

# SNA Closeness 1.2 - ALGUM RELACIONAMENTO (x2)

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SNA Descriptive Analysis from “Projeto Redes de Atenção às pessoas que consomem álcool e outras Drogas em Juiz de Fora-MG Brazil” - SNArRDJF

Here you can find a basic script to analysis data from SNArRDJF - this script was elaborated considering its use for orther matrix adjacency data from SNArRDJF - Here we are going to analyse:

## 1 1.2 - ALGUM RELACIONAMENTO (x2)

##### Basic Preparation ##### ‘#####

## 2 Loading objects generated with previous script

```
rm(list = ls()) # removing previous objects to be sure that we don't have objects conflicts name
load("~/SNArRDJF/Robjct/2_degree_x2.RData")
```

### 2.1 Reload packages

```
suppressMessages(library(RColorBrewer))
suppressMessages(library(car))
suppressMessages(library(xtable))
suppressMessages(library(igraph))
suppressMessages(library(miniCRAN))
suppressMessages(library(magrittr))
suppressMessages(library(keyplayer))
suppressMessages(library(dplyr))
suppressMessages(library(feather))
suppressMessages(library(visNetwork))
suppressMessages(library(knitr))
suppressMessages(library(DT))
```

### 2.2 Adding phantom tools

```
#In order to get dinamic javascript object install those ones. If you get problems installing go to Sta
#devtools::install_github("wch/webshot")
#webshot::install_phantomjs()
```

### 2.3 Setting a random seed - this is a good strategy to keep the same graph pattern layout in a new report generation

```
set.seed(123)
```

## 2.4 Simplify Graph - removing loops and duple edges

```
x2<-simplify(x2) #Simplify
```

## 3 Closeness - centrality based on distance to others in the graph

*How close an actor to all the other actors in network?*

High closeness centrality - short communication path to others, minimal number of steps to reach others.

Answers the “Kevin Bacon” question:

*How many steps are required to access every other vertex from a given vertex?*

One practical implication of this metric: it helps you gauge how information might spread within your network, and who might be the best people to leverage if you need to make sure information gets around. Link here: <http://www.tc.umn.edu/~alink/R-social-network-analysis.html>

Closeness centrality can be defined as a measure of how far other nodes are from the node in question. Nodes with high closeness centrality are likely to be relatively efficient in receiving or transmitting information to/from distant parts of the social network.

Scores may be interpreted as arising from a reciprocal process in which the centrality of each actor is proportional to the sum of the centralities of those actors to whom he or she is connected.

In general, vertices with high eigenvector centralities are those which are connected to many other vertices which are, in turn, connected to many others (and so on). (The perceptive may realize that this implies that the largest values will be obtained by individuals in large cliques (or high-density substructures))

### 3.1 Closeness Non-normalized

#### 3.1.1 Saving to Igraph object

```
V(x2)$incloseness <- closeness(x2, mode = "in", weights = E(x2)$x2) %>% round(6)
V(x2)$outcloseness <- closeness(x2, mode = "out", weights = E(x2)$x2) %>% round(6)
V(x2)$totalcloseness <- closeness(x2, mode = "total", weights = E(x2)$x2) %>% round(4)
```

#### 3.1.2 Saving to Environment

```
x2_incloseness<- closeness(x2, mode = "in", weights = E(x2)$x2) %>% round(6)
x2_outcloseness<- closeness(x2, mode = "out", weights = E(x2)$x2) %>% round(6)
x2_totalcloseness<- closeness(x2, mode = "total", weights = E(x2)$x2) %>% round(6)
```

#### 3.1.3 Closeness Non-normalized - in

```
summary(x2_incloseness)
```

```
##      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
## 0.0000290 0.0001310 0.0001320 0.0001308 0.0001320 0.0001360
```

```
sd(x2_incloseness)
```

```
## [1] 7.64145e-06
```

### 3.2 Network Plotting Based On Non-normalized Closeness - IN

```
V(x2)$incloseness<-closeness(x2, weights = E(x2)$x2, mode="in")

#Get Variable
V(x2)$x2_color_degree<-round(V(x2)$incloseness,4)

#Creating brewer palette
vertex_x2_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
    V(x2)$x2_color_degree)), "RdBu"))(
    length(unique(V(x2)$x2_color_degree)))

#Saving as Vertex properties
V(x2)$vertex_x2_color_degree<-
  vertex_x2_color_degree[as.numeric(
    cut(V(x2)$x2_color_degree,
      breaks=length(unique(V(x2)$x2_color_degree))))]

set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(x2, es=E(x2), names=F)[,1]

# Fixing ego
minC <- rep(-Inf, vcount(x2))
maxC <- rep(Inf, vcount(x2))
minC[1] <- maxC[1] <- 0
co <- layout_with_fr(x2, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(x2)$weight)

#Plotting
plot(x2,
  layout=co,
  edge.color=V(x2)$vertex_x2_color_degree[edge.start],
  edge.arrow.size=closeness(x2, weights = E(x2)$x2, mode="in"),
  edge.width=E(x2)$weight/mean(E(x2)$weight),
  edge.curved = TRUE,
  vertex.color=V(x2)$vertex_x2_color_degree,
  vertex.size=closeness(x2, weights = E(x2)$x2, mode="in")*10^5,
  vertex.frame.color="black",
  vertex.label.color="black",
  vertex.label=get.vertex.attribute(x2,"LABEL_COR"),
  vertex.label.cex=(closeness(x2, weights = E(x2)$x2, mode="in")+10^-5)*2000,
  vertex.label.dist=0,
  rescale=F,
  xlim=range(co[,1]),
  ylim=range(co[,2]))
```

```

    )
axis(1)
axis(2)

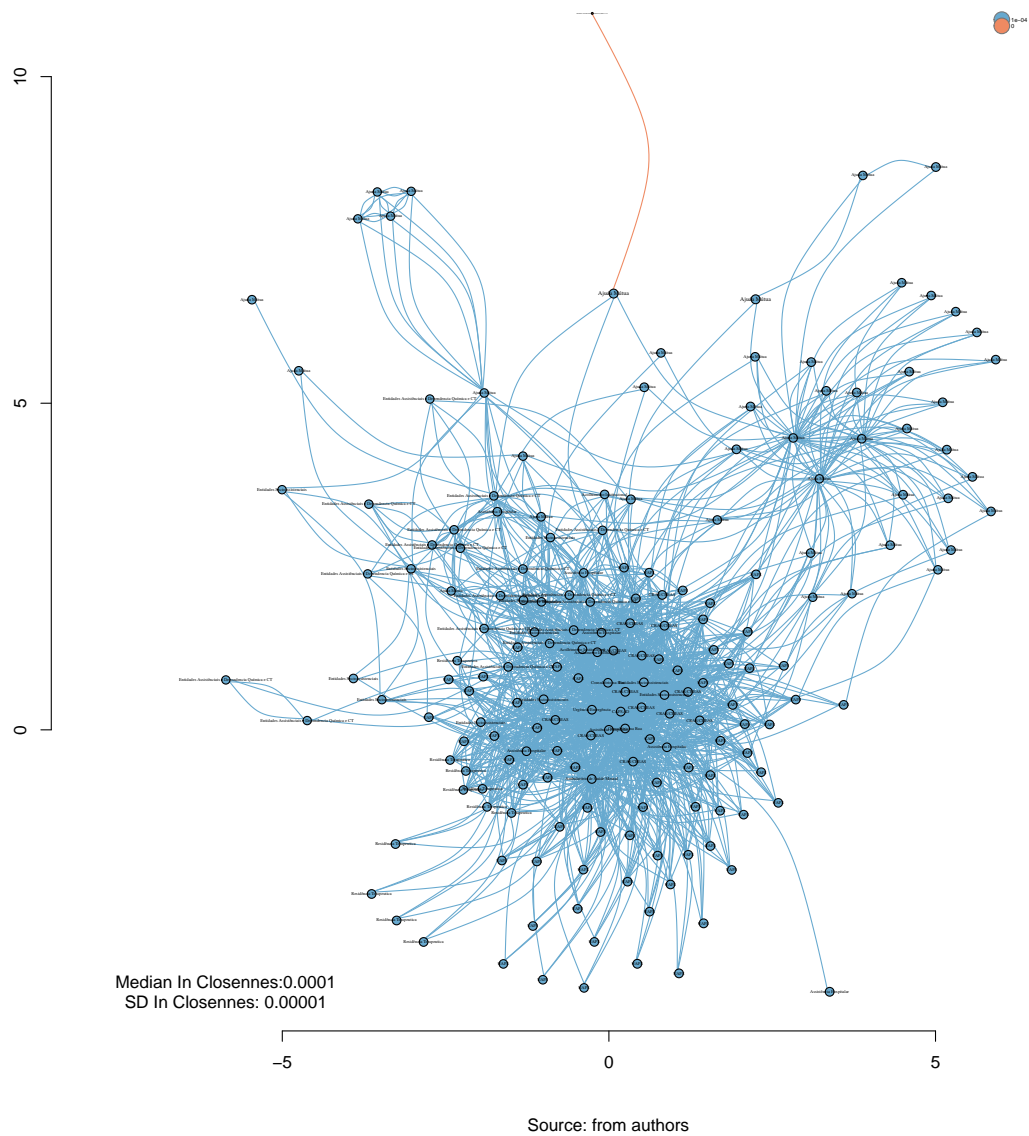
#Solving Problems with legend rendering
a<-V(x2)$x2_color_degree
b<-V(x2)$vertex_x2_color_degree
c<-table(a,b)
d<-as.data.frame(c)
e<-subset(d, d$Freq>0)
e<-e[order(e$a,decreasing=T),]
f<-t(e$a)
g<-t(e$b)

#Adding Legend
legend(x=range(co[,1])[2],
      y=range(co[,2])[2],
      legend=as.character(f),
      pch=21,
      col = "#777777",
      pt.bg=as.character(g),
      pt.cex=2,
      bty="n",
      ncol=1,
      lty=1,
      cex = .3)

#Adding Title
title("Network Closeness Degree Sized and Colored In - 1.2 - ALGUM RELACIONAMENTO (x2)", sub = "Source")
text(
  x=range(co[,1])[1],
  y=range(co[,2])[1],
  labels = sprintf(
    "Median In Closennes: %.4f\nSD In Closennes: %.5f",
    median(closeness(x2, mode="in", weights = E(x2)$x2)),
    sd(closeness(x2, mode="in", weights = E(x2)$x2))
  )
)

```

## Network Closeness Degree Sized and Colored In – 1.2 – ALGUM RELACIONAMENTO (x2)



### 3.2.1 Closeness Non-normalized - OUT

```
summary(x2_outcloseness)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.000029 0.001233 0.001538 0.001234 0.001590 0.002105
```

```
sd(x2_outcloseness)
```

```
## [1] 0.0006349249
```

### 3.3 Network Plotting Based On Non-normalized Closeness - OUT

```
V(x2)$outcloseness<-closeness(x2, weights = E(x2)$x2, mode="out")

#Get Variable
V(x2)$x2_color_degree<-round(V(x2)$outcloseness,4)

#Creating brewer palette
vertex_x2_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
    V(x2)$x2_color_degree)), "RdBu"))(
    length(unique(V(x2)$x2_color_degree)))

#Saving as Vertex properties
V(x2)$vertex_x2_color_degree<-
  vertex_x2_color_degree[as.numeric(
    cut(V(x2)$x2_color_degree,
      breaks=length(unique(V(x2)$x2_color_degree))))]

set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(x2, es=E(x2), names=F)[,1]

# Fixing ego
minC <- rep(-Inf, vcount(x2))
maxC <- rep(Inf, vcount(x2))
minC[1] <- maxC[1] <- 0
co <- layout_with_fr(x2, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(x2)$weight)

#Plotting
plot(x2,
  layout=co,
  #edge.color=V(x2)$vertex_x2_color_degree[edge.start],
  edge.arrow.size=closeness(x2, weights = E(x2)$x2, mode="out"),
  edge.width=E(x2)$weight/2*mean(E(x2)$weight),
  edge.curved = TRUE,
  vertex.color=V(x2)$vertex_x2_color_degree,
  vertex.size=closeness(x2, weights = E(x2)$x2, mode="out")*10^4,
  vertex.frame.color="white",
  vertex.label.color="black",
  vertex.label=get.vertex.attribute(x2,"LABEL_COR"),
  vertex.label.cex=closeness(x2, weights = E(x2)$x2, mode="out")*200,
  vertex.label.dist=0,
  rescale=F,
  xlim=range(co[,1]),
  ylim=range(co[,2])
)
axis(1)
axis(2)

#Solving Problems with legend rendering
```

```

a<-V(x2)$x2_color_degree
b<-V(x2)$vertex_x2_color_degree
c<-table(a,b)
d<-as.data.frame(c)
e<-subset(d, d$Freq>0)
e<-e[order(e$a,decreasing=T),]
f<-t(e$a)
g<-t(e$b)

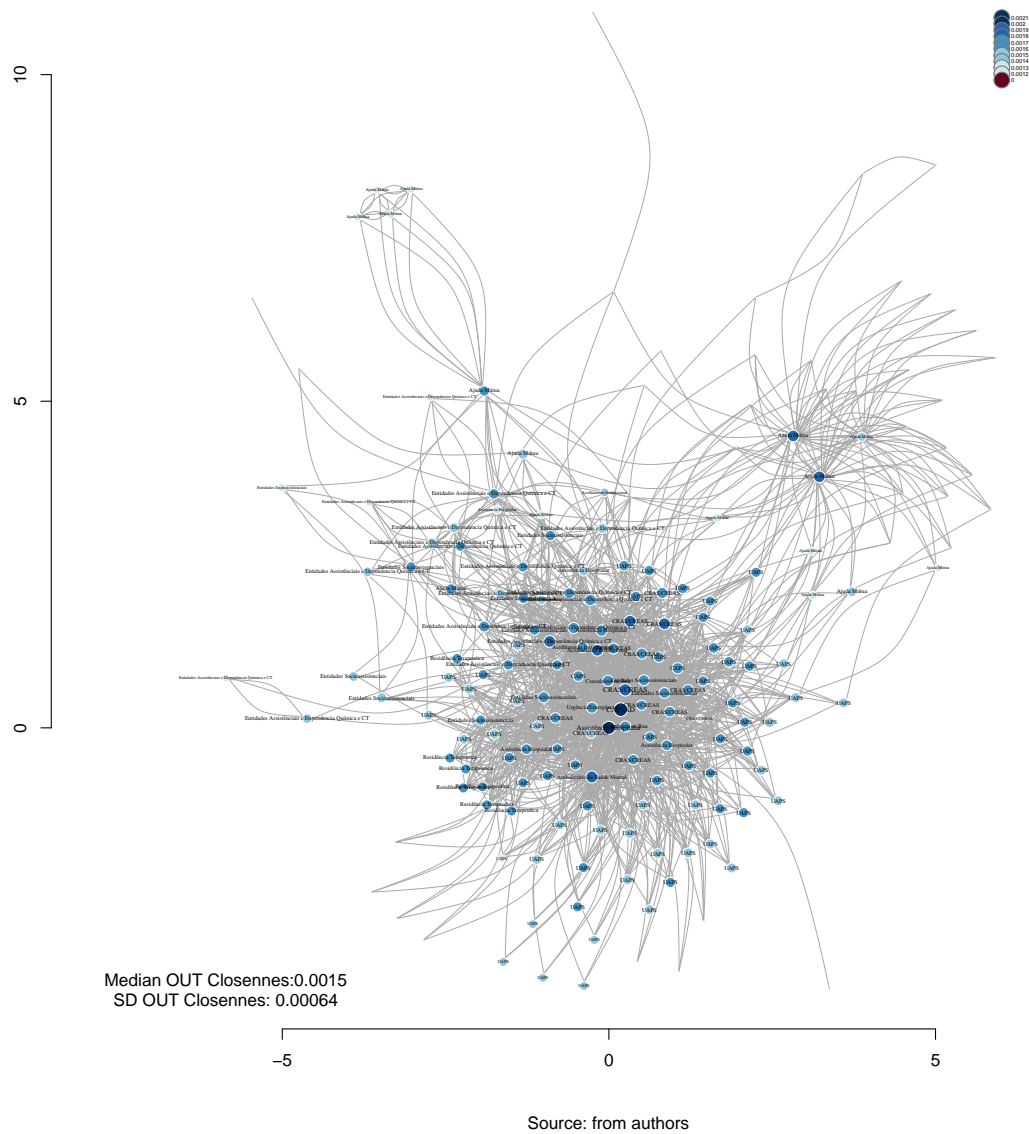
#Adding Legend
legend(x=range(co[,1])[2],
      y=range(co[,2])[2],
      legend=as.character(f),
      pch=21,
      col = "#777777",
      pt.bg=as.character(g),
      pt.cex=2,
      bty="n",
      ncol=1,
      lty=1,
      cex = .3)

#Adding Title
title("Network Closeness Degree Sized and Colored OUT - 1.2 - ALGUM RELACIONAMENTO (x2)", sub = "Source")
text(
  x=range(co[,1])[1],
  y=range(co[,2])[1],
  labels = sprintf(
    "Median OUT Closennes: %.4f\nSD OUT Closennes: %.5f",
    median(closeness(x2, mode="out", weights = E(x2)$x2)),
    sd(closeness(x2, mode="out", weights = E(x2)$x2))
  )
)

```



### Network Closeness Degree Sized and Colored OUT – 1.2 – ALGUM RELACIONAMENTO (x2)



#### 3.3.1 Closeness Non-normalized - ALL

```
summary(x2_totalcloseness)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.001527 0.002265 0.002445 0.002404 0.002561 0.003968
```

```
sd(x2_totalcloseness)
```

```
## [1] 0.0003242546
```

### 3.4 Network Plotting Based On Non-normalized Closeness - ALL

```
V(x2)$allcloseness<-closeness(x2, weights = E(x2)$x2, mode="all")

#Get Variable
V(x2)$x2_color_degree<-round(V(x2)$allcloseness,4)

#Creating brewer palette
vertex_x2_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
    V(x2)$x2_color_degree)), "RdBu"))(
    length(unique(V(x2)$x2_color_degree)))

#Saving as Vertex properties
V(x2)$vertex_x2_color_degree<-
  vertex_x2_color_degree[as.numeric(
    cut(V(x2)$x2_color_degree,
      breaks=length(unique(V(x2)$x2_color_degree))))]

set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(x2, es=E(x2), names=F)[,1]

# Fixing ego
minC <- rep(-Inf, vcount(x2))
maxC <- rep(Inf, vcount(x2))
minC[1] <- maxC[1] <- 0
co <- layout_with_fr(x2, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(x2)$weight)

#Plotting
plot(x2,
  layout=co,
  #edge.color=V(x2)$vertex_x2_color_degree[edge.start],
  edge.arrow.size=closeness(x2, weights = E(x2)$x2, mode="all"),
  edge.width=E(x2)$weight/2*mean(E(x2)$weight),
  edge.curved = TRUE,
  vertex.color=V(x2)$vertex_x2_color_degree,
  vertex.size=closeness(x2, weights = E(x2)$x2, mode="all")*10^4,
  vertex.frame.color="white",
  vertex.label.color="black",
  vertex.label=get.vertex.attribute(x2,"LABEL_COR"),
  vertex.label.cex=closeness(x2, weights = E(x2)$x2, mode="all")*200,
  vertex.label.dist=0,
  rescale=F,
  xlim=range(co[,1]),
  ylim=range(co[,2])
)
axis(1)
axis(2)

#Solving Problems with legend rendering
```

```

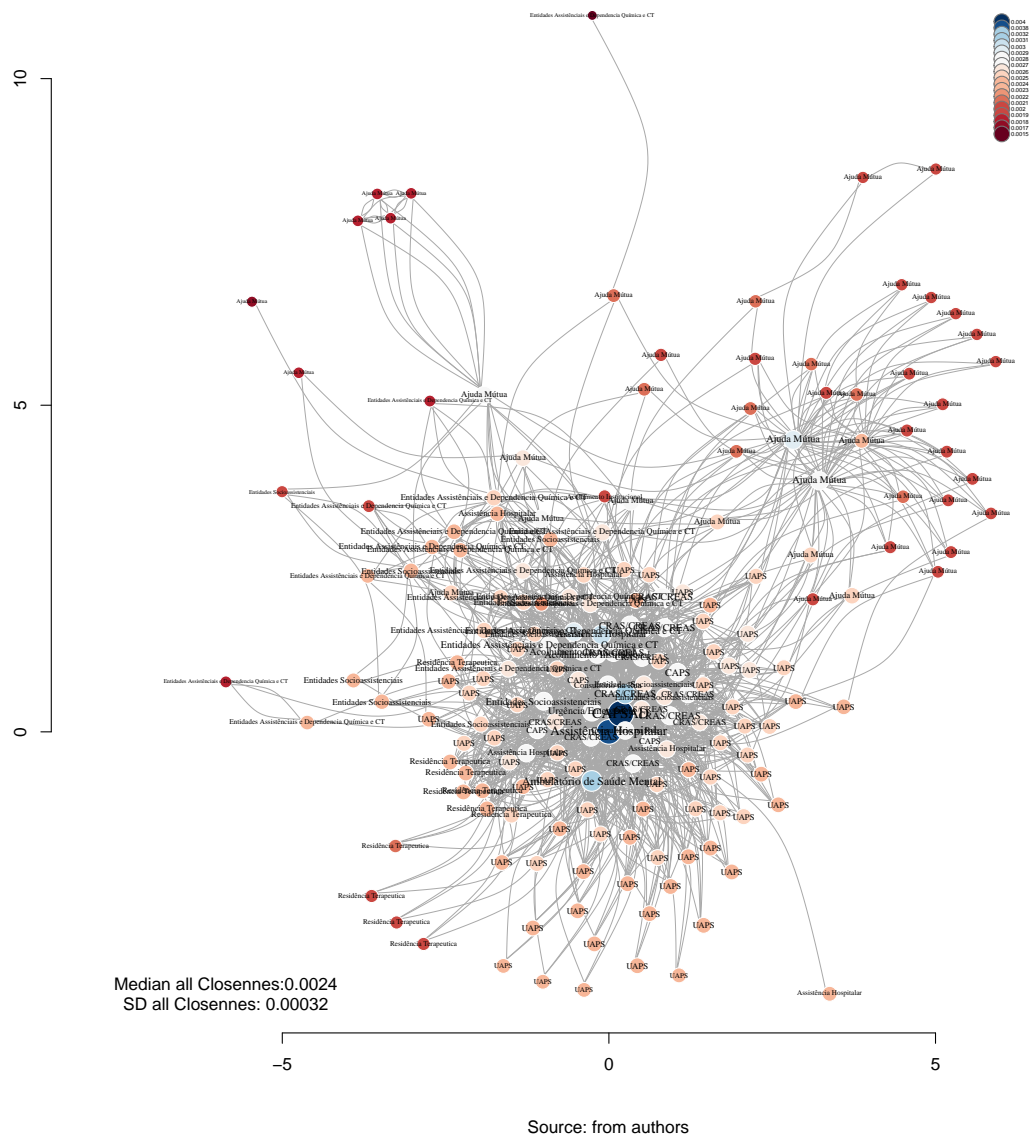
a<-V(x2)$x2_color_degree
b<-V(x2)$vertex_x2_color_degree
c<-table(a,b)
d<-as.data.frame(c)
e<-subset(d, d$Freq>0)
e<-e[order(e$a,decreasing=T),]
f<-t(e$a)
g<-t(e$b)

#Adding Legend
legend(x=range(co[,1])[2],
      y=range(co[,2])[2],
      legend=as.character(f),
      pch=21,
      col = "#777777",
      pt.bg=as.character(g),
      pt.cex=2,
      bty="n",
      ncol=1,
      lty=1,
      cex = .3)

#Adding Title
title("Network Closeness Degree Sized and Colored all - 1.2 - ALGUM RELACIONAMENTO (x2)", sub = "Source")
text(
  x=range(co[,1])[1],
  y=range(co[,2])[1],
  labels = sprintf(
    "Median all Closennes: %.4f\nSD all Closennes: %.5f",
    median(closeness(x2, mode="all", weights = E(x2)$x2)),
    sd(closeness(x2, mode="all", weights = E(x2)$x2))
  )
)

```

## Network Closeness Degree Sized and Colored all – 1.2 – ALGUM RELACIONAMENTO (x2)



### 3.5 Closeness Normalized

#### 3.5.1 Saving to Igraph object

```
V(x2)$incloseness_n <- closeness(x2, mode = "in", weights = E(x2)$x2, normalized = T) %>% round(10)
V(x2)$outcloseness_n <- closeness(x2, mode = "out", normalized = T, weights = E(x2)$x2) %>% round(6)
V(x2)$totalcloseness_n <- closeness(x2, mode = "total", normalized = T, weights = E(x2)$x2) %>% round(6)
```

### 3.5.2 Saving to Environment

```
x2_incloseness_n<- closeness(x2, mode = "in", normalized = T, weights = E(x2)$x2) %>% round(6)
x2_outcloseness_n<- closeness(x2, mode = "out", normalized = T, weights = E(x2)$x2) %>% round(6)
x2_totalcloseness_n<- closeness(x2, mode = "total", normalized = T, weights = E(x2)$x2) %>% round(6)
```

### 3.5.3 Closeness Normalized - IN

```
summary(x2_incloseness_n)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.005348 0.024300 0.024460 0.024310 0.024510 0.025310
```

```
sd(x2_incloseness_n)
```

```
## [1] 0.001421472
```

## 3.6 Network Plotting Based On Normalized Closeness - IN

```
V(x2)$incloseness_n<-closeness(x2, weights = E(x2)$x2, mode="in", normalized = T)

#Get Variable
V(x2)$x2_color_degree<-round(V(x2)$incloseness_n,4)

#Creating brewer palette
vertex_x2_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
    V(x2)$x2_color_degree)), "RdBu"))(
    length(unique(V(x2)$x2_color_degree)))

#Saving as Vertex properties
V(x2)$vertex_x2_color_degree<-
  vertex_x2_color_degree[as.numeric(
    cut(V(x2)$x2_color_degree,
      breaks=length(unique(V(x2)$x2_color_degree))))]

set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(x2, es=E(x2), names=F)[,1]

# Fixing ego
minC <- rep(-Inf, vcount(x2))
maxC <- rep(Inf, vcount(x2))
minC[1] <- maxC[1] <- 0
co <- layout_with_fr(x2, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(x2)$weight)
```

```

#Plotting
plot(x2,
     layout=co,
     #edge.color=V(x2)$vertex_x2_color_degree[edge.start],
     edge.arrow.size=closeness(x2, weights = E(x2)$x2, mode="in",normalized = T),
     edge.width=E(x2)$weight/10*mean(E(x2)$weight),
     edge.curved = TRUE,
     vertex.color=V(x2)$vertex_x2_color_degree,
     vertex.size=(closeness(x2, weights = E(x2)$x2, mode="in",normalized = T))*1000,
     vertex.frame.color="black",
     vertex.label.color="black",
     vertex.label=get.vertex.attribute(x2,"LABEL_COR"),
     vertex.label.cex=closeness(x2, weights = E(x2)$x2, mode="in",normalized = T)*10,
     vertex.label.dist=0,
     rescale=F,
     xlim=range(co[,1]),
     ylim=range(co[,2])
)
axis(1)
axis(2)

#Solving Problems with legend rendering
a<-V(x2)$x2_color_degree
b<-V(x2)$vertex_x2_color_degree
c<-table(a,b)
d<-as.data.frame(c)
e<-subset(d, d$Freq>0)
e<-e[order(e$a,decreasing=T),]
f<-t(e$a)
g<-t(e$b)

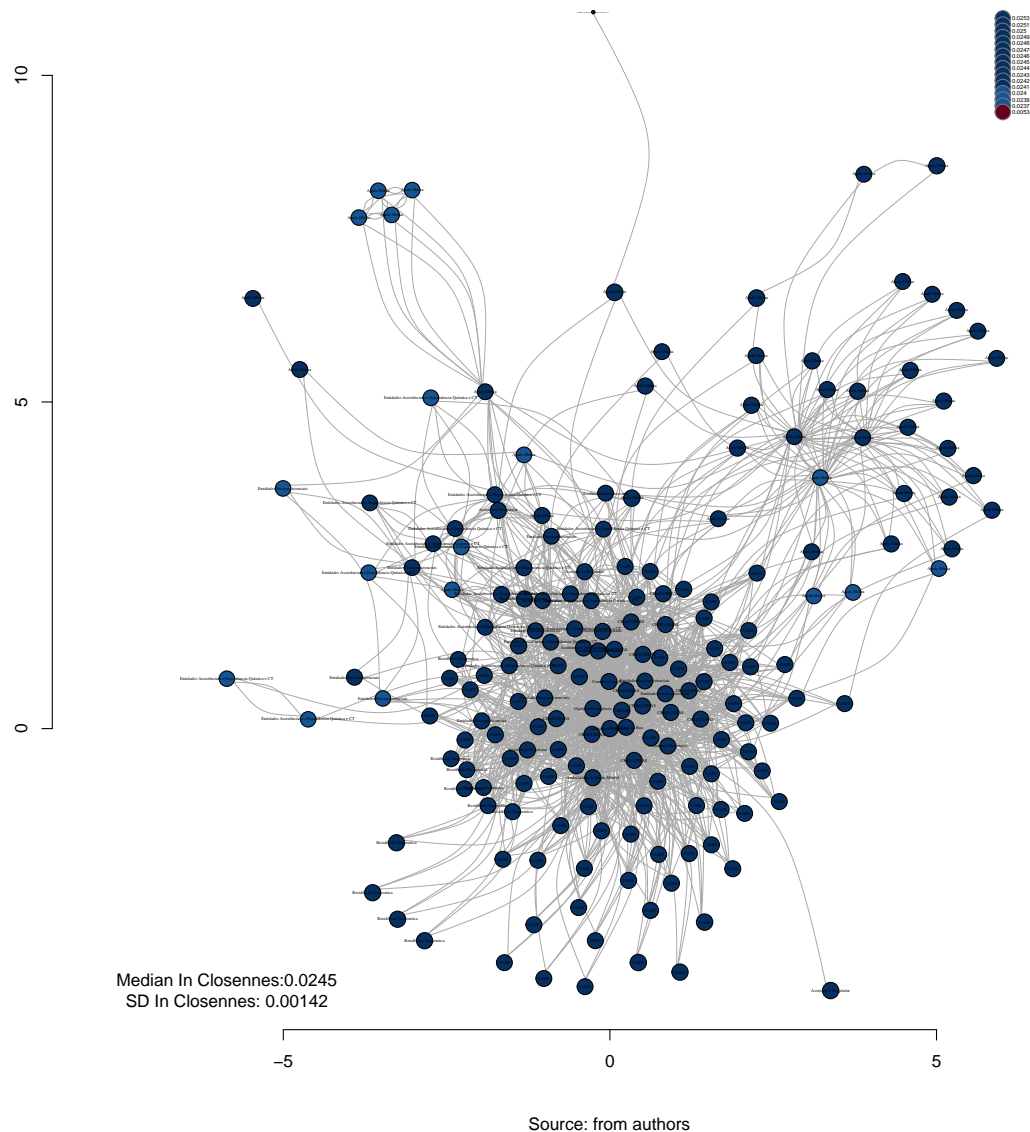
#Adding Legend
legend(x=range(co[,1])[2],
      y=range(co[,2])[2],
      legend=as.character(f),
      pch=21,
      col = "#777777",
      pt.bg=as.character(g),
      pt.cex=2,
      bty="n",
      ncol=1,
      lty=1,
      cex = .3)

#Adding Title
title("Network Closeness Degree Sized Normalized In - 1.2 - ALGUM RELACIONAMENTO (x2)", sub = "Source
text(
  x=range(co[,1])[1],
  y=range(co[,2])[1],
  labels = sprintf(
    "Median In Closennes:%.4f\nSD In Closennes: %.5f",
    median(closeness(x2, mode="in", weights = E(x2)$x2, normalized = T)),

```

```
sd(closeness(x2, mode="in", weights = E(x2)$x2, normalized = T))
)
)
```

Network Closeness Degree Sized Normalized In – 1.2 – ALGUM RELACIONAMENTO (x2)



###Closeness Normalized - OUT

```
summary(x2_outcloseness_n)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.005348 0.229300 0.286200 0.229600 0.295700 0.391600
```

```
sd(x2_outcloseness_n)
```

```
## [1] 0.1181212
```

### 3.7 Network Plotting Based On Normalized Closeness - OUT

```
V(x2)$outcloseness_n<-closeness(x2, weights = E(x2)$x2, mode="out", normalized = T)

#Get Variable
V(x2)$x2_color_degree<-round(V(x2)$outcloseness_n,2)

#Creating brewer palette
vertex_x2_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
    V(x2)$x2_color_degree)), "RdBu"))(
    length(unique(V(x2)$x2_color_degree)))

#Saving as Vertex properties
V(x2)$vertex_x2_color_degree<-
  vertex_x2_color_degree[as.numeric(
    cut(V(x2)$x2_color_degree,
      breaks=length(unique(V(x2)$x2_color_degree))))]

set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(x2, es=E(x2), names=F)[,1]

# Fixing ego
minC <- rep(-Inf, vcount(x2))
maxC <- rep(Inf, vcount(x2))
minC[1] <- maxC[1] <- 0
co <- layout_with_fr(x2, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(x2)$weight)

#Plotting
plot(x2,
  layout=co,
  #edge.color=V(x2)$vertex_x2_color_degree[edge.start],
  edge.arrow.size=closeness(x2, weights = E(x2)$x2, mode="out",normalized = T),
  edge.width=E(x2)$weight/10*mean(E(x2)$weight),
  edge.curved = TRUE,
  vertex.color=V(x2)$vertex_x2_color_degree,
  vertex.size=(closeness(x2, weights = E(x2)$x2, mode="out",normalized = T))*100,
  vertex.frame.color="black",
  vertex.label.color="black",
  vertex.label=get.vertex.attribute(x2,"LABEL_COR"),
  vertex.label.cex=closeness(x2, weights = E(x2)$x2, mode="out",normalized = T)*1.5,
  vertex.label.dist=0,
  rescale=F,
  xlim=range(co[,1]),
  ylim=range(co[,2])
)
axis(1)
axis(2)

#Solving Problems with legend rendering
```



```

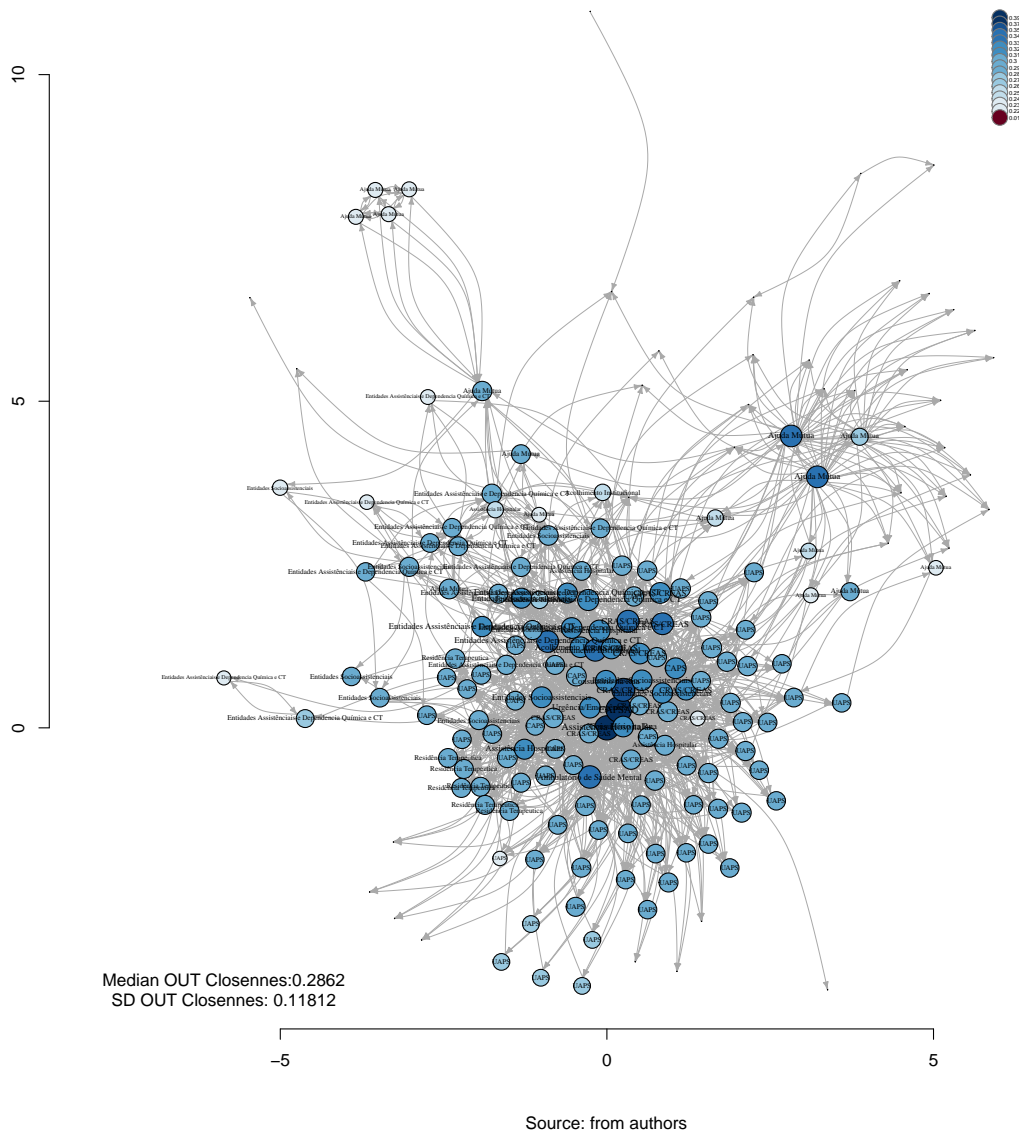
a<-V(x2)$x2_color_degree
b<-V(x2)$vertex_x2_color_degree
c<-table(a,b)
d<-as.data.frame(c)
e<-subset(d, d$Freq>0)
e<-e[order(e$a,decreasing=T),]
f<-t(e$a)
g<-t(e$b)

#Adding Legend
legend(x=range(co[,1])[2],
      y=range(co[,2])[2],
      legend=as.character(f),
      pch=21,
      col = "#777777",
      pt.bg=as.character(g),
      pt.cex=2,
      bty="n",
      ncol=1,
      lty=1,
      cex = .3)

#Adding Title
title("Network Closeness Degree Sized Normalized OUT - 1.2 - ALGUM RELACIONAMENTO (x2)", sub = "Source")
text(
  x=range(co[,1])[1],
  y=range(co[,2])[1],
  labels = sprintf(
    "Median OUT Closennes: %.4f\nSD OUT Closennes: %.5f",
    median(closeness(x2, mode="out", weights = E(x2)$x2, normalized = T)),
    sd(closeness(x2, mode="out", weights = E(x2)$x2, normalized = T))
  )
)

```

### Network Closeness Degree Sized Normalized OUT – 1.2 – ALGUM RELACIONAMENTO (x2)



#### 3.7.1 Closeness Normalized - ALL

```
summary(x2_totalcloseness_n)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  0.2840  0.4213  0.4548  0.4472  0.4763  0.7381
```

```
sd(x2_totalcloseness_n)
```

```
## [1] 0.06030622
```

### 3.8 Network Plotting Based On Normalized Closeness - ALL

```
V(x2)$allcloseness_n<-closeness(x2, weights = E(x2)$x2, mode="all", normalized = T)

#Get Variable
V(x2)$x2_color_degree<-round(V(x2)$allcloseness_n,2)

#Creating brewer palette
vertex_x2_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
    V(x2)$x2_color_degree)), "RdBu"))(
    length(unique(V(x2)$x2_color_degree)))

#Saving as Vertex properties
V(x2)$vertex_x2_color_degree<-
  vertex_x2_color_degree[as.numeric(
    cut(V(x2)$x2_color_degree,
      breaks=length(unique(V(x2)$x2_color_degree))))]

set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(x2, es=E(x2), names=F)[,1]

# Fixing ego
minC <- rep(-Inf, vcount(x2))
maxC <- rep(Inf, vcount(x2))
minC[1] <- maxC[1] <- 0
co <- layout_with_fr(x2, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(x2)$weight)

#Plotting
plot(x2,
  layout=co,
  #edge.color=V(x2)$vertex_x2_color_degree[edge.start],
  edge.arrow.size=closeness(x2, weights = E(x2)$x2, mode="all",normalized = T),
  edge.width=E(x2)$weight/10*mean(E(x2)$weight),
  edge.curved = TRUE,
  vertex.color=V(x2)$vertex_x2_color_degree,
  vertex.size=(closeness(x2, weights = E(x2)$x2, mode="all",normalized = T))*100,
  vertex.frame.color="black",
  vertex.label.color="black",
  vertex.label=get.vertex.attribute(x2,"LABEL_COR"),
  vertex.label.cex=closeness(x2, weights = E(x2)$x2, mode="all",normalized = T)*1.5,
  vertex.label.dist=0,
  rescale=F,
  xlim=range(co[,1]),
  ylim=range(co[,2])
)
axis(1)
axis(2)

#Solving Problems with legend rendering
```

```

a<-V(x2)$x2_color_degree
b<-V(x2)$vertex_x2_color_degree
c<-table(a,b)
d<-as.data.frame(c)
e<-subset(d, d$Freq>0)
e<-e[order(e$a,decreasing=T),]
f<-t(e$a)
g<-t(e$b)

#Adding Legend
legend(x=range(co[,1])[2],
      y=range(co[,2])[2],
      legend=as.character(f),
      pch=21,
      col = "#777777",
      pt.bg=as.character(g),
      pt.cex=2,
      bty="n",
      ncol=1,
      lty=1,
      cex = .3)

#Adding Title
title("Network Closeness Degree Sized Normalized ALL - 1.2 - ALGUM RELACIONAMENTO (x2)", sub = "Source")
text(
  x=range(co[,1])[1],
  y=range(co[,2])[1],
  labels = sprintf(
    "Median ALL Closennes: %.4f\nSD ALL Closennes: %.5f",
    median(closeness(x2, mode="all", weights = E(x2)$x2, normalized = T)),
    sd(closeness(x2, mode="all", weights = E(x2)$x2, normalized = T))
  )
)

```

## Network Closeness Degree Sized Normalized ALL – 1.2 – ALGUM RELACIONAMENTO (x2)



### 3.9 Closeness Normalized

#### 3.9.1 Saving to Igraph object

```
V(x2)$incloseness_n <- closeness(x2, weights = E(x2)$x2, mode = "in", normalized = T) %>% round(6)
V(x2)$outcloseness_n <- closeness(x2, weights = E(x2)$x2, mode = "out", normalized = T) %>% round(6)
V(x2)$totalcloseness_n <- closeness(x2, weights = E(x2)$x2, mode = "total", normalized = T) %>% round(6)
```

### 3.10 Centralization Closeness

```
V(x2)$x2_centr_closeness<- centralization.closeness(x2)$res
x2_centr_closeness<- centralization.closeness(x2)$res
x2_centr_closeness_all<- centralization.closeness(x2)
```

#### 3.10.1 Centralization

```
x2_centr_closeness_all$centralization
```

```
## [1] 0.1637188
```

#### 3.10.2 Theoretical Max

```
x2_centr_closeness_all$theoretical_max
```

```
## [1] 185.0053
```

### 3.11 Network Plotting Based On Centralization Closeness

```
V(x2)$x2_centr_closeness<- centralization.closeness(x2)$res

#Get Variable
V(x2)$x2_color_degree<-round(V(x2)$x2_centr_closeness,2)

#Creating brewer palette
vertex_x2_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
    V(x2)$x2_color_degree)), "Spectral"))(
    length(unique(V(x2)$x2_color_degree)))

#Saving as Vertex properties
V(x2)$vertex_x2_color_degree<-
  vertex_x2_color_degree[as.numeric(
    cut(V(x2)$x2_color_degree,
      breaks=length(unique(V(x2)$x2_color_degree))))]

set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(x2, es=E(x2), names=F)[,1]

# Fixing ego
minC <- rep(-Inf, vcount(x2))
maxC <- rep(Inf, vcount(x2))
minC[1] <- maxC[1] <- 0
co <- layout_with_fr(x2, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(x2)$weight)
```

```
#Plotting
```

```
plot(x2,  
     layout=co,  
     edge.color=V(x2)$vertex_x2_color_degree[edge.start],  
     edge.arrow.size=centralization.closeness(x2)$res,  
     edge.width=E(x2)$weight/10*mean(E(x2)$weight),  
     edge.curved = TRUE,  
     vertex.color=V(x2)$vertex_x2_color_degree,  
     vertex.size=centralization.closeness(x2)$res*100,  
     vertex.frame.color="black",  
     vertex.label.color="black",  
     vertex.label=get.vertex.attribute(x2,"LABEL_COR"),  
     vertex.label.cex=centralization.closeness(x2)$res,  
     vertex.label.dist=0,  
     rescale=F,  
     xlim=range(co[,1]),  
     ylim=range(co[,2])  
 )  
axis(1)  
axis(2)
```

```
#Solving Problems with legend rendering
```

```
a<-V(x2)$x2_color_degree  
b<-V(x2)$vertex_x2_color_degree  
c<-table(a,b)  
d<-as.data.frame(c)  
e<-subset(d, d$Freq>0)  
e<-e[order(e$a,decreasing=T),]  
f<-t(e$a)  
g<-t(e$b)
```

```
#Adding Legend
```

```
legend(x=range(co[,1])[2],  
       y=range(co[,2])[2],  
       legend=as.character(f),  
       pch=21,  
       col = "#777777",  
       pt.bg=as.character(g),  
       pt.cex=2,  
       bty="n",  
       ncol=1,  
       lty=1,  
       cex = .3)
```

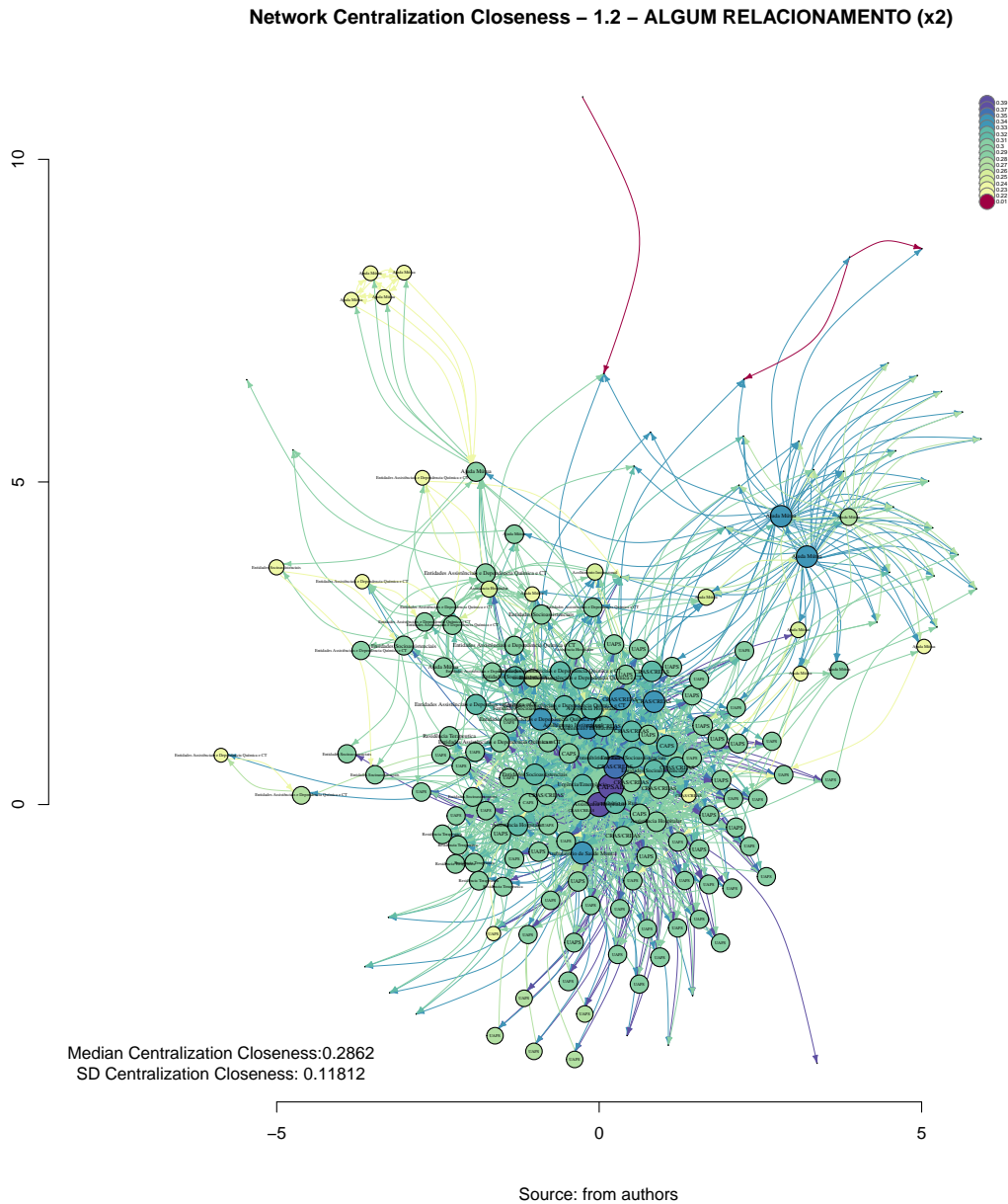
```
#Adding Title
```

```
title("Network Centralization Closeness - 1.2 - ALGUM RELACIONAMENTO (x2)", sub = "Source: from author")  
text(  
     x=range(co[,1])[1],  
     y=range(co[,2])[1],  
     labels = sprintf(  
         "Median Centralization Closeness:%.4f\nSD Centralization Closeness: %.5f",
```

```

median(centralization.closeness(x2)$res),
sd(centralization.closeness(x2)$res)
)

```



## 4 Closeness Dinamic Table

### 4.1 Getting Closeness Measures



```

x2_incloseness<- closeness(x2, weights = E(x2)$x2, mode = "in") %>% round(6)
x2_outcloseness<- closeness(x2, weights = E(x2)$x2, mode = "out") %>% round(6)
x2_totalcloseness<- closeness(x2, weights = E(x2)$x2, mode = "total") %>% round(6)
x2_incloseness_n<- closeness(x2,weights = E(x2)$x2, mode = "in", normalized = T) %>% round(6)
x2_outcloseness_n<- closeness(x2,weights = E(x2)$x2, mode = "out", normalized = T) %>% round(6)
x2_totalcloseness_n<- closeness(x2,weights = E(x2)$x2, mode = "total", normalized = T) %>% round(6)
x2_centr_closeness <- centralization.closeness(x2)$res %>% round(6)

```

## 4.2 Creating a datagram of measures

```

x2_df_closeness <- data.frame(
  x2_incloseness,
  x2_outcloseness,
  x2_totalcloseness,
  x2_incloseness_n,
  x2_outcloseness_n,
  x2_totalcloseness_n,
  x2_centr_closeness) %>% round(6)

#Adding type
x2_df_closeness <-cbind(x2_df_closeness, V(x2)$LABEL_COR)

#Adding names
names(x2_df_closeness) <- c("In Closeness", "Out Closeness", "Total Closeness","In Closeness Normalized", "Out Closeness Normalized", "Total Closeness Normalized")

#Ordering Variables
x2_df_closeness<-x2_df_closeness[c("Type","In Closeness", "Out Closeness", "Total Closeness","In Closeness Normalized", "Out Closeness Normalized", "Total Closeness Normalized")]

```

## 4.3 General tabel - DT

```

datatable(x2_df_closeness, filter = 'top')

```

Show 10 entries

Search:

	Type	In Closeness	Out Closeness	Total Closeness	In Closeness Normalized	Out Closeness Normalized	Total Closeness Normalized	Centralization Closeness
	<input type="text" value="All"/>	<input type="text" value="All"/>	<input type="text" value="All"/>	<input type="text" value="All"/>	<input type="text" value="All"/>	<input type="text" value="All"/>	<input type="text" value="All"/>	<input type="text" value="All"/>
ASS_HOS_ Hospital de Pronto Socorro – HPS	Assistência Hospitalar	0.000133	0.001996	0.003774	0.024714	0.371257	0.701887	0.371257
AMB_SAM_ Centro de Atenção à Saúde Mental (CASM)	Ambulatório de Saúde Mental	0.000132	0.001845	0.003165	0.024561	0.343173	0.588608	0.343173
CAPS_AD	CAPSAD	0.000134	0.002105	0.003968	0.024856	0.391579	0.738095	0.391579
CRAS_AS_ CRAS Sudeste Costa Carvalho	CRAS/CREAS	0.000132	0.001792	0.002793	0.024506	0.333333	0.519553	0.333333
CRE_SOC_ CREAS Infância e Juventude	CRAS/CREAS	0.000132	0.001631	0.00277	0.024529	0.303426	0.515235	0.303426
CRE_SOC_ CREAS Norte	CRAS/CREAS	0.000132	0.001592	0.002695	0.024509	0.296178	0.501348	0.296178
ASS_HOS_ Serviço de Controle e Prevenção e Tratamento do Tabagismo (SECOPTT)	Assistência Hospitalar	0.000131	0.001634	0.002618	0.024442	0.303922	0.486911	0.303922
EA_DQCT_ Centro de Recuperação Resgatando Vidas (Escritório)	Entidades Assistenciais e Dependencia Química e CT	0.000132	0.001709	0.002857	0.024499	0.317949	0.531429	0.317949
EA_DQCT_ Comunidade Terapêutica Geração de Adoradores – CTGA	Entidades Assistenciais e Dependencia Química e CT	0.000129	0.001669	0.002532	0.024081	0.310518	0.470886	0.310518
EA_DQCT_ Centro de Recuperação Resgatando Vidas	Entidades Assistenciais e Dependencia Química e CT	0.000131	0.001548	0.0025	0.024442	0.287926	0.465	0.287926

Showing 1 to 10 of 187 entries

Previous  2 3 4 5 ... 19 Next

#### 4.4 Aggregating data from previous table - mean

```
aggdata_mean <-aggregate(x2_df_closeness, by=list(x2_df_closeness$Type), FUN=mean, na.rm=TRUE)

names(aggdata_mean) <- c("Group","Type","In Closeness(M)", "Out Closeness(M)", "Total Closeness(M)","In Closeness Normalized(M)", "Out Closeness Normalized(M)", "Total Closeness Normalized(M)", "Centralization Closeness")

#Removing Type variable
aggdata_mean<-aggdata_mean[,-c(2)]
```

## 4.5 Aggregating data from previous table - sd

```
aggdata_sd <-aggregate(x2_df_closeness, by=list(x2_df_closeness$Type), FUN=sd, na.rm=TRUE)

names(aggdata_sd) <- c("Group","Type","In Closeness(SD)", "Out Closeness(SD)", "Total Closeness(SD)","In Closeness(M)", "Out Closeness(M)", "Total Closeness(M)", "In Closeness(SD)", "Out Closeness(SD)", "Total Closeness(SD)", "In Closeness(M)", "Out Closeness(M)", "Total Closeness(M)")

#Removing Type variable
aggdata_sd<-aggdata_sd[,-c(2)]

#Merging mean and standart deviation
total_table <- merge(aggdata_mean,aggdata_sd,by="Group")

#Rounding
Group<-total_table[,c(1)] #Keeping group
total_table<-total_table[,-c(1)] %>% round(6) #Rouding
total_table<-cbind(Group,total_table) #Binding togheter

#Organizing Variabels
total_table<-total_table[c("Group","In Closeness(M)", "In Closeness(SD)", "Out Closeness(M)", "Out Closeness(SD)", "Total Closeness(M)", "Total Closeness(SD)", "In Closeness(M)", "In Closeness(SD)", "Out Closeness(M)", "Out Closeness(SD)", "Total Closeness(M)", "Total Closeness(SD)", "Centralization Closeness(M)", "Centralization Closeness(SD)")]
```

## 4.6 Plotting final table with round for Closeness

```
datatable(total_table, filter = 'top')
```

Show 10 entries

Search:

	Group	In Closeness(M)	In Closeness(SD)	Out Closeness(M)	Out Closeness(SD)	Total Closeness(M)	Total Closeness(SD)	In Closeness Normalized(M)	In Closeness Normalized(SD)	Out Closeness Normalized(M)	Out Closeness Normalized(SD)	Total Closeness Normalized(M)	Total Closeness Normalized(SD)	Centralization Closeness(M)	Centralization Closeness(SD)
1	Acolhimento Institucional	0.000131	0.000001	0.001616	0.000231	0.002545	0.000453	0.024385	0.000229	0.300585	0.043014	0.473387	0.08415	0.300585	0.043014
2	Ajuda Mútua	0.000131	0.000002	0.000507	0.000672	0.002116	0.000297	0.024378	0.000401	0.09424	0.125042	0.393559	0.055153	0.09424	0.125042
3	Ambulatório de Saúde Mental	0.000132		0.001845		0.003165		0.024561		0.343173		0.588608		0.343173	
4	Assistência Hospitalar	0.000132	0.000001	0.001417	0.000644	0.002744	0.000506	0.024582	0.000206	0.263607	0.119764	0.510281	0.094089	0.263607	0.119764
5	CAPS	0.000132	0	0.00161	0.000081	0.002667	0.000029	0.024542	0.000026	0.299481	0.015084	0.496044	0.005453	0.299481	0.015084
6	CAPSAD	0.000134		0.002105		0.003968		0.024856		0.391579		0.738095		0.391579	
7	Consultório na Rua	0.000132	0.000001	0.001704	0	0.002688	0.000021	0.024458	0.000013	0.316865	0	0.500014	0.003802	0.316865	0
8	CRAS/CREAS	0.000132	0	0.001644	0.000164	0.002728	0.000119	0.024519	0.000017	0.305785	0.030577	0.507332	0.022116	0.305785	0.030577
9	Entidades Assistenciais e Dependência Química e CT	0.000124	0.000023	0.001454	0.000385	0.002371	0.00035	0.023121	0.004311	0.270431	0.071675	0.440936	0.065121	0.270431	0.071675
10	Entidades Socioassistenciais	0.00013	0.000002	0.001577	0.000122	0.002436	0.000221	0.024236	0.000278	0.293281	0.022725	0.453105	0.041074	0.293281	0.022725

Showing 1 to 10 of 13 entries

Previous 1 2 Next

## 5 Saving objects with new variables and changes

```
save.image("~/SNArRDJF/Robject/3_closeness_x2.RData")
```