

HW9 Solution CS6220-Data Mining

```
library(igraph)
```

```
##  
## Attaching package: 'igraph'  
##  
## The following objects are masked from 'package:stats':  
##  
##   decompose, spectrum  
##  
## The following object is masked from 'package:base':  
##  
##   union
```

Problem 1

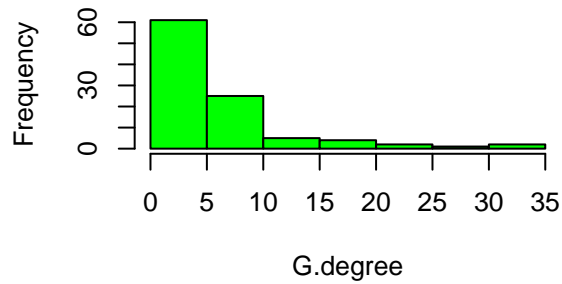
(a)

```
#Load the data  
setwd('/Users/ovitek/Dropbox/Olga/Teaching/CS6220/Fall15/Homeworks/Hw9')  
mytable <- read.table("yeast_broad.tsv", header = FALSE, stringsAsFactors = FALSE)  
load("yeast_names.Rdata")  
ls()
```

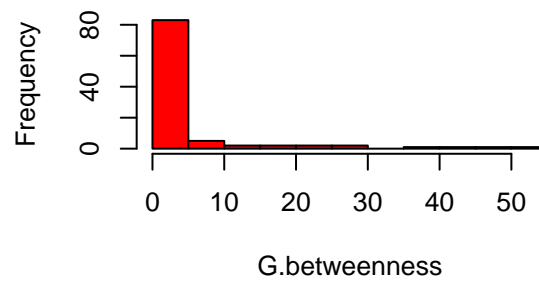
```
## [1] "mytable"      "vertex.table"
```

```
G <- graph.data.frame(mytable, vertices = vertex.table, directed = TRUE)  
  
G.degree = degree(G)  
G.betweenness = betweenness(G)  
G.pagerank = page.rank(G)$vector  
  
par(mfrow = c(2,2))  
hist(G.degree, breaks = 10, col = "green" )  
hist(G.betweenness, col = "red" )  
hist(G.pagerank, col = "blue" )
```

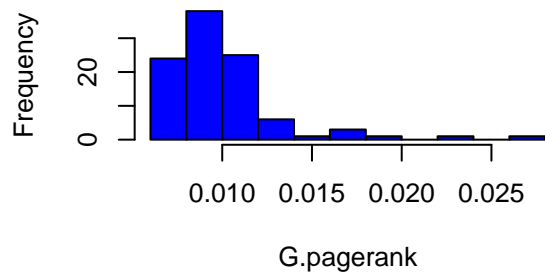
Histogram of G.degree



Histogram of G.betweenness



Histogram of G.pagerank



(b)

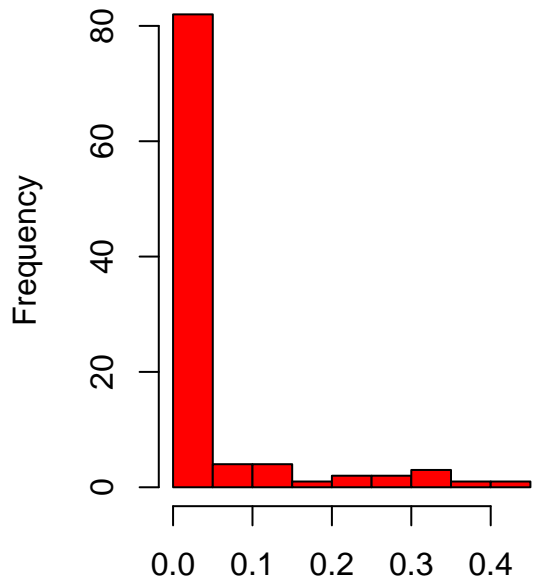
```
# Implementation of the HITS algorithm
#-----
hitsFunc = function(A, times) {
  h = rep(1, dim(A)[1])
  a = rep(1, dim(A)[1])

  for (i in 1:times) {
    a <- t(A) %*% h
    a <- a / sqrt(sum(a ^ 2))
    h <- A %*% a
    h <- h / sqrt(sum(h ^ 2))
  }
  return(list(hub = h, auth = a))
}

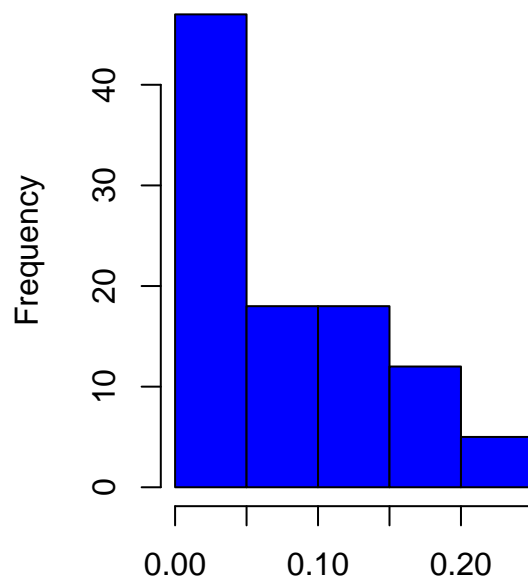
myMatrix <- as.matrix(get.adjacency(G))
hits = hitsFunc(myMatrix, 20)
G.hub = hits$hub
G.auth = hits$auth

par(mfrow = c(1,2))
hist(G.hub, col = "red")
hist(G.auth, col = "blue")
```

Histogram of G.hub



Histogram of G.auth

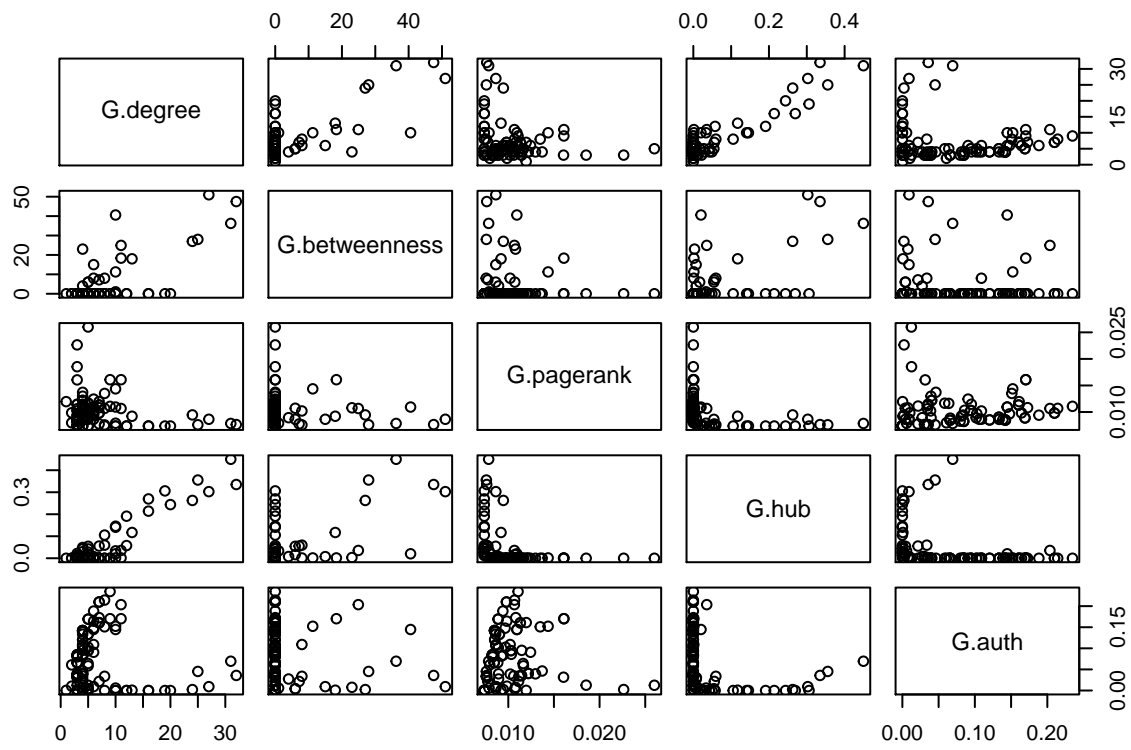


G.hub

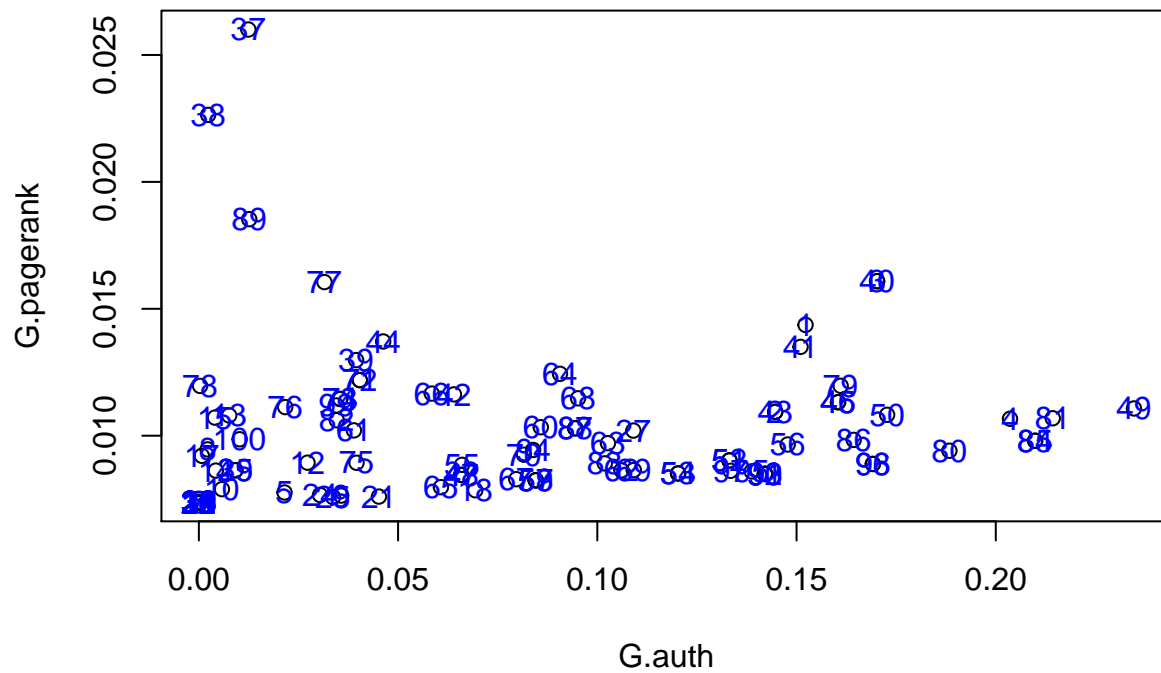
G.auth

(c)

```
pairs( data.frame(G.degree, G.betweenness, G.pagerank, G.hub, G.auth) )
```



```
plot (G.auth, G.pagerank)
text(G.auth, G.pagerank, col = "blue", cex=1)
```



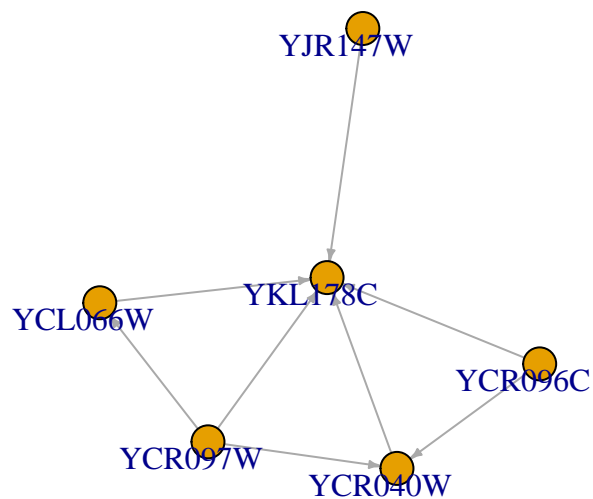
Node 37 and 38 seems to have low authority but high page rank.

```
n.name <- V(G)[37]$name
n.name
```

```
## [1] "YKL178C"
```

```
names = V(G)[nei(n.name)]$name
sub = c(names, n.name)
subG = induced.subgraph(G, V(G)[sub])
```

```
plot.igraph(subG, layout = layout.auto(subG), vertex.label.dist = -.5, edge.arrow.size = .3)
```



Problem 2

(a)

```
p = 1/20
numnode = seq(from = 2000, to = 3000, by = 100)
len = length(numnode)
prtime = rep(0, len)
hitstime = rep(0, len)

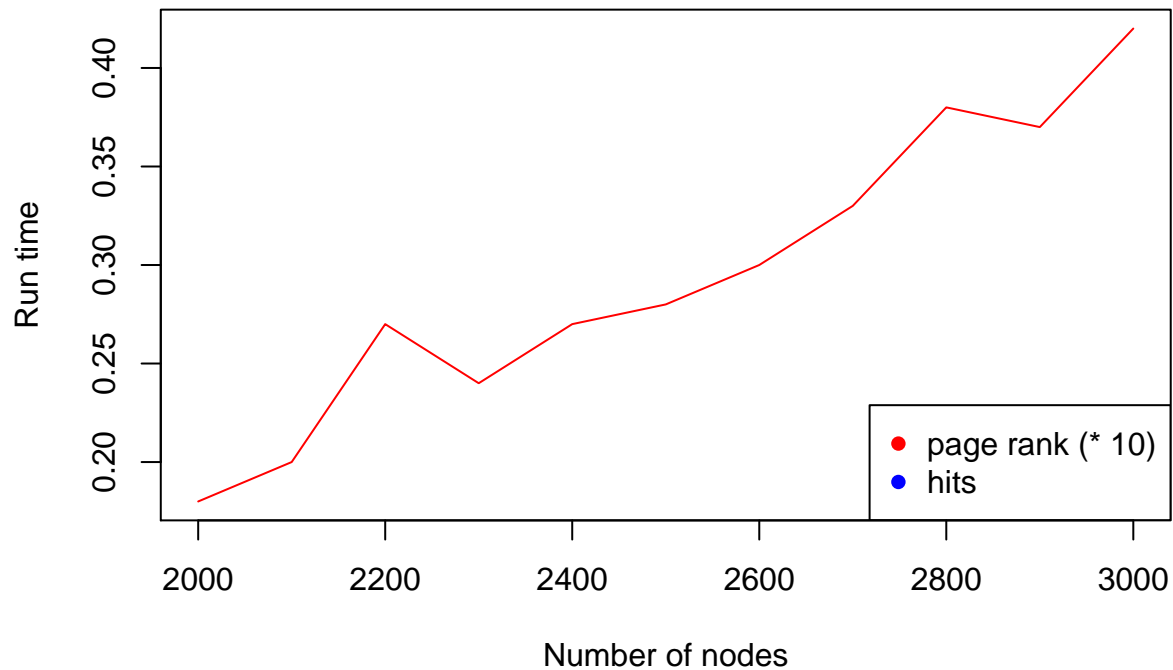
for (i in 1:len) {
  g <- sample_gnp(numnode[i], 1/20, directed=TRUE)
  prtime[i] = system.time(page.rank(g))
  mat = as.matrix(get.adjacency(g))
  hitstime[i] = system.time(hitsFunc(mat, 20))
}
#
prtime

## [1] 0.018 0.020 0.027 0.024 0.027 0.028 0.030 0.033 0.038 0.037 0.042

hitstime

## [1] 1.397 1.236 1.402 1.644 1.641 2.101 2.273 2.479 2.528 3.034 3.399

plot (numnode, 10 * prtime, col = "red", type = 'l', ylab = "Run time", xlab = "Number of nodes")
lines (numnode, hitstime, col = "blue", type = 'l')
legend("bottomright", pch=16, col=c("red", "blue"), c("page rank (* 10)", "hits"))
```



(b)

```

for (i in 1:len) {
  g <- sample_gnp(numnode[i], 1/3, directed=TRUE)
  prtime[i] = system.time(page.rank(g))
  mat = as.matrix(get.adjacency(g))
  hitstime[i] = system.time(hitsFunc(mat, 20))
}
#
prtime

```

```
## [1] 0.122 0.135 0.158 0.194 0.185 0.217 0.228 0.246 0.299 0.301 0.350
```

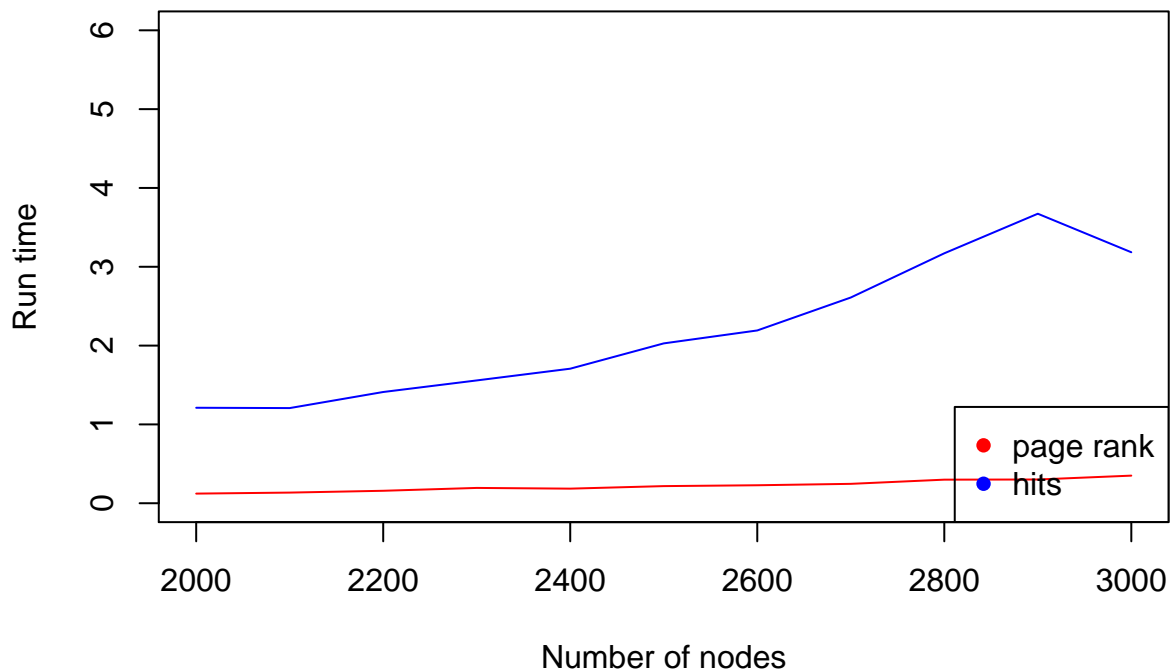
```
hitstime
```

```
## [1] 1.211 1.207 1.411 1.558 1.707 2.028 2.192 2.610 3.171 3.673 3.184
```

```

plot (numnode, prtime, col = "red", type = 'l', ylim = c(0,6), ylab = "Run time", xlab = "Number of nodes")
lines (numnode, hitstime, col = "blue", type = 'l')
legend("bottomright", pch=16, col=c("red", "blue"), c("page rank", "hits"))

```



(c)

Increasing the edge probability means increasing the number of edges. The run time of the pagerank algorithm increases as either the number of nodes or edges increase. In the other hand, while the run time of the HITS algorithm increases with the number of nodes, it doesn't change when the number of edges increases.

The pattern is due to our inefficient implementation of the HITS algorithm. Matrix multiplications are very expensive and do not scale well.