

WNS Assignment 5

PART 1

Reading the graph

Reference to Code Appendix- [read_graph](#)

The first step is to read the graph using the igraph method and make it undirected. I also removed the edges that pointed to the same vertices they emerged from by using simplify() method.

Finding out cue words from the graph

Reference to Code Appendix- [scan](#)

In the next step, I read the file- cue.txt using the scan() method. After reading it in a variable I set it as a vertex attribute. This will help us to eliminate those words that are not cue words. After this I deleted the non-cue words from the graph.

Pre-processing I

Reference to Code Appendix- [deleteing_vertices](#)

This involved deletion of those nodes that had degree 0. I also scaled the edge weights from 0 to 1 at this point.

edgedeletion() function

Reference to Code Appendix- [edgedeletion](#)

I created a function that takes a graph and a numeric value (threshold) as parameter and deletes those edges from the graph that have edge weight less than the threshold. This method is called for each of my target words (ART, KNOWLEDGE, MIND) to delete edges that have weights less than threshold.

Random Walk

Reference to Code Appendix- [random_walk](#)

I now create 3 vectors that will store the results of random walks for my target words. Then using a for loop that runs 1000 times, I called the random_walk() method from each target word and got the neighborhood around each target word in these vectors.

vertexdeletion() function

Reference to Code Appendix-[vertexdeletion](#)

This method I created takes a graph and the random_walk_vector as input and deletes those vertices that are not in the neighbourhood, by deleting vertices that are not in the random_walk_vector. After calling this method for each of my target words I get 3 graphs. Each of these graphs are a subset of the original graph and consists of only the neighborhood of my target word.

Pre-processing II

After this I delete from each of the three graphs those vertices that have degree=1.

plotgraph() function

Reference to Code Appendix-[plotgraph](#)

This function takes an igraph as an input and returns a ggraph plot. Using this function I have plotted association networks for each of my target words.

Authority measures and authority scores for each target word

Reference to Code Appendix-[AUTHORITY_MEASURES](#)

I have selected Page Rank algorithm for authority score calculation of target word- ART, this is done with the help of page_rank() method. Similarly for target word-KNOWLEDGE I have calculated the authority score using authority_score() method. Finally for the third target word- MIND I have used degree centrality for calculating the authority. Below is a snippet of the authorities in each target word.

| Target Word: ART | | Target Word: KNOWLEDGE | | Target Word: MIND | |
|------------------|-----------------|------------------------|-----------------|-------------------|-----------------|
| | Authority Score | | Authority Score | | Authority Score |
| ART | 0.0965221 | SMART | 1.0000000 | THINK | 14 |
| PICTURE | 0.0720184 | DUMB | 0.5671321 | MIND | 13 |
| DRAW | 0.0432245 | INTELLIGENT | 0.4771101 | BRAIN | 12 |

PART 2:

Using Louvain algorithm for community detection

Reference to Code Appendix-[clusterdf](#)

I have used Louvain algorithm for community detection which can be done with the help of cluster_louvain() method. This is extremely easy to use and simple and efficient. I have created a method called clusterdf() which takes an igraph as a parameter and returns a dataframe that consists of Cluster Number, Members of that cluster (Words In our case), Total count of members in that cluster. I had called this method for each of my target words and got the Cluster Information as shown below

| ART | | |
|----------------|---|-------|
| Cluster Number | Members | Count |
| 1 | ABILITY SKILL TALENT | 3 |
| 2 | MAKE DESTROY DESIGN CREATE INVENT DEVELOP | 6 |
| 3 | ABSTRACT ART PAINTING DRAWING SCULPTURE CRAFT CREATIVE ARTS CRAFTS MUSEUM WRITING | 11 |
| 4 | HOUSE ARTIST PAINTER PAINT WALL BRUSH PLASTER CREATIVITY | 8 |
| 5 | CANVAS SHOES TENT CANVASS | 4 |
| 6 | GIRL BEAUTY GOD MATERIAL CREATION GODDESS | 6 |
| 7 | DRAW PENCIL WRITE SKETCH COMPOSE SCRIBBLE | 6 |
| 8 | CAMERA PICTURE PHOTO FILM DIAGRAM PORTRAIT | 6 |

KNOWLEDGE

| Cluster Number | Members | Count |
|----------------|---|-------|
| 1 | KNOWLEDGE GOOD NEWSPAPER NEWS WISDOM LOOK INFORMATION EXPERIENCE INSIGHT | 9 |
| 2 | MIND THOUGHT THINK HEAD DECIDE MEMORY NEURON NERVE BRAIN SKULL THINKING MENTAL | 12 |
| 3 | STUPID SMART LOGIC GENIUS BRIGHT INTELLIGENCE INTELLIGENT BRILLIANT EXCEPTIONAL WIT CLEVER EINSTEIN GIFTED INTELLECT | 14 |
| 4 | DICTIONARY BOOK ENCYCLOPEDIA READ BRITANNICA | 5 |
| 5 | WORK SCHOOL LEARN TEACH TEACHER STUDY PROFESSOR EDUCATION LEARNING LESSON EDUCATE | 11 |
| 6 | UNDERSTAND ACKNOWLEDGE KNOW RECOGNITION REALIZE SYMPATHETIC UNDERSTANDING COMPREHEND INTUITION PERCEIVE | 10 |

MIND

| Cluster Number | Members | Count |
|----------------|--|-------|
| 1 | THOUGHT IDEA OPINION FACT NOTION | 5 |
| 2 | EASY COMPLEX BASIC HARD SIMPLE EASE | 6 |
| 3 | HEAD ATOM CELL NEURON NERVE BRAIN SKULL THINKING | 8 |
| 4 | MIND PSYCHOLOGY CLASS AWAKE AWARE SCIENCE EXPERIMENT CONSCIOUS SUBJECT CONSCIENCE MATTER FEEBLE | 12 |
| 5 | SMART LOGIC INTELLIGENCE IMAGINATION CREATIVITY INTELLECT | 6 |
| 6 | BODY PHYSICAL ANATOMY PHYSIOLOGY MENTAL | 5 |
| 7 | FORGET MEMORY REMEMBER RECALL REMINISCENCE | 5 |
| 8 | THINK WONDER STUDY ANALYZE DREAM GUESS SUPPOSE CONCENTRATE PONDER IMAGINE | 10 |

Interpretation

Reference to Code Appendix-[Interpretation](#)

ART CLUSTERS INTERPRETATION

Cluster No. Interpretation

- | | |
|---|---|
| 1 | Art as a kind of Talent |
| 2 | Art for Renovation |
| 3 | Art for Restoration |
| 4 | Result of Art like House/Artist/Painter etc |
| 5 | Art Canvas |
| 6 | Art as in Divine Creation |
| 7 | Book Art like Sketching/Scribbling |
| 8 | Digital Art |

KNOWLEDGE CLUSTERS INTERPRETATION

Cluster No. Interpretation

- | | |
|---|---|
| 1 | Getting Knowledge with Experience and Information |
| 2 | Where knowledge is probably stored |
| 3 | Classification of a person based on the knowledge they have |
| 4 | Resources to obtain knowledge |
| 5 | Academic and Educational Knowledge |
| 6 | Traits of a person having knowledge |

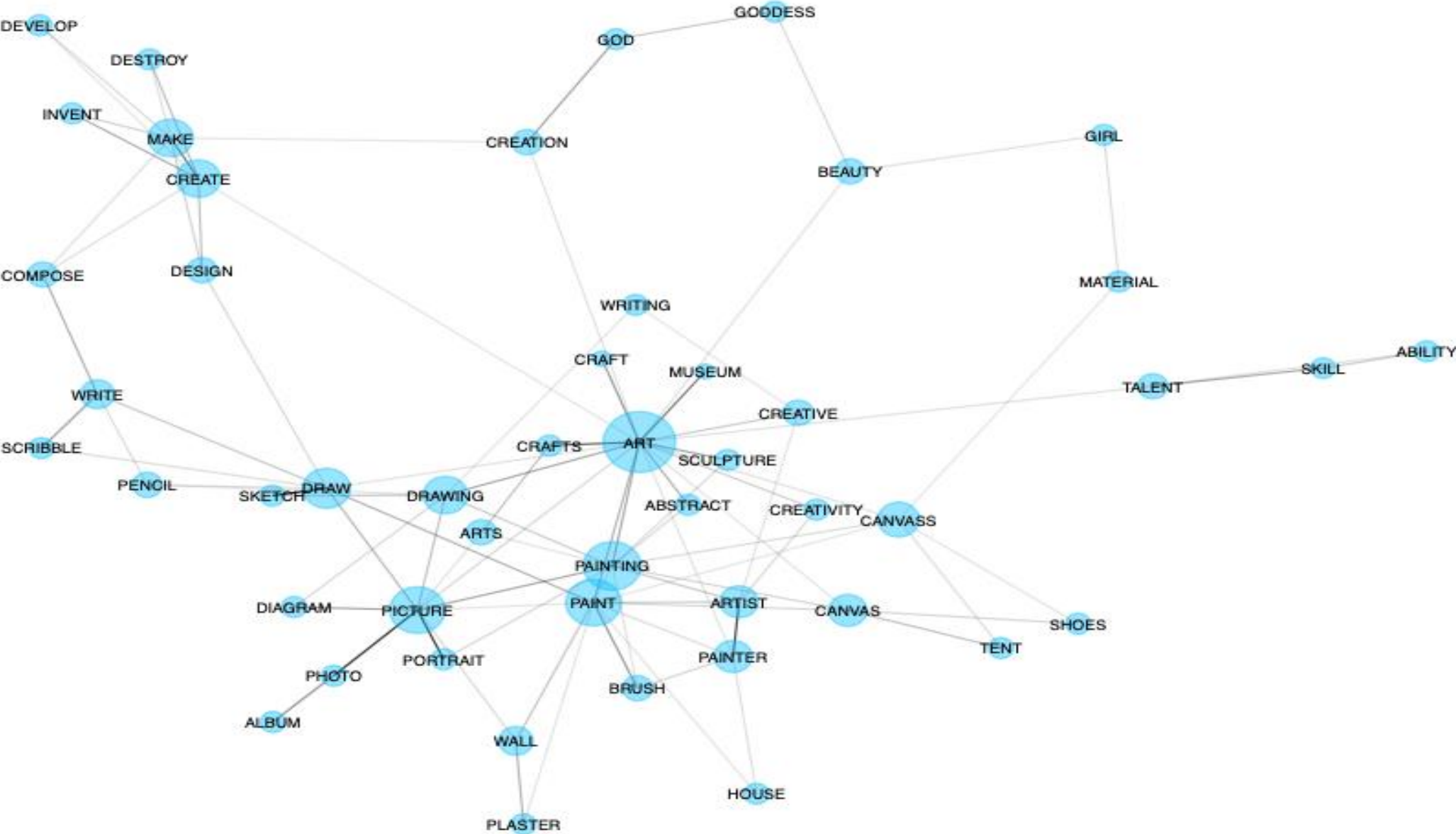
MIND CLUSTERS INTERPRETATION

Cluster No. Interpretation

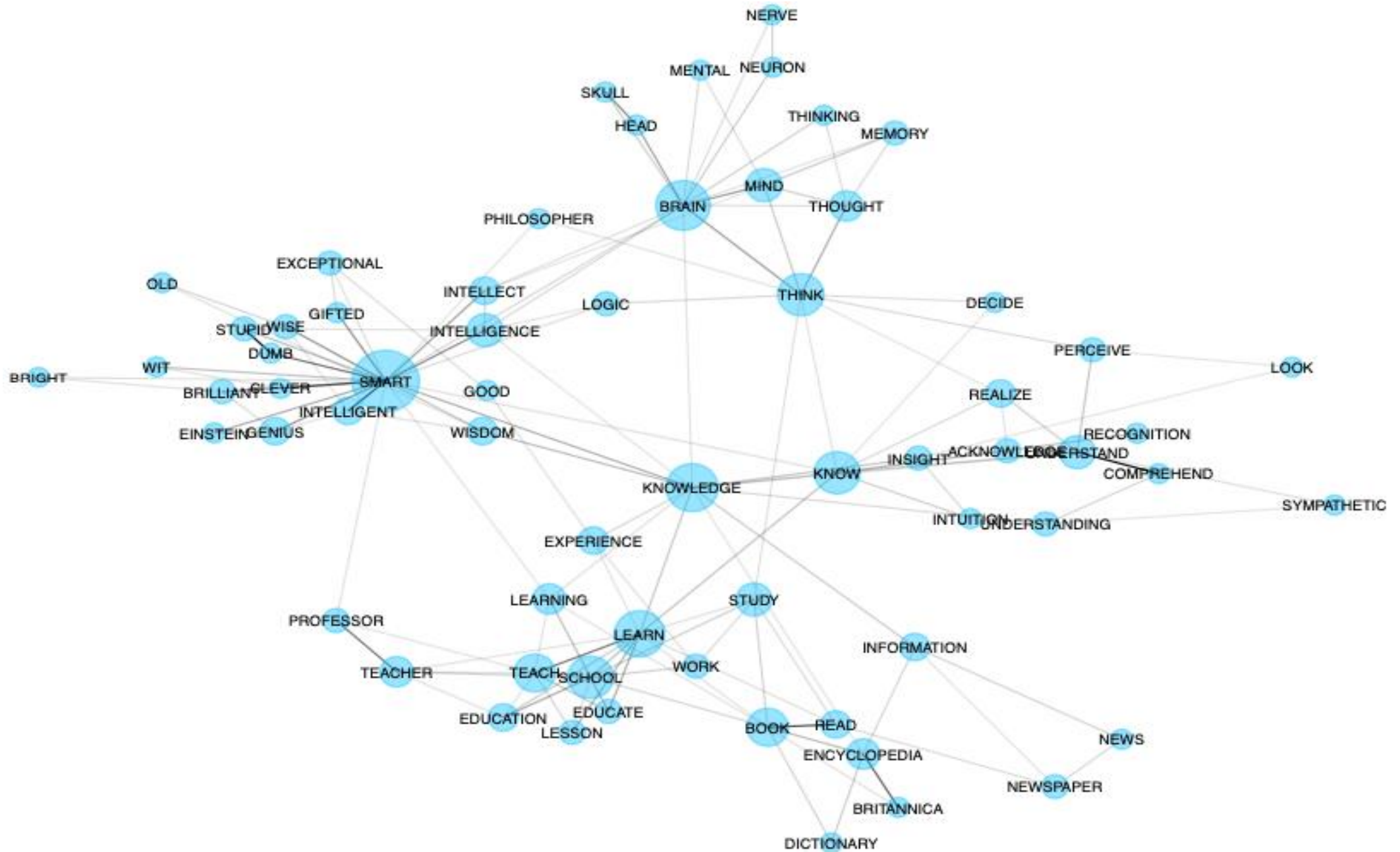
- | | |
|---|---|
| 1 | Getting Knowledge with Experience and Information |
| 2 | Where knowledge is probably stored |
| 3 | Classification of a person based on the knowledge they have |
| 4 | Resources to obtain knowledge |
| 5 | Academic and Educational Knowledge |
| 6 | Traits of a person having knowledge |

GRAPH PLOTS

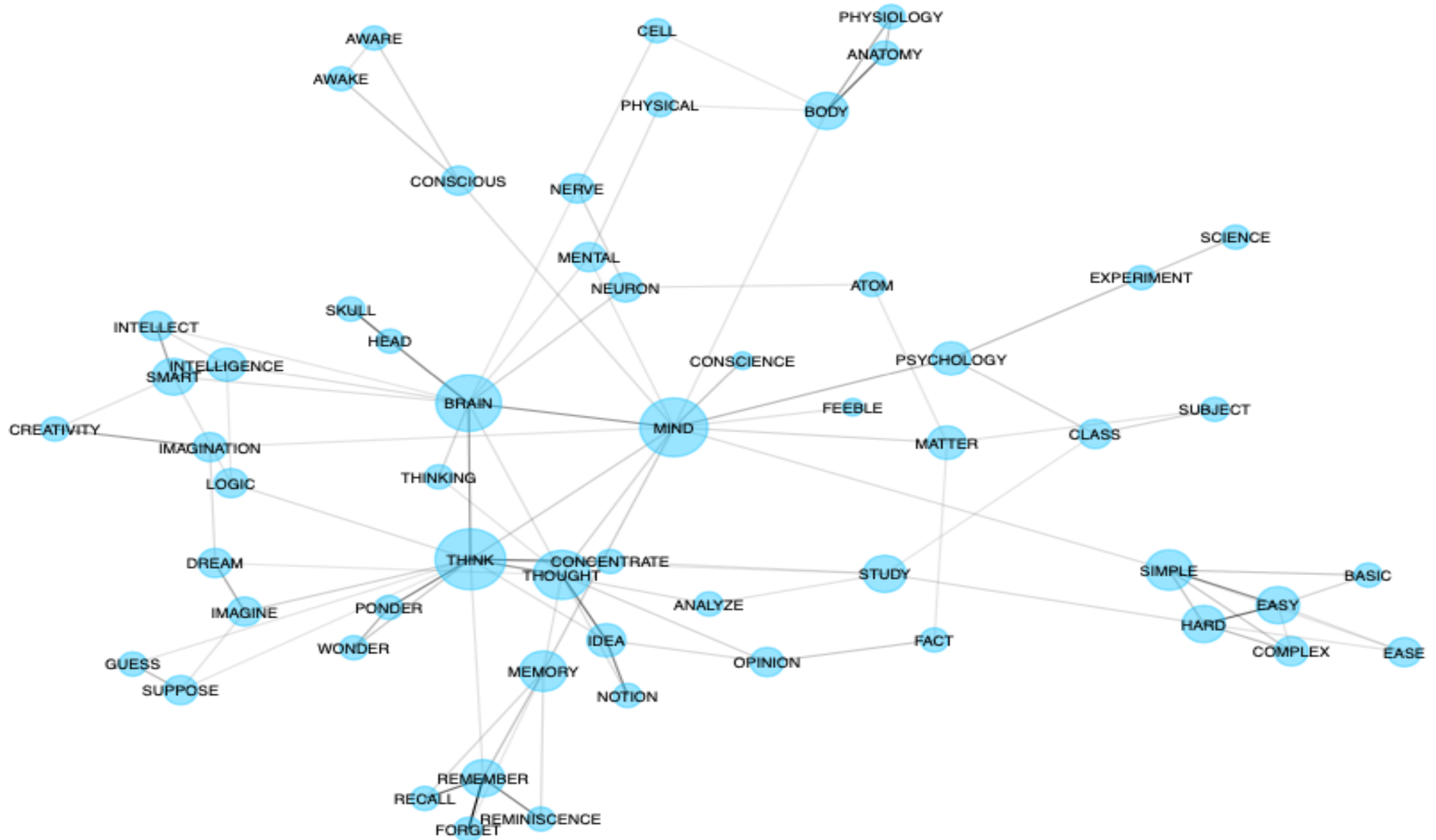
ART



KNOWLEDGE



MIND



REFERENCES:

Kable: <https://www.rdocumentation.org/packages/knitr/versions/1.32/topics/kable>

Converting .txt file to vector for reading cue.txt: <https://stackoverflow.com/questions/23678691/converting-a-text-file-into-a-vector-in-r>

Authority score using Page Rank, Degree Centrality, Kleinberg's Method:

https://nuigalway.blackboard.com/webapps/blackboard/execute/content/file?cmd=view&content_id=2331181_1&course_id=123624_1&framesetWrapped=true

Louvain algorithm: https://nuigalway.blackboard.com/bbcswebdav/pid-2344041-dt-content-rid-21696439_1/courses/2021-CT5113/2021-Community-Detection-Algorithms.html

CODE APPENDIX

```
#importing libraries
library(igraph)
library("scales")
library(ggraph)
library(knitr)
library(tidyr)
library(kableExtra)

#read the graph
g<- read_graph(file="WordPairs.txt",format="pajek")
#make it undirected
g<- as.undirected(g)
#removing the edges pointing to the same node they emerge from
g<-simplify(g)

#reading cue.txt
cueval <- scan("cue.txt", character(), quote = '')
#removing unwanted values
cueval <- cueval[-1:-24]
#making it numeric
cueval <- as.numeric(cueval)
#setting cue attribute to vertex
g <- set_vertex_attr(g, "cueindicator", value = cueval )
length(V(g))
#deleting non cuewords from graph
```

```

g<- delete.vertices(g,which(V(g)$cueindicator == 0)) # remove nodes with degree zero
length(V(g))
#deleteing_vertices with zero degree
g<- delete.vertices(g,which(degree(g)==0))
length(V(g))
#scaling weights from 0 to 1
E(g)$weight<-rescale(E(g)$weight)
#Copying g into three different graphs
gart<-g
gknowledge<-g
gmind<-g

edgedeleletion <- function(gvar, wthreshold)
{
  #deleting edges in gvar that are smaller than the threshold
  return(delete.edges(gvar, which(E(gvar)$weight < wthreshold)))
}
#deleting edges
gart<- edgedeleletion(gart,0.028)
gknowledge<- edgedeleletion(gknowledge,0.026)
gmind<- edgedeleletion(gmind, 0.032)
#KNOWLEDGE, MIND, ART
t1<- V(gart)$name %in% c("ART")
t1 <- V(gart)[t1] #Stores node ART
t2<- V(gknowledge)$name %in% c("KNOWLEDGE")
t2 <- V(gknowledge)[t2] #Stores node KNOWLEDGE

```

```

t3<- V(gmind)$name %in% c("MIND")
t3 <- V(gmind)[t3] #Stores node MIND
#vectors to store random walk result
walk_rand_art<-c()
walk_rand_knowledge<-c()
walk_rand_mind<-c()
#for loop to run random walk 1000 times
for(i in 1:1000)
{
  walk_rand_art <- c(walk_rand_art, random_walk(gart, start= t1, steps=3, stuck = "return"))
  walk_rand_knowledge <- c(walk_rand_knowledge, random_walk(gknowledge, start= t2, steps=3, stuck = "return"))
  walk_rand_mind <- c(walk_rand_mind, random_walk(gmind, start= t3, steps=3, stuck = "return"))
}
vertexdeletion <- function(gvar, walk_rand_var)
{
  vec<- as.numeric(rownames(table(walk_rand_var)))
  # remove nodes not in random walk
  gresult<- delete.vertices(gvar,V(gvar)[which(!(V(gvar)$name %in% V(gvar)[vec]$name))])
  return(gresult)
}
# calling vertexdeletion to delete vertices not in random walk
gart<-vertexdeletion(gart,walk_rand_art)
gknowledge<-vertexdeletion(gknowledge,walk_rand_knowledge)
gmind<-vertexdeletion(gmind,walk_rand_mind)

gart<- delete.vertices(gart,which(degree(gart)==1)) # remove nodes with degree one

```

```

gknowledge<- delete.vertices(gknowledge,which(degree(gknowledge)==1)) # remove nodes with degree one
gmind<- delete.vertices(gmind,which(degree(gmind)==1)) # remove nodes with degree one
# function to plot graphs
plotgraph <- function(gvar)
{
  gnew<- gggraph(gvar, layout = "fr") +
  geom_edge_link2(aes(edge_alpha = weight),
                  edge_width=0.3, show.legend = FALSE) +
  geom_node_point(aes(size=degree(gvar)),
                  color = "deepskyblue",
                  alpha=0.4, show.legend = FALSE) +
  geom_node_text(aes(label = name), size = 1.75,
                 repel = FALSE) +
  scale_size_area(max_size=10) +
  theme_void()
  return(gnew)
}
# plotting graphs
artplot <- plotgraph(gart)
plot(artplot)
knowledgeplot <- plotgraph(gknowledge)
plot(knowledgeplot)
mindplot <- plotgraph(gmind)
plot(mindplot)
#AUTHORITY_MEASURES
### PAGE RANK FOR ART####

```

```

alpha = 0.85

pr<- page_rank(gart, algo = "arpack", directed = TRUE, damping = alpha ,personalized = NULL, options = NULL)
as1<-sort(pr$vector, TRUE)[1:3]

knitr::kable(as1,row.names=TRUE,col.names=c("Authority Score"), "html")

####AUTHORITY SCORE FOR KNOWLEDGE####

authority<- authority_score(gknowledge, scale = TRUE)$vector
as2<-sort(authority, TRUE)[1:3]

knitr::kable(as2,row.names=TRUE,col.names=c("Authority Score"), "html")

####DEGREE CENTRALITY FOR MIND ####

degreecent<-degree(gmind)
as3<-sort(degreecent, TRUE)[1:3]

knitr::kable(as3,row.names=TRUE,col.names=c("Authority Score"), "html")

###Louvain algorithm###

clusterdf <- function(gvar)
{
  set.seed(4292)

  communitiesvar<-cluster_louvain(gvar) #Louvain algorithm for community detection
  tmpr<-as.list(membership(communitiesvar)) #Members stored in tmpr as a list
  wordlist<- names(tmpr) #Names(Words) stored in wordlist
  clusternumber <- matrix(0, length(names(tmpr)))
  df <- data.frame(wordlist,clusternumber) #Dataframe having wordlist and clusternumbers (all zeros at this point)
  for( 1 in names(tmpr))
  {
    rownum<- which(df$wordlist==1) #Getting rownumber which has word 1
    df[rownum,2]<-tmpr[[1]] #Storing clusternumber in that rownumber and column 2
  }
}

```

```

clusternames<-unique(unlist(as.list(df["clusternumber"]), use.names = FALSE)) #Storing all unique clusternumbers
newdf <- matrix(ncol = 2, nrow=0)
newdf <- data.frame(newdf)
for(s in clusternames) #For each clusternumber
{
  rowindex<-which(df["clusternumber"]==s) #get the position in dataframe df
  newlist<-c()
  for(r in rowindex) #at all those position we also have word in column word
  {
    newlist<-c(newlist,df[r,1]) #append each word to newlist
  }
  comcount<-length(newlist) #get the count
  newlist<- paste( unlist(newlist), collapse=' ')
  newdf<-rbind(newdf,c(s,newlist,comcount)) #store both words list (newlist) and cluster numbers in newdf
}
colnames(newdf)<-c("Cluster Number","Members","Count") #giving column headings
newdf<-newdf[order(newdf["Cluster Number"]),] #sorting
return(newdf)
}

artdf<-clusterdf(gart)
knowledgedf<-clusterdf(gknowledge)
minddf<-clusterdf(gmind)
#plotting the tables using kable
knitr::kable(artdf,row.names=FALSE, "html")%>% add_header_above(header= c("ART"=3))
knitr::kable(knowledgedf,row.names=FALSE, "html")%>% add_header_above(header= c("KNOWLEDGE"=3))

```



```
knitr::kable(minddf,row.names=FALSE, "html")%>% add_header_above(header= c("MIND"=3))
```

```
#Interpretation
```

```
artint<-c(" Art as a kind of Talent"," Art for Renovation"," Art for Restoration"," Result of Art like House/Artist/Painter  
etc"," Art Canvas"," Art as in Divine Creation"," Book Art like Sketching/Scribbling"," Digital Art")
```

```
aclnum<-seq(1:8)
```

```
artint<-data.frame(aclnum,artint)
```

```
colnames(artint)<-c("Cluster No.,"Interpretation")
```

```
knitr::kable(artint,row.names=FALSE,align = "ll", "html")%>% add_header_above(header= c("ART CLUSTERS INTERPRETATION"=2))
```

```
knowledgeint<-c("Getting Knowledge with Experience and Information","Where knowledge is probably stored","Classification of  
a person based on the knowledge they have","Resources to obtain knowledge","Academic and Educational Knowledge","Traits of a  
person having knowldege")
```

```
kclnum<-seq(1:6)
```

```
knowledgeint<-data.frame(kclnum,knowledgeint)
```

```
colnames(knowledgeint)<-c("Cluster No.,"Interpretation")
```

```
knitr::kable(knowledgeint,row.names=FALSE,align = "ll", "html")%>% add_header_above(header= c("KNOWLEDGE CLUSTERS INTERPRETA  
TION"=2))
```

```
mindint<-c("Something mind does","Comprehending ability of Mind","Part of body where mind if thought to be there","Mind in t  
he context of Science and experiments","Classifying thought process of a person as in creative, intellectual, etc","Mind as  
in a body part","Ability of mind to recollect or process past information","Something that puts mind to work")
```

```
mclnum<-seq(1:8)
```

```
mindint<-data.frame(mclnum,mindint)
```

```
colnames(mindint)<-c("Cluster No.,"Interpretation")
```

```
knitr::kable(knowledgeint,row.names=FALSE,align = "ll", "html")%>% add_header_above(header= c("MIND CLUSTERS INTERPRETATION"  
=2))
```