0.9,
     layout = lay)
```



Affiliation Network Basics

creating affiliation networks from incidence matrices

```
bn <- graph.incidence(aff.df)
bn

## IGRAPH 40297e6 UN-B 10 11 --
## + attr: type (v/l), name (v/c)
## + edges from 40297e6 (vertex names):
## [1] S1--C1 S2--C1 S2--C2 S3--C1 S3--C2 S3--C3 S4--C2 S4--C3 S5--C3 S5--C4
## [11] S6--C4
```

```
get.incidence(bn)
```

```
##      C1 C2 C3 C4
## S1   1  0  0  0
## S2   1  1  0  0
## S3   1  1  1  0
## S4   0  1  1  0
## S5   0  0  1  1
## S6   0  0  0  1
```

```
V(bn)$type
```

```
## [1] FALSE FALSE FALSE FALSE FALSE FALSE TRUE TRUE TRUE TRUE
```

```
V(bn)$name
```

```
## [1] "S1" "S2" "S3" "S4" "S5" "S6" "C1" "C2" "C3" "C4"
```

```
##Creating Affiliation network from edge list
```

```
el.df <- data.frame(rbind(c("S1", "C1"),
                           c("S2", "C1"),
                           c("S2", "C2"),
                           c("S3", "C1"),
                           c("S3", "C2"),
                           c("S3", "C3"),
                           c("S4", "C2"),
                           c("S4", "C3"),
                           c("S5", "C3"),
                           c("S5", "C4"),
                           c("S6", "C4")))
```

```
el.df
```

```
##      X1 X2
## 1   S1 C1
## 2   S2 C1
## 3   S2 C2
## 4   S3 C1
## 5   S3 C2
## 6   S3 C3
## 7   S4 C2
## 8   S4 C3
## 9   S5 C3
## 10  S5 C4
## 11  S6 C4
```

```
bn2 <- graph.data.frame(el.df,
                        directed = FALSE)
```

```
bn2
```

```
## IGRAPH 40347ae UN-- 10 11 --
## + attr: name (v/c)
## + edges from 40347ae (vertex names):
## [1] S1--C1 S2--C1 S2--C2 S3--C1 S3--C2 S3--C3 S4--C2 S4--C3 S5--C3 S5--C4
## [11] S6--C4
```

```
V(bn2)$type <- V(bn2)$name %in% el.df[,1]
```

```
bn2
```

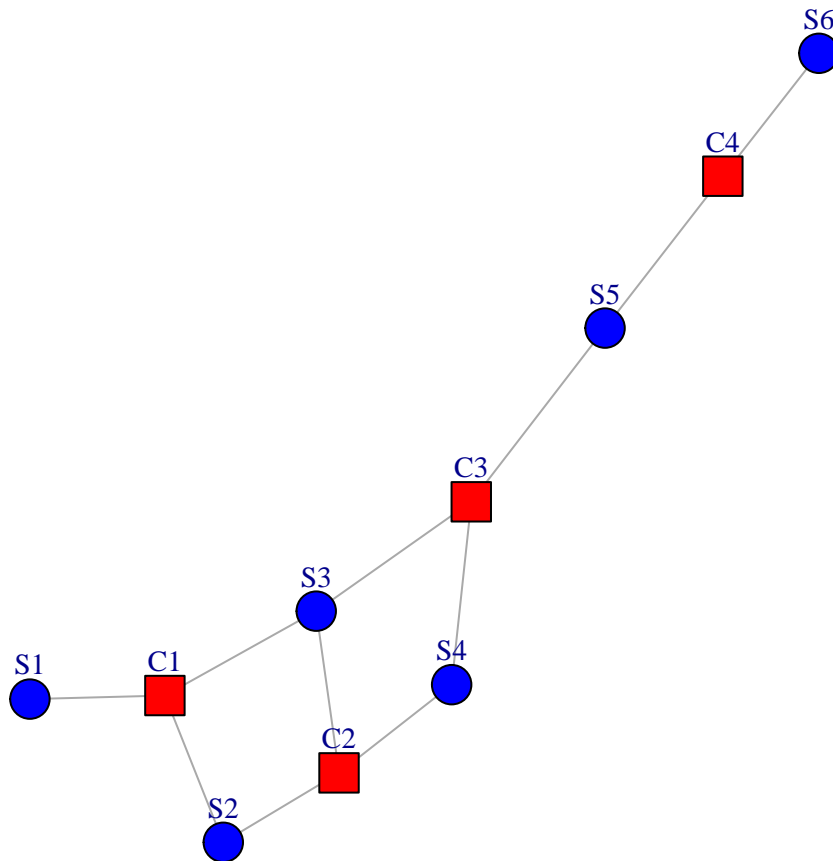
```
## IGRAPH 40347ae UN-B 10 11 --
## + attr: name (v/c), type (v/l)
## + edges from 40347ae (vertex names):
## [1] S1--C1 S2--C1 S2--C2 S3--C1 S3--C2 S3--C3 S4--C2 S4--C3 S5--C3 S5--C4
## [11] S6--C4
```

```
graph.density(bn) == graph.density(bn2)
```

```
## [1] TRUE
```

plotting Affiliation networks

```
shapes <- c("circle",
            "square")
colors <- c("blue",
            "red")
plot(bn,
      vertex.color = colors[V(bn)$type + 1],
      vertex.shape = shapes[V(bn)$type + 1],
      vertex.size = 10,
      vertex.label.degree = -pi/2,
      vertex.label.dist = 1.2,
      vertex.label.cex = 0.9)
```



```

bn.pr <- bipartite.projection(bn)
bn.pr

## $proj1
## IGRAPH 403e8d0 UNW- 6 8 --
## + attr: name (v/c), weight (e/n)
## + edges from 403e8d0 (vertex names):
## [1] S1--S2 S1--S3 S2--S3 S2--S4 S3--S4 S3--S5 S4--S5 S5--S6
##
## $proj2
## IGRAPH 403e8d0 UNW- 4 4 --
## + attr: name (v/c), weight (e/n)
## + edges from 403e8d0 (vertex names):
## [1] C1--C2 C1--C3 C2--C3 C3--C4

```

```
graph.density(bn.pr$proj1)
```

```
## [1] 0.5333333
```

```
bn.students <- bn.pr$proj1
```

```
bn.class <- bn.pr$proj2
```

```
graph.density(bn.students)
```

```
## [1] 0.5333333
```

```
get.adjacency(bn.students,  
              sparse = FALSE,  
              attr = "weight")
```

```
##      S1 S2 S3 S4 S5 S6  
## S1  0  1  1  0  0  0  
## S2  1  0  2  1  0  0  
## S3  1  2  0  2  1  0  
## S4  0  1  2  0  1  0  
## S5  0  0  1  1  0  1  
## S6  0  0  0  0  1  0
```

```
get.adjacency(bn.class,  
              sparse = FALSE,  
              attr = "weight")
```

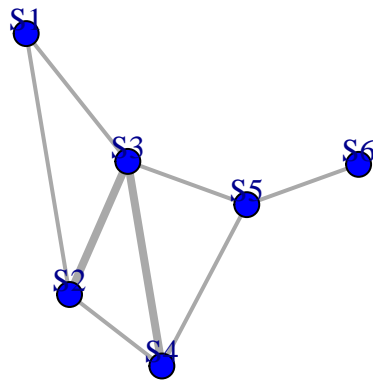
```
##      C1 C2 C3 C4  
## C1  0  2  1  0  
## C2  2  0  2  0  
## C3  1  2  0  1  
## C4  0  0  1  0
```

```
op <- par(mfrow = c(1,2))
```

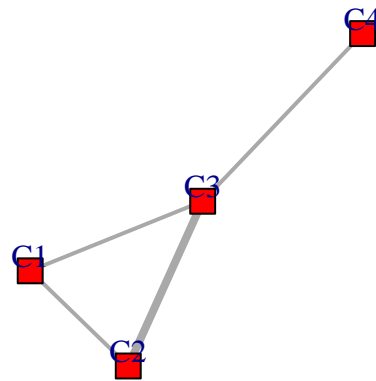
```
plot(bn.students,  
     vertex.color = "blue",  
     vertex.shape = "circle",  
     main = "Students",  
     edge.width = E(bn.students)$weight * 2,  
     vertex.size = 15,  
     vertex.label.degree = -pi/2,  
     vertex.label.dist = 1.2,  
     vertex.label.cex = 1)
```

```
plot(bn.class,  
     vertex.color = "red",  
     vertex.shape = "square",  
     main = "Class",  
     edge.width = E(bn.students)$weight * 2,  
     vertex.size = 15,  
     vertex.label.degree = -pi/2,  
     vertex.label.dist = 1.2,  
     vertex.label.cex = 1)
```

Students



Class



```
par(op)
```

examples

Hollywood Actors as an affiliation network

```
data(hwd)
h1 <- hwd
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
V(h1)$name[1:10]

## [1] "Inception"
## [2] "Alice in Wonderland"
## [3] "Kick-Ass"
## [4] "Toy Story 3"
## [5] "How to Train Your Dragon"
## [6] "Despicable Me"
## [7] "Scott Pilgrim vs. the World"
## [8] "Hot Tub Time Machine"
## [9] "Harry Potter and the Deathly Hallows: Part 1"
## [10] "Tangled"
V(h1)$type[1:10]

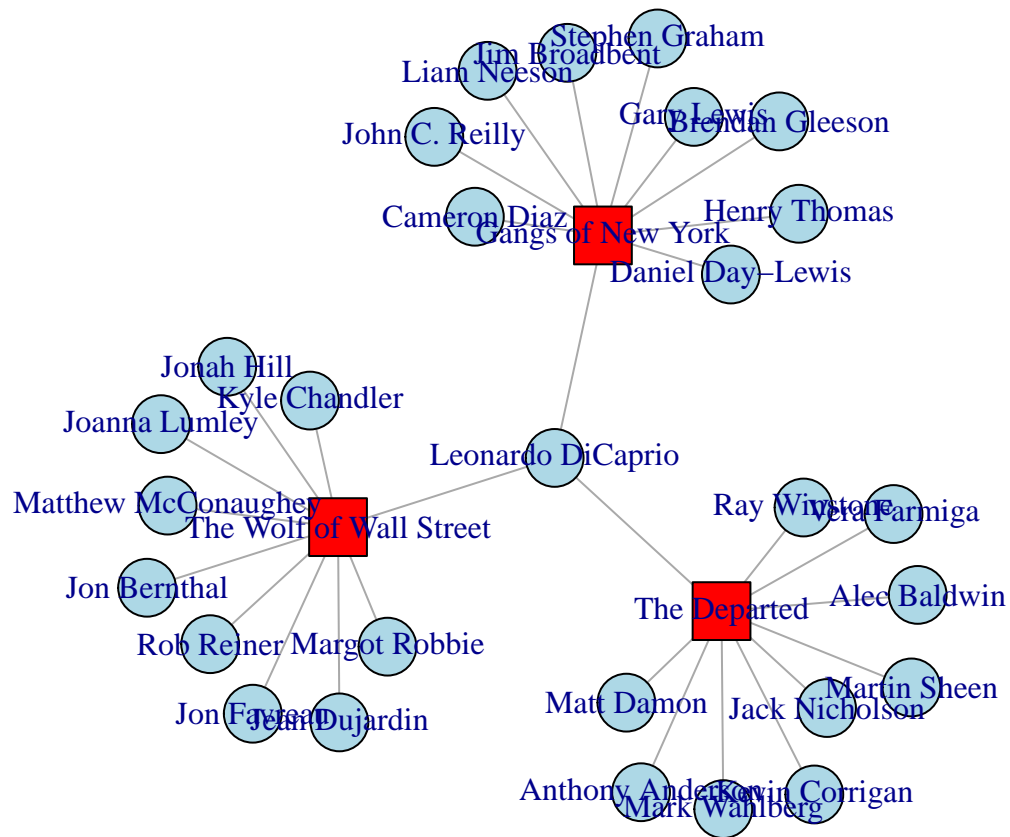
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
V(h1)$IMDBrating[1:10]

## [1] 8.8 6.5 7.8 8.4 8.2 7.7 7.5 6.5 7.7 7.9
V(h1)$name[155:165]

## [1] "Notting Hill"          "Eyes Wide Shut"
## [3] "The Green Mile"        "10 Things I Hate About You"
## [5] "American Pie"          "Girl, Interrupted"
## [7] "Leonardo DiCaprio"    "Joseph Gordon-Levitt"
## [9] "Ellen Page"           "Tom Hardy"
## [11] "Ken Watanabe"
V(h1)$shape <- ifelse(V(h1)$type == TRUE,
                      "square",
                      "circle")
V(h1)$shape[1:10]

## [1] "square" "square" "square" "square" "square" "square" "square" "square"
## [9] "square" "square"
V(h1)$color <- ifelse(V(h1)$type == TRUE,
                      "red",
                      "lightblue")
h2 <- subgraph.edges(h1,
                     E(h1)[inc(V(h1)[name %in%
                                c("The Wolf of Wall Street",
                                  "Gangs of New York",
                                  "The Departed")])])
plot(h2,
     layout = layout_with_kk)

```



```
graph.density(h1)
```

```
## [1] 0.001718711
```

```
table(degree(h1, v=V(h1)[type==FALSE]))
```

```
##
```

```
## 1 2 3 4 5 6 7 8
```

```
## 955 165 47 23 11 2 1 1
```

```
mean(degree(h1, v=V(h1)[type==FALSE]))
```

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

```
V(h1)$deg <- degree(h1)
```

```
V(h1)[type == FALSE & deg > 4]$name
```

```
## [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"
```

```
busy_actor <- data.frame(cbind(
  Actor = V(h1)[type == FALSE & deg > 4]$name,
  Movies = V(h1)[type == FALSE & deg > 4]$deg
))
busy_actor[order(busy_actor$Movies,
  decreasing = TRUE), ]
```

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

```
for(i in 161:1365){
  V(h1)[i]$totrating <- sum(V(h1)[nei(i)]$IMDBrating)
}
max(V(h1)$totrating, na.rm = TRUE)
```

```
## [1] 60.9
```

```
pop_actor <- data.frame(cbind(
  Actor = V(h1)[type == FALSE &
    totrating > 40]$name,
  Popularity = V(h1)[type == FALSE &
    totrating > 40]$totrating))
pop_actor[order(pop_actor$Popularity,
  decreasing = TRUE), ]
```

```
##           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
```

```
for(i in 161:1365){
  V(h1)[i]$avgrating <- mean(V(h1)[nei(i)]$IMDBrating)
}
num <- V(h1)[type == FALSE]$deg
```

```

avgpop <- V(h1)[type == FALSE]$avgrating
summary(lm(avgpop ~ num))

##
## Call:
## lm(formula = avgpop ~ num)
##
## Residuals:
##      Min       1Q   Median       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 ***
## num          0.04714    0.03527   1.337   0.182
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 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

scatter.smooth(num,
               avgpop,
               col = "lightblue",
               ylim = c(2, 10),
               span = 0.8,
               xlab = "Number of Movies",
               ylab = "Avg. Popularity")

## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## pseudoinverse used at 0.965
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## neighborhood radius 1.035
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## reciprocal condition number 5.4652e-015
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## There are other near singularities as well. 1
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## pseudoinverse used at 0.965
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## neighborhood radius 1.035
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## reciprocal condition number 1.9145e-015
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## There are other near singularities as well. 1
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## pseudoinverse used at 0.965
##
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :
## neighborhood radius 1.035

```

```
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## reciprocal condition number 1.2706e-015  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## There are other near singularities as well. 1  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## pseudoinverse used at 0.965  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## neighborhood radius 1.035  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## reciprocal condition number 2.3999e-015  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## There are other near singularities as well. 1  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## pseudoinverse used at 0.965  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## neighborhood radius 1.035  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## reciprocal condition number 5.4652e-015  
  
## Warning in simpleLoess(y, x, w, span, degree = degree, parametric = FALSE, :  
## There are other near singularities as well. 1
```

