# **Grey's Anatomy**

## HilaReut

## 16 May 2016

library(igraph)

```
##
 ## Attaching package: 'igraph'
 ## The following objects are masked from 'package:stats':
 ##
 ##
        decompose, spectrum
 ## The following object is masked from 'package:base':
 ##
 ##
        union
 require(igraph)
Grey's anatomy data
 edges.data = read.csv('ga edgelist.csv',header = T)
 ga edges = graph.data.frame(edges.data,directed = F)
 summary(ga edges)
 ## IGRAPH UN-- 32 34 --
 ## + attr: name (v/c)
 V(ga edges)$name
                         "owen"
                                         "sloan"
     [1] "lexi"
                                                         "torres"
 ##
                                         "o'malley"
     [5] "derek"
                         "karev"
                                                         "yang"
 ##
     [9] "grey"
                         "chief"
                                         "ellis grey"
                                                         "susan grey"
 ## [13] "bailey"
                         "izzie"
                                         "altman"
                                                         "arizona"
 ## [17] "colin"
                         "preston"
                                         "kepner"
                                                         "addison"
 ## [21] "nancy"
                                         "mrs. seabury" "adele"
                         "olivia"
                                         "hank"
 ## [25] "thatch grey"
                         "tucker"
                                                         "denny"
                                         "ben"
 ## [29] "finn"
                         "steve"
                                                         "avery"
 #Remove self-Loops is exist
```

ga\_edges = simplify(ga\_edges)

```
#Calculate betweenness
ga_bet = betweenness(ga_edges)
ga_bet = sort(ga_bet, decreasing = T)
names(ga_bet[1])
```

```
## [1] "sloan"
```

#### Calculate closeness

```
#Calculate closeness
ga_close = closeness(ga_edges)
ga_close = sort(ga_close, decreasing = T)
names(ga_close[1])
```

```
## [1] "torres"
```

#### Calculate eigenvector

```
#Calculate eigenvector
ga_eigen = evcent(ga_edges)
ga_eigen = sort(ga_eigen$vector, decreasing = T)
names(ga_eigen[1])
```

```
## [1] "karev"
```

### Find community with Girvan-Newman community detection.

```
fc = edge.betweenness.community(ga_edges)
```

#### Cheack what is the modularity

```
#Cheack what is the modularity fc$modularity
```

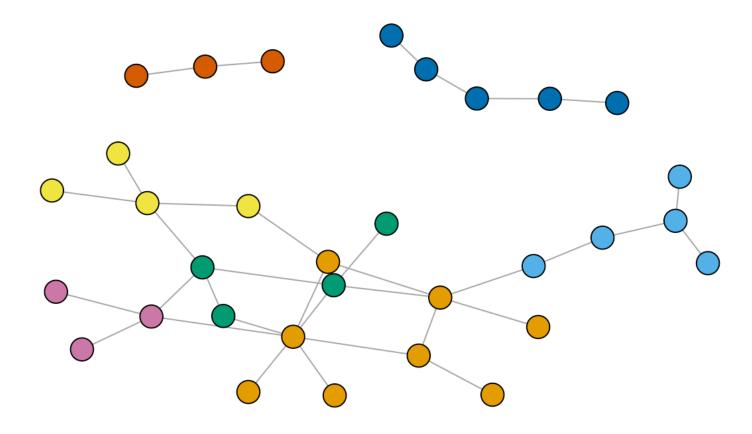
```
[1] -0.04584775 -0.01816609
                              0.01038062 0.03892734 0.06747405
##
##
   [6] 0.09515571 0.12326990
                              0.14922145 0.17560554 0.20328720
## [11] 0.23053633 0.25821799
                               0.28633218
                                         0.31358131
                                                     0.34169550
## [16] 0.36851211 0.39576125
                               0.42344291 0.44247405 0.46712803
## [21] 0.49134948 0.50778547
                               0.52681661
                                          0.54974048
                                                     0.57050173
## [26] 0.57742215 0.56098616
                              0.53416955
                                         0.45804498 0.30449827
```

```
#What partition is the best?
max(fc$modularity)
```

```
## [1] 0.5774221
```

```
which.max(fc$modularity)
```

```
## [1] 26
```



```
#How many communities received max(levels(as.factor(memb)))
```

```
## [1] "7"
```

```
#What size of each commuinty
summary(as.factor(memb))
```

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

This function implements the multi-level modularity optimization algorithm for finding community structure.

```
ml = multilevel.community(ga_edges)

#Cheack what is the modularity
ml$modularity
```

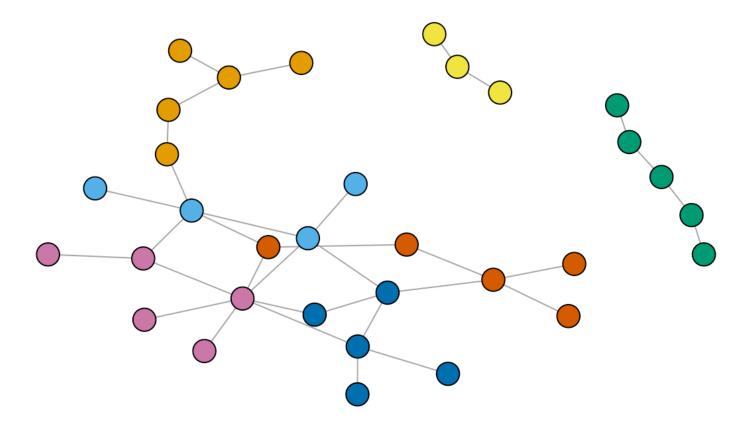
```
## [1] 0.4762111 0.5804498
```

```
#What partition is the best?
max(ml$modularity)
```

```
## [1] 0.5804498
```

```
which.max(ml$modularity)
```

```
## [1] 2
```



```
#How many communities received
max(levels(as.factor(memb)))
```

```
## [1] "7"
```

```
#What size of each commuinty
summary(as.factor(memb))
```

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

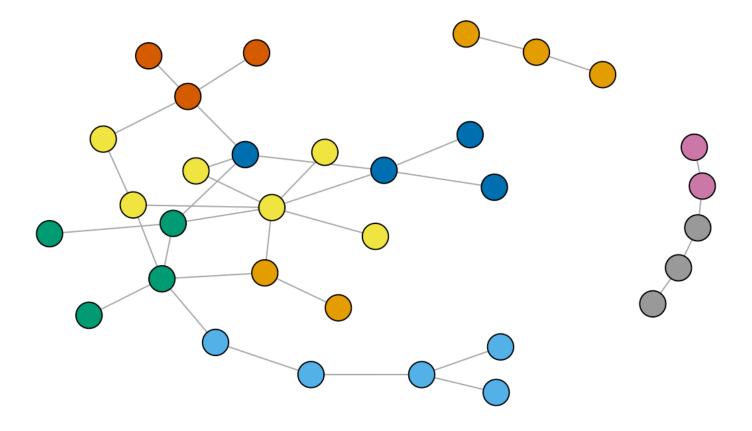
# Find commuinty with propagating labels algorithm

This is a fast, nearly linear time algorithm for detecting community structure in networks. In works by labeling the vertices with unique labels and then updating the labels by majority voting in the neighborhood of the vertex.

```
networks.
#In works by labeling the vertices with unique labels and then updating the labels
by majority voting in the neighborhood of the vertex.
pl = label.propagation.community(ga edges)
#Cheack what is the modularity
pl$modularity
## [1] 0.533737
#What partition is the best?
max(pl$modularity)
## [1] 0.533737
which.max(pl$modularity)
## [1] 1
#Color nodes by partitions
memb = membership(pl)
plot(ga edges, vertex.size=8, vertex.label=NA,
     vertex.color=memb, asp=FALSE)
```

#This is a fast, nearly linear time algorithm for detecting community structure in

#Find commuinty with propagating labels algorithm



#How many communities received
max(levels(as.factor(memb)))

## [1] "9"

#What size of each commuinty
summary(as.factor(memb))

## 1 2 3 4 5 6 7 8 9 ## 2 5 4 6 4 3 2 3 3