

Applying Community Detection

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```
library(igraph)
library(RColorBrewer)
library(scales)

#my_color_pal <- c("#dc661b", "#c9da0b", "#2eb83c", "#157184", "#7a8fe1", "#5525a2", "#fd74d8", "#b948d5")

#add.alpha <- function(cols, alpha) rgb(t(col2rgb(cols)/255), alpha = alpha)
#colours<-add.alpha(my_color_pal, 0.9)
colours <- brewer.pal(n=12,name = "Paired")

show_col(colours)
```

#A6CEE3	#1F78B4	#B2DF8A	#33A02C
#FB9A99	#E31A1C	#FDBF6F	#FF7F00
#CAB2D6	#6A3D9A	#FFFF99	#B15928

```
#code block for single matrix
```

```
#check read.table doc skip param
```

```
data <- read.table("similarity_matrix_for_graph.csv",sep = ',',header = TRUE, check.names = FALSE, skip
```

```
#DF to matrix
```

```
data_matrix <- data.matrix(data)
```

```
#colnames of the matrix
```

```
colnames(data_matrix)
```

```
## [1] "87153398" "87153401" "91158359" "91253500" "87212280" "89045465"  
## [7] "87153394" "87153396" "87153393" "91303201" "87153395" "89353055"  
## [13] "87153399" "91349874" "87153397" "91042911" "91158353" "91303191"  
## [19] "91157419" "90079999" "89034877" "90054688" "88019051" "87272402"  
## [25] "87214971" "90294859" "87269956" "90045364" "88273558" "87181747"  
## [31] "88048748" "91194654" "89300773" "89090775" "88070619" "89210525"  
## [37] "87154824" "88105976" "91157909" "87170682" "88093278" "87212275"  
## [43] "89090802" "88132177" "89057110" "87266840" "91048168" "87127179"  
## [49] "89205573" "91029107"
```

```
#making all lower triangle values zero in the matrix including diag
```

```
data_matrix[lower.tri(data_matrix,diag = TRUE)] <- 0
```

```
#number of zeros in the matrix
```

```
#length(which(data_matrix == 0))
```

```
#avg of non-zero values in the matrix
```

```
sum(data_matrix[data_matrix != 0])/length(data_matrix[data_matrix != 0])
```

```
## [1] 0.1903989
```

```
summary(data_matrix[data_matrix != 0])
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
## 0.05502 0.10655 0.18149 0.19040 0.25077 0.62199
```

```
fivenum(data_matrix[data_matrix != 0])
```

```
## [1] 0.05501601 0.10654775 0.18148821 0.25076880 0.62198535
```

```
#3rd quadrant
```

```
threshold <- fivenum(data_matrix[data_matrix != 0])[4]
```

```
#number of values less than threshold (including zeros and lower tria)
```

```
length(which(data_matrix >= threshold))
```

```
## [1] 203
```

```

#filter out values below threshold
data_matrix[data_matrix < threshold] <- 0
#data_matrix[data_matrix >= threshold] <- 1
#length(which(data_matrix == 0))
#weighted check
g <- graph_from_adjacency_matrix(data_matrix, weighted=TRUE, mode="upper", diag = FALSE, add.colnames =
#V(g)
#V(g)$medline_ui
#E(g)
#E(g)$weight

summary(g)

```

```

## IGRAPH 6c446ec U-W- 50 203 --
## + attr: medline_ui (v/c), weight (e/n)

```

```

#summary(E(g)$weight)

#fivenum(E(g)$weight)[2]

#plot(g)

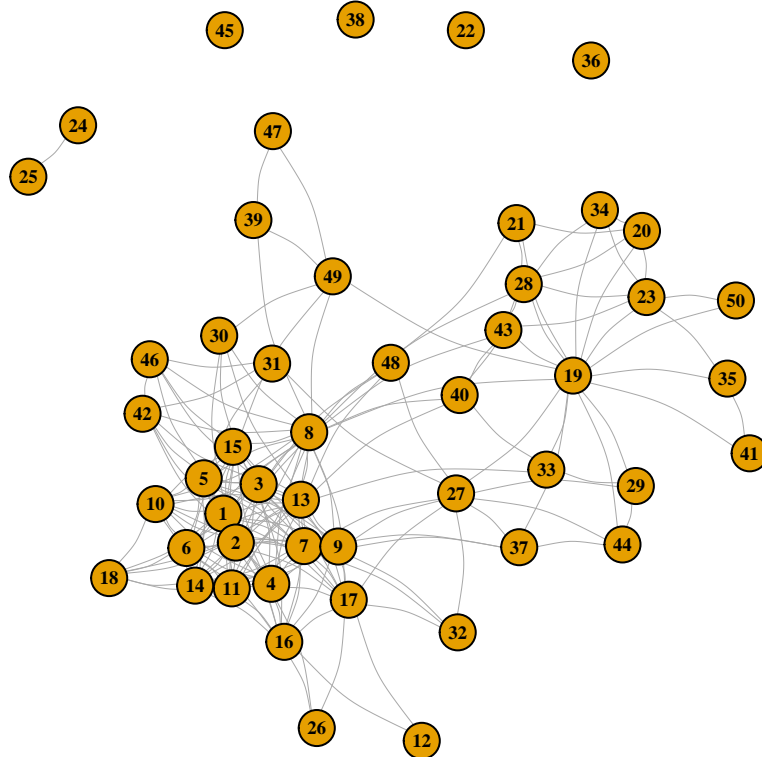
```

```

vertex_size <- 10
cex_size <-0.6
par(mar=c(0,0,1,0)+.1)
plot( g,
layout=layout_nicely,
vertex.label.cex=cex_size,
vertex.label.color="black",
#vertex.label.dist = 0.7,
vertex.label.font = 2,
vertex.size = vertex_size,
#edge.color = "gray90",
edge.width=0.5,
edge.curved=0.2,
main = paste("Sample graph")
)

```

Sample graph



```
#By default the 'weight' edge attribute is used as weights.Larger edge weights correspond to stronger c
fg.community<- cluster_fast_greedy(g)
table(fg.community$membership)
```

```
##
##  1  2  3  4  5  6  7  8
## 10 17 17  2  1  1  1  1
```

```
#modularity of actual split
modularity(fg.community)
```

```
## [1] 0.3036332
```

```
#modularity of actual split
modularity(g,membership(fg.community),E(g)$weight)
```

```
## [1] 0.3036332
```

```
#plot(as.dendrogram(as.hclust((fg.community))))

#is_hierarchical(fg.community)
fg.10_comm <- cut_at(fg.community,no = 10)
table(fg.10_comm)
```

```
## fg.10_comm
##  1  2  3  4  5  6  7  8  9 10
##  2 17  2  6 17  2  1  1  1  1
```

```
#modularity of cut at 10
modularity(g,fg.10_comm,E(g)$weight)
```

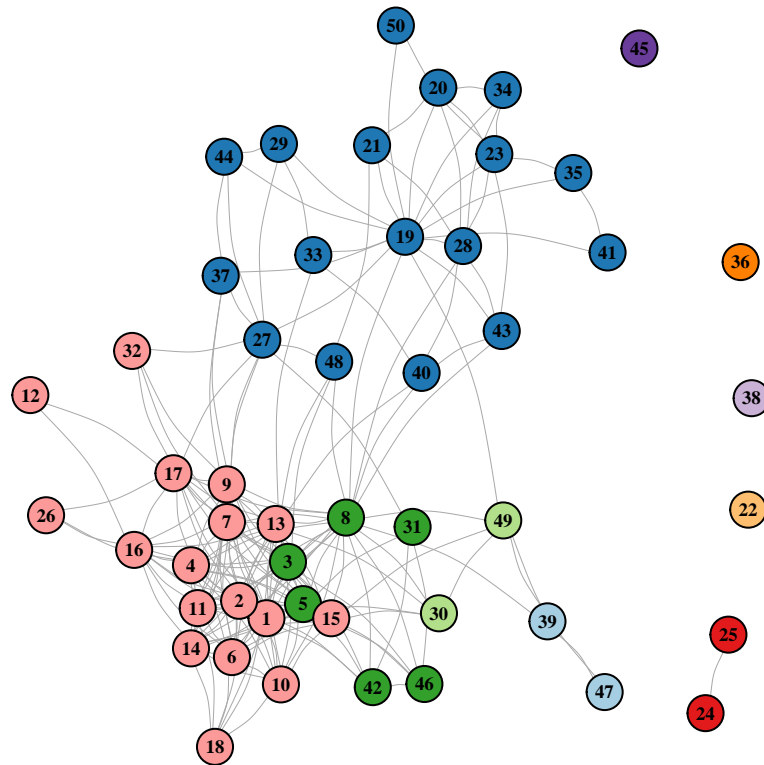
```
## [1] 0.2947539
```

```
#length(fg.10_comm)
```

```
#weights: Optional positive weight vector. If the graph has a weight edge attribute, then this is used
#louvain.community<- cluster_louvain(g)
#length(louvain.community)
```

```
par(mar=c(0,0,1,0)+.1)
plot( g,
      layout=layout_nicely,
      vertex.color=colours[fg.10_comm],
      vertex.label.cex=cex_size,
      vertex.label.color="black",
      #vertex.label.dist = 0.7,
      vertex.label.font = 2,
      vertex.size = vertex_size,
      #edge.color = "gray90",
      edge.width=0.5,
      edge.curved=0.2,
      main = paste("Cut at 10")
)
```

Cut at 10

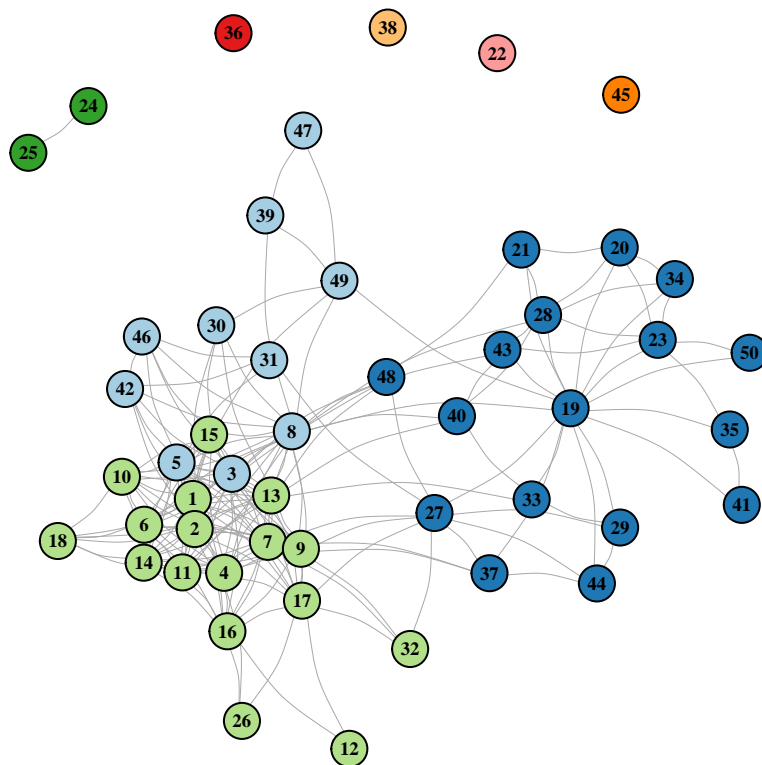


```

par(mar=c(0,0,1,0)+.1)
plot( g,
      layout=layout_nicely,
      vertex.color=colours[fg.community$membership],
      vertex.label.cex=cex_size,
      vertex.label.color="black",
      #vertex.label.dist = 0.7,
      vertex.label.font = 2,
      vertex.size = vertex_size,
      #edge.color = "gray90",
      edge.width=0.5,
      edge.curved=0.2,
      main = paste("Actual communities")
)

```

Actual communities



```
#code for all 63 queries
# to find clusters and new top 10 docs

actual_clusters = c()
final_clusters = c()

orig_top_50 <- list()
new_top_10 <- list()

for (i in 1:63) {

data <- read.table("similarity_matrix_for_graph.csv",sep = ',',header = TRUE, check.names = FALSE, skip

#DF to matrix
data_matrix <- data.matrix(data)

#colnames of the matrix
#colnames(data_matrix)

#making all lower triangle values zero in the matrix including diag
data_matrix[lower.tri(data_matrix,diag = TRUE)] <- 0

#3rd quadrant
threshold <- fivenum(data_matrix[data_matrix != 0])[4]
```

```

#filter out values below threshold
data_matrix[data_matrix < threshold] <- 0

#weighted check
g <- graph_from_adjacency_matrix(data_matrix, weighted=TRUE, mode="upper", diag = FALSE, add.colnames=

#summary(g)

fg.community <- cluster_fast_greedy(g)
actual_clusters[i] <- length(fg.community)
#sizes(fg.community)

if(length(fg.community) >= 10)
{
  final_clusters[i] <- length(fg.community)
  print(sizes(fg.community))
  print(fg.community$membership)
  mem_vector <- fg.community$membership
  n_comm <- length(fg.community)
}
else
{
  fg.10_comm <- cut_at(fg.community,no = 10)
  final_clusters[i] <- 10
  print(table(fg.10_comm))
  print(fg.10_comm)
  mem_vector <- fg.10_comm
  n_comm <- 10
}

#initialisations
index <- c(1)
seen_clusters <- c(mem_vector[1])

for (n in 2:50)
{
  if (!(mem_vector[n] %in% seen_clusters))
  {
    seen_clusters <- append(seen_clusters,mem_vector[n])
    index <- append(index,n)
  }
}

#taking only top 10
new_top_10 <- append(new_top_10,list(index[1:10]))
orig_top_50 <- append(orig_top_50,list(V(g)$medline_ui))
}

```



```

## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 16 26 1 1 1 1 1 1 1 1
## [1] 2 2 2 2 1 1 2 2 1 1 1 1 2 1 1 1 2 3 2 2 1 2 4 2
## [26] 2 2 2 5 2 6 2 1 1 1 2 2 2 1 7 8 1 2 2 2 2 2 9 2 10
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 5 10 13 6 11 1 1 1 1 1
## [1] 3 5 4 6 2 2 5 7 8 2 9 5 5 4 5 2 2 1 2 2 5 3 4 3 1
## [26] 3 3 4 3 5 4 3 4 3 1 3 2 3 3 5 2 5 3 1 1 5 10 2 3 5
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 17 2 10 15 1 1 1 1 1 1
## [1] 1 3 3 1 5 3 3 6 1 4 7 1 1 3 1 1 4 4 1 4 8 3 1 4 4
## [26] 4 4 1 3 9 4 1 3 1 1 1 2 1 4 4 1 3 4 3 1 2 4 4 10 4
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 2 17 2 6 17 2 1 1 1 1
## [1] 5 5 4 5 4 5 5 4 5 5 5 5 5 5 5 5 5 2 2 2 7 2 6 6
## [26] 5 2 2 2 3 4 5 2 2 2 8 2 9 1 2 2 4 2 2 10 4 1 2 3 2
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 19 10 2 13 1 1 1 1 1 1
## [1] 2 1 1 2 1 4 4 2 2 4 4 4 4 4 2 1 2 1 4 1 5 1 1 1 2
## [26] 1 1 1 2 1 4 4 1 1 3 4 1 1 6 4 7 8 1 9 1 10 2 3 4 2
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 12 10 12 9 2 1 1 1 1 1
## [1] 5 5 2 3 6 1 3 4 2 3 1 1 2 1 3 1 3 2 2 4 1 3 2 1 1
## [26] 7 4 8 1 4 4 2 9 2 4 1 1 4 1 3 4 10 3 4 2 3 3 3 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13
## 19 2 6 14 1 1 1 1 1 1 1 1 1
## [1] 1 4 1 4 4 1 4 1 4 4 1 4 2 1 1 5 6 3 1 4 4 1 1 4 1
## [26] 1 4 1 1 1 3 1 4 3 1 2 4 7 3 8 9 10 11 3 3 4 12 13 1 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## 6 10 2 2 4 8 5 2 2 1 1 1 1 1 1 1 1 1
## [1] 7 1 7 5 10 2 2 1 5 2 6 2 2 7 11 1 2 6 12 13 2 14 15 6 1
## [26] 1 8 1 2 4 4 6 8 6 7 2 6 9 7 6 2 6 3 16 17 3 9 5 18 5
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 16 12 14 2 1 1 1 1 1 1
## [1] 1 2 3 1 3 1 2 1 3 1 3 3 2 2 1 5 3 2 3 1 1 3 6 4 3
## [26] 2 3 1 1 2 3 1 3 2 1 1 3 4 7 1 2 1 8 2 9 10 1 3 2 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11
## 15 16 2 4 2 6 1 1 1 1 1
## [1] 1 2 7 2 2 2 2 2 1 1 4 1 2 2 1 2 1 2 3 2 1 1 1 1 4
## [26] 1 4 2 8 6 6 5 6 3 9 1 1 2 2 2 1 2 4 6 6 10 6 5 1 11
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 20 20 3 1 1 1 1 1 1 1
## [1] 1 1 1 2 1 2 1 4 3 1 1 1 2 2 1 2 3 2 2 1 1 1 3 2 1

```

```

## [26] 2 2 1 1 1 5 2 1 6 1 2 7 1 1 2 2 2 2 2 8 2 9 10 2 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## 14 9 3 9 2 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 4 2 5 6 4 7 2 2 4 2 1 4 2 4 2 5 4 4 2 4 2 1 8 9 1
## [26] 10 11 4 12 2 1 3 13 1 1 1 14 15 16 1 3 1 3 1 1 17 1 18 1 1
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 4 12 9 10 2 9 1 1 1 1
## [1] 2 3 6 2 2 6 2 2 2 3 4 3 4 1 4 2 6 6 6 3 6 7 6 1 4
## [26] 4 3 3 2 8 1 9 4 4 3 2 5 1 3 4 5 6 2 4 2 6 3 4 10 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11
## 15 17 3 8 1 1 1 1 1 1 1
## [1] 3 2 1 1 4 1 4 1 2 1 2 2 3 5 2 1 6 1 7 2 1 8 2 1 4
## [26] 1 4 4 2 4 2 4 9 2 2 10 1 2 2 4 2 1 2 1 2 1 11 1 2 3
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 15 12 2 15 1 1 1 1 1 1
## [1] 1 4 1 1 5 4 1 1 2 4 2 1 1 2 1 6 4 1 1 1 7 1 3 1 4
## [26] 2 2 8 4 1 9 10 2 2 4 4 2 2 4 4 4 1 4 3 2 4 4 2 2 4
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 11 18 9 4 3 1 1 1 1 1
## [1] 2 2 5 3 2 1 5 3 2 2 2 2 2 1 6 3 3 2 2 3 4 7 5 3 2
## [26] 1 1 3 8 2 2 4 3 9 1 1 1 1 2 1 2 4 3 1 4 2 10 2 1 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10
## 15 12 6 6 6 1 1 1 1 1
## [1] 3 1 1 2 1 1 1 2 1 5 1 1 1 5 2 2 2 5 3 4 6 5 4 2 1
## [26] 3 2 2 7 4 1 1 8 3 5 2 5 2 4 2 2 1 1 4 1 9 3 4 3 10
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 14 2 16 11 2 1 1 1 1 1
## [1] 3 1 3 5 3 3 3 3 1 3 1 4 3 3 6 7 3 4 1 8 5 1 3 9 1
## [26] 3 4 3 4 10 1 1 1 1 1 4 1 3 4 4 4 1 4 4 3 4 2 3 1 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
## 11 7 4 11 3 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 4 4 4 4 4 4 4 4 4 6 4 4 4 7 8 5 2 3 3 5 9 10 11 2 1 1
## [26] 2 1 1 12 1 1 5 2 13 14 2 15 1 16 17 18 1 2 19 1 1 3 1 3 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11
## 21 12 5 4 2 1 1 1 1 1 1
## [1] 2 2 1 1 6 2 1 2 1 2 2 7 1 3 1 8 1 1 1 1 1 2 1 9 10
## [26] 1 1 1 11 2 2 2 2 4 1 1 1 3 1 5 3 1 3 3 2 5 4 4 4 1
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 18 4 2 19 2 1 1 1 1 1
## [1] 4 4 1 4 6 4 4 4 4 2 4 1 7 4 1 8 4 4 1 1 4 1 3 9 1
## [26] 4 4 5 10 1 1 4 2 4 3 1 4 1 1 2 1 1 4 1 1 2 1 4 5 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## 8 7 2 12 3 3 2 1 1 1 1 1 1 1 1 1 1 1 1 1

```

```

## [1] 2 4 4 8 1 4 9 10 2 4 4 7 11 6 4 4 4 2 2 1 1 12 1 2 2
## [26] 3 7 6 13 14 15 16 1 5 17 18 4 1 1 6 4 19 4 20 2 5 3 4 1 5
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12
## 9 9 2 2 10 12 1 1 1 1 1 1
## [1] 4 5 5 6 2 2 7 6 8 9 6 1 5 6 6 5 2 5 2 1 5 1 6 4 2
## [26] 1 6 6 10 1 2 1 1 1 5 3 5 6 2 2 6 11 6 5 12 5 6 2 1 3
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13
## 16 10 3 9 4 1 1 1 1 1 1 1 1
## [1] 1 1 1 1 1 6 7 8 5 3 1 1 2 2 9 2 4 1 4 1 4 1 3 4 10
## [26] 4 4 1 1 1 4 11 5 2 12 1 1 1 4 2 4 3 2 2 2 2 2 5 5 13
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13
## 19 6 2 5 3 8 1 1 1 1 1 1 1
## [1] 1 1 6 2 6 2 1 4 1 1 7 1 2 1 1 1 8 2 6 1 5 1 1 2 1
## [26] 5 6 1 5 9 6 10 11 1 4 1 1 2 12 3 1 4 4 6 3 4 6 13 1 6
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 2 10 19 13 1 1 1 1 1 1
## [1] 4 4 4 2 4 2 2 2 2 2 2 4 4 2 2 4 2 3 3 3 3 4 4 3 4
## [26] 3 3 4 5 3 4 3 6 3 4 3 7 1 1 3 8 9 3 3 3 3 3 3 3 10
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22
## 10 7 3 2 4 3 3 4 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 2 2 1 9 1 10 7 3 11 1 12 13 6 14 15 6 2 5 5 1 5 2 3 16 2
## [26] 8 2 17 18 19 8 1 1 8 1 20 7 2 4 3 7 21 1 22 6 4 8 1 5 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26
## 8 6 7 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## 27 28 29 30
## 1 1 1 1
## [1] 3 3 1 6 3 7 8 9 10 11 12 13 4 14 1 5 15 16 17 18 19 1 2 2 20
## [26] 3 5 21 3 2 1 22 23 2 24 1 25 26 1 27 28 1 4 3 1 2 29 3 30 2
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 5 17 21 1 1 1 1 1 1 1
## [1] 3 1 2 3 3 4 1 5 2 6 2 3 1 2 3 7 3 2 3 2 2 2 3 3 2
## [26] 3 3 3 3 3 2 2 3 3 2 3 3 2 2 3 1 2 2 1 3 3 2 8 9 10
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 9 3 16 16 1 1 1 1 1 1
## [1] 1 3 4 2 4 4 4 4 3 2 3 5 4 4 3 4 3 3 1 4 4 4 3 1 4
## [26] 6 2 7 4 3 8 4 1 3 9 3 4 3 1 3 1 10 1 3 3 3 4 3 1 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
## 15 13 5 2 5 1 1 1 1 1 1 1 1 1 1
## [1] 1 2 2 6 1 5 5 1 5 1 1 1 5 1 7 8 9 2 3 10 2 4 3 3 3
## [26] 2 5 1 1 3 4 11 2 2 1 2 1 1 12 2 13 14 1 2 2 1 15 1 2 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14
## 9 10 5 4 6 2 7 1 1 1 1 1 1 1
## [1] 2 3 2 1 1 1 5 7 6 1 4 7 5 7 3 2 2 1 8 5 3 2 3 2 9
## [26] 1 4 7 1 7 4 2 3 2 5 10 1 2 5 7 5 1 11 6 12 4 2 13 7 14

```

```

## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## 11 5 8 11 2 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 4 4 5 2 1 6 1 1 1 1 2 2 7 4 2 4 8 9 10 3 3 3 11 3 1
## [26] 12 1 5 1 13 14 4 4 4 3 15 1 2 4 4 4 3 3 16 4 17 18 1 1 3
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 18 7 2 2 4 13 1 1 1 1
## [1] 1 6 6 5 6 6 6 6 6 5 1 1 4 2 1 6 6 5 2 1 7 1 1 1 2
## [26] 1 1 8 5 2 1 6 1 1 2 1 1 1 1 2 1 6 6 3 3 9 2 4 6 10
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13
## 12 5 7 17 1 1 1 1 1 1 1 1 1
## [1] 4 1 3 1 2 4 2 4 2 4 4 4 1 4 3 1 1 1 3 1 4 3 5 1 3
## [26] 4 4 1 4 6 4 3 7 1 8 3 4 4 1 4 4 9 1 10 2 11 2 4 12 13
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 8 7 8 9 13 1 1 1 1 1
## [1] 1 1 1 3 2 1 1 2 2 5 6 5 5 4 1 4 1 4 2 5 2 1 5 5 3
## [26] 3 7 5 8 4 4 4 3 5 4 9 5 10 5 5 4 3 5 3 5 3 4 2 3 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## 15 5 2 11 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 3 1 6 2 5 5 1 7 3 1 8 1 1 1 9 4 10 11 12 13 1 1 1 14 4
## [26] 1 4 1 15 16 4 1 1 4 4 4 17 4 18 2 1 4 19 4 1 2 2 20 4 2
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 2 17 19 6 1 1 1 1 1 1
## [1] 2 3 3 3 3 2 4 2 2 2 2 2 2 2 4 3 3 4 2 2 5 2 4 1 2
## [26] 4 4 3 3 3 3 6 3 3 3 3 1 7 8 3 9 2 10 3 2 3 2 2 3 3
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## 7 9 5 8 5 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 1 4 5 6 1 3 4 4 7 5 2 3 8 2 5 9 2 10 11 1 2 4 5 1 3
## [26] 12 4 2 1 4 13 14 2 15 1 5 16 17 3 4 2 4 18 19 3 1 2 20 21 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## 14 4 8 2 9 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 5 1 5 5 6 1 5 3 3 5 1 1 7 1 3 1 2 1 1 5 1 2 3 5 1
## [26] 1 3 3 8 5 1 1 2 3 9 10 1 3 11 12 4 13 14 5 15 4 16 2 17 18
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14
## 13 9 5 13 1 1 1 1 1 1 1 1 1 1
## [1] 4 2 1 4 2 4 4 1 3 1 1 2 5 6 4 2 1 3 1 2 7 3 8 2 9
## [26] 4 1 2 10 1 1 4 3 4 2 11 1 4 4 1 2 1 3 4 12 1 13 14 4 4
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11
## 6 7 8 9 14 1 1 1 1 1 1
## [1] 1 3 3 4 2 4 3 5 5 5 3 1 4 5 5 6 2 2 5 5 5 4 3 3 2
## [26] 7 1 8 5 5 9 4 2 4 3 4 2 1 10 5 2 4 3 4 5 1 5 1 11 5
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 19 8 16 1 1 1 1 1 1 1
## [1] 3 1 3 1 1 1 3 1 3 3 3 3 1 3 3 3 3 3 2 1 1 3 4 1 1

```

```

## [26] 3 3 2 1 5 1 2 1 1 2 6 1 1 1 2 3 2 7 8 1 2 1 9 2 10
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 9 13 2 15 2 2 3 2 1 1
## [1] 7 3 2 6 1 2 7 8 7 4 4 2 3 4 8 4 2 9 1 2 4 5 1 4 2
## [26] 4 1 2 4 4 1 2 4 2 1 6 4 5 1 4 4 2 10 2 4 2 4 1 1 2
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## 13 10 8 2 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 2 2 2 2 2 1 3 2 2 2 1 3 2 1 2 3 6 1 4 1 7 8 1 1 9
## [26] 3 3 1 3 10 11 5 1 12 1 1 1 5 13 4 14 15 1 3 16 3 17 18 19 20
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16
## 13 8 15 2 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 1 1 1 2 5 2 6 3 3 3 1 4 3 1 3 2 3 3 7 8 3 9 2 10 4
## [26] 1 3 2 11 12 1 3 2 3 1 3 2 13 14 1 1 3 3 2 15 3 1 1 16 1
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 10 4 17 2 12 1 1 1 1 1
## [1] 3 3 3 2 3 3 3 3 3 5 1 5 3 5 3 1 1 3 2 5 1 3 4 1 1
## [26] 1 1 1 5 5 4 2 2 5 5 6 5 1 3 3 3 7 3 5 8 5 9 3 5 10
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 15 9 3 6 10 2 2 1 1 1
## [1] 3 7 7 1 4 4 4 5 4 3 1 1 2 2 1 1 1 4 2 1 1 1 5 5 2
## [26] 5 5 1 2 3 2 5 4 8 1 1 9 2 5 2 5 1 10 5 2 1 6 1 5 6
## Community sizes
## 1 2 3 4 5 6 7 8 9 10
## 11 5 4 13 2 11 1 1 1 1
## [1] 1 6 5 6 4 4 3 2 6 4 6 1 6 1 4 6 6 1 7 1 4 5 4 4 2
## [26] 2 8 1 4 6 4 6 2 6 3 1 1 3 4 3 1 9 10 2 6 4 4 1 1 4
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 15 7 13 9 1 1 1 1 1 1
## [1] 1 4 2 1 5 2 1 1 2 2 1 4 1 1 3 3 3 6 1 1 4 1 3 3 1
## [26] 3 3 4 2 1 7 8 3 4 2 1 4 3 9 2 4 3 3 1 3 4 4 1 3 10
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 8 3 12 5 17 1 1 1 1 1
## [1] 3 3 3 6 4 5 3 5 5 3 3 3 5 5 5 1 1 1 5 5 1 5 5 5 1
## [26] 3 2 4 5 7 5 5 1 3 2 4 4 1 3 5 4 8 5 3 9 2 3 10 1 5
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 18 4 15 7 1 1 1 1 1 1
## [1] 1 4 2 4 1 1 3 1 2 5 1 6 1 1 7 1 8 3 2 3 1 3 3 2 1
## [26] 1 1 1 3 4 3 3 1 3 1 3 4 1 1 4 4 3 1 3 3 9 3 3 10 4
## Community sizes
## 1 2 3 4 5 6 7 8 9 10
## 17 2 12 13 1 1 1 1 1 1
## [1] 5 3 3 6 7 3 3 1 3 1 1 1 4 4 1 1 1 4 4 1 4 4 1 4 3
## [26] 3 8 4 1 4 4 3 1 1 4 3 1 9 1 2 4 10 3 3 1 3 2 1 4 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25
## 9 4 5 3 7 3 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1

```

```

## [1] 1 1 3 3 1 1 2 3 7 3 3 5 1 5 2 8 4 1 9 1 10 6 5 4 11
## [26] 4 12 13 14 15 16 5 6 17 5 18 19 6 20 5 21 22 5 2 1 23 24 1 2 25
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## 7 5 6 6 3 4 2 4 2 1 1 1 1 1 1 1 1 1 1 1
## [1] 7 4 1 1 3 10 3 6 1 4 8 11 1 4 5 2 12 1 9 5 13 4 6 5 2
## [26] 14 15 2 1 8 16 6 2 4 4 8 3 17 18 6 7 8 3 9 2 3 3 19 20 1
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 14 4 10 5 12 1 1 1 1 1
## [1] 3 5 1 5 3 5 5 3 2 2 4 5 1 1 1 5 3 1 1 4 6 3 2 3 3
## [26] 1 5 1 7 3 1 4 5 4 5 8 1 1 1 9 1 4 3 1 3 5 10 5 2 5
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11
## 10 13 11 9 1 1 1 1 1 1 1
## [1] 5 1 3 2 6 1 3 1 2 2 3 3 4 3 1 2 4 3 4 2 1 2 4 1 7
## [26] 3 2 8 9 4 3 4 2 4 2 10 4 2 3 4 3 1 2 1 2 1 3 2 11 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## 9 5 11 5 5 2 2 1 1 1 1 1 1 1 1 1 1 1
## [1] 8 1 5 1 5 6 6 4 7 3 9 3 10 7 1 11 4 3 1 3 2 1 5 3 1
## [26] 3 12 3 3 13 5 14 1 2 3 3 4 15 16 2 4 17 2 2 3 18 4 1 5 1
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 17 7 7 13 1 1 1 1 1 1
## [1] 1 4 1 1 2 5 1 3 4 4 1 6 1 3 1 1 1 1 2 4 1 7 8 2 3
## [26] 3 4 1 1 3 4 9 4 1 2 2 4 4 4 4 2 1 1 3 4 3 2 10 4 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
## 9 7 8 4 4 3 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 7 2 5 2 5 5 8 5 9 10 6 3 3 3 1 1 11 12 2 4 2 1 4 2 6
## [26] 1 3 13 1 14 6 15 2 1 2 16 3 1 17 1 18 3 19 20 1 4 4 3 21 3
## fg.10_comm
## 1 2 3 4 5 6 7 8 9 10
## 16 14 9 3 2 2 1 1 1 1
## [1] 4 4 2 2 3 2 2 3 2 2 5 3 6 2 1 1 2 2 3 2 3 2 1 1 4
## [26] 1 1 7 2 6 1 8 1 1 1 5 2 3 9 1 3 3 2 1 1 1 3 10 1 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
## 11 6 15 2 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 5 1 6 1 3 3 3 3 1 1 1 3 3 3 3 1 3 2 3 1 7 8 9 2 2
## [26] 1 1 4 10 4 3 11 2 3 1 12 3 3 13 14 15 2 16 3 2 17 18 19 20 1
## Community sizes
## 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18
## 11 9 5 4 7 2 1 1 1 1 1 1 1 1 1 1 1 1
## [1] 2 1 1 2 5 2 1 5 7 3 2 4 5 2 1 1 5 1 1 8 5 2 1 5 9
## [26] 10 4 11 12 1 5 2 2 13 14 4 2 6 15 4 1 3 16 3 1 3 3 17 18 6

```

```
length(which(actual_clusters <10))
```

```
## [1] 29
```

```
actual_clusters
```

```
## [1]  9  8  3  8  7  9 13 18  8 11  8 18  6 11  8  8 10  7 19 11  4 20 12 13 13
## [26]  6 22 30  4  5 15 14 18  7 13  7 20  8 21 18 14 11  5  8 20 16  5  8 10  5
## [51]  9  6 10 25 20  7 11 18  6 21  9 20 18
```

```
mean(actual_clusters)
```

```
## [1] 11.93651
```

```
final_clusters
```

```
## [1] 10 10 10 10 10 10 13 18 10 11 10 18 10 11 10 10 10 10 19 11 10 20 12 13 13
## [26] 10 22 30 10 10 15 14 18 10 13 10 20 10 21 18 14 11 10 10 20 16 10 10 10 10
## [51] 10 10 10 25 20 10 11 18 10 21 10 20 18
```

```
#read rel_results of top k from file
```

```
rel_results <- read.table("relevant_doc_list_top_50_for_63q.csv",sep = ',',header = FALSE, nrows = 63)
```

```
#rel_results
```

```
new_rel_list_10 = list()
```

```
new_rel_df_10 <- data.frame()
```

```
for (j in 1:63)
```

```
{
```

```
  rel_vec <- c()
```

```
  for (k in 1:10)
```

```
  {
```

```
    if (new_top_10[[j]][k] %in% (which(rel_results[j,] == 1)))
```

```
    {
```

```
      rel_vec <- append(rel_vec,1)
```

```
    }
```

```
    else
```

```
    {
```

```
      rel_vec <- append(rel_vec,0)
```

```
    }
```

```
  }
```

```
  new_rel_list_10 <- append(new_rel_list_10,list(rel_vec))
```

```
  new_rel_df_10 <- rbind(new_rel_df_10,rel_vec)
```

```
}
```

```
# write new rel_list to file for jupyter to read
```

```
write.table(new_rel_df_10,'new_relevant_docs_top_10.csv',sep = ',',row.names = FALSE,col.names = FALSE)
```

```
#Fetching similarity scores before and after
```

```
#new_top_10
```

```
#orig_top_50
```

```
sum_old = 0
```

```
sum_new = 0
```

```
for (i in 1:63) {
```

```

sim_matrix <- read.table("similarity_matrix_for_graph.csv",sep = ',',header = TRUE, check.names = FALSE)

#DF to matrix
data_matrix_k50 <- data.matrix(sim_matrix)
#print(data_matrix_k50)

#old sim matrix
#first subtract 10 because we dont want similarity if i,i
print(i)
print((sum(data_matrix_k50[1:10,1:10]) - 10) / 90)

#new sim matrix
print((sum(data_matrix_k50[new_top_10[[i]],new_top_10[[i]]]) - 10)/ 90)
if ((sum(data_matrix_k50[1:10,1:10]) - 10)/ 90 < (sum(data_matrix_k50[new_top_10[[i]],new_top_10[[i]]]))
{
  print("Similarity score did not decrease")
}

sum_old <- sum_old + (sum(data_matrix_k50[1:10,1:10]) - 10)/ 90

sum_new <- sum_new + (sum(data_matrix_k50[new_top_10[[i]],new_top_10[[i]]]) - 10)/ 90

}

```

```

## [1] 1
## [1] 0.2563354
## [1] 0.0837172
## [1] 2
## [1] 0.2460477
## [1] 0.2379253
## [1] 3
## [1] 0.2443066
## [1] 0.1699285
## [1] 4
## [1] 0.3706662
## [1] 0.08869959
## [1] 5
## [1] 0.4366277
## [1] 0.2413287
## [1] 6
## [1] 0.0729461
## [1] 0.03478602
## [1] 7
## [1] 0.2334765
## [1] 0.03366682
## [1] 8
## [1] 0.05842281
## [1] 0.02276489
## [1] 9
## [1] 0.1897575

```



```
## [1] 0.07015433
## [1] 10
## [1] 0.1861767
## [1] 0.08523318
## [1] 11
## [1] 0.2123987
## [1] 0.09005878
## [1] 12
## [1] 0.101768
## [1] 0.01952708
## [1] 13
## [1] 0.2317843
## [1] 0.1729311
## [1] 14
## [1] 0.1643413
## [1] 0.06887271
## [1] 15
## [1] 0.2125598
## [1] 0.1210474
## [1] 16
## [1] 0.2889223
## [1] 0.2241185
## [1] 17
## [1] 0.3434223
## [1] 0.1437698
## [1] 18
## [1] 0.2678921
## [1] 0.1297733
## [1] 19
## [1] 0.1531355
## [1] 0.02233494
## [1] 20
## [1] 0.2445565
## [1] 0.05713545
## [1] 21
## [1] 0.2037441
## [1] 0.1300189
## [1] 22
## [1] 0.1166815
## [1] 0.05890232
## [1] 23
## [1] 0.1262211
## [1] 0.1520665
## [1] "Similarity score did not decrease"
## [1] 24
## [1] 0.1314883
## [1] 0.07952669
## [1] 25
## [1] 0.1410797
## [1] 0.02416795
## [1] 26
## [1] 0.2859148
## [1] 0.1005429
## [1] 27
```

```
## [1] 0.06948585
## [1] 0.02113052
## [1] 28
## [1] 0.06776753
## [1] 0.04351239
## [1] 29
## [1] 0.1672062
## [1] 0.150239
## [1] 30
## [1] 0.370146
## [1] 0.189698
## [1] 31
## [1] 0.1274758
## [1] 0.03818007
## [1] 32
## [1] 0.1024267
## [1] 0.04696082
## [1] 33
## [1] 0.1317069
## [1] 0.02822534
## [1] 34
## [1] 0.1850538
## [1] 0.07935507
## [1] 35
## [1] 0.179695
## [1] 0.06774597
## [1] 36
## [1] 0.2813692
## [1] 0.1664044
## [1] 37
## [1] 0.05740431
## [1] 0.03020277
## [1] 38
## [1] 0.259178
## [1] 0.07015495
## [1] 39
## [1] 0.04919711
## [1] 0.01443058
## [1] 40
## [1] 0.206638
## [1] 0.0708872
## [1] 41
## [1] 0.1517131
## [1] 0.06421687
## [1] 42
## [1] 0.1115283
## [1] 0.05800115
## [1] 43
## [1] 0.4072721
## [1] 0.2634537
## [1] 44
## [1] 0.1898762
## [1] 0.1347691
## [1] 45
```

```
## [1] 0.2612499
## [1] 0.0334909
## [1] 46
## [1] 0.1620739
## [1] 0.07055952
## [1] 47
## [1] 0.3239874
## [1] 0.1976758
## [1] 48
## [1] 0.1366266
## [1] 0.1191719
## [1] 49
## [1] 0.1939679
## [1] 0.106183
## [1] 50
## [1] 0.1793986
## [1] 0.0918093
## [1] 51
## [1] 0.248814
## [1] 0.120699
## [1] 52
## [1] 0.2740029
## [1] 0.1923383
## [1] 53
## [1] 0.149835
## [1] 0.07213528
## [1] 54
## [1] 0.2216449
## [1] 0.0578513
## [1] 55
## [1] 0.02111827
## [1] 0.004046148
## [1] 56
## [1] 0.163082
## [1] 0.06020714
## [1] 57
## [1] 0.2648054
## [1] 0.1493191
## [1] 58
## [1] 0.018704
## [1] 0.009513986
## [1] 59
## [1] 0.2399083
## [1] 0.1268516
## [1] 60
## [1] 0.03718176
## [1] 0.02278422
## [1] 61
## [1] 0.1444798
## [1] 0.0314394
## [1] 62
## [1] 0.1719409
## [1] 0.05146471
## [1] 63
```

```
## [1] 0.1217586  
## [1] 0.03119817
```

```
#mean of (avg (sim score for every i,j))  
print("Before and after")
```

```
## [1] "Before and after"
```

```
sum_old/63
```

```
## [1] 0.1900062
```

```
sum_new/63
```

```
## [1] 0.09125882
```