# SNA Closeness 1.1 - REDE COMPLETA

# (full\_no\_zero)

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#### 5 Saving objects with new variables and changes

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SNA Descritive 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 1.1 - REDE COMPLETA (full no zero)

# 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/Robject/2_degree_full_no_zero.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 duble edges

```
full_no_zero<-simplify(full_no_zero) #Simplify</pre>
```

# 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(full_no_zero)$incloseness <- closeness(full_no_zero, mode = "in", weights = E(full_no_zero)$full_no_z
V(full_no_zero)$outcloseness <- closeness(full_no_zero, mode = "out", weights = E(full_no_zero)$full_no
V(full_no_zero)$totalcloseness <- closeness(full_no_zero, mode = "total", weights = E(full_no_zero)$full_no_zero)$full_no_zero
```

#### 3.1.2 Saving to Environment

```
full_no_zero_incloseness<- closeness(full_no_zero, mode = "in", weights = E(full_no_zero)$full_no_zero)
full_no_zero_outcloseness<- closeness(full_no_zero, mode = "out", weights = E(full_no_zero)$full_no_zer
full_no_zero_totalcloseness<- closeness(full_no_zero, mode = "total", weights = E(full_no_zero)$full_no
```

## 3.1.3 Closeness Non-normalized - in

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.0000290 0.0001480 0.0001500 0.0001486 0.0001500 0.0001550
```

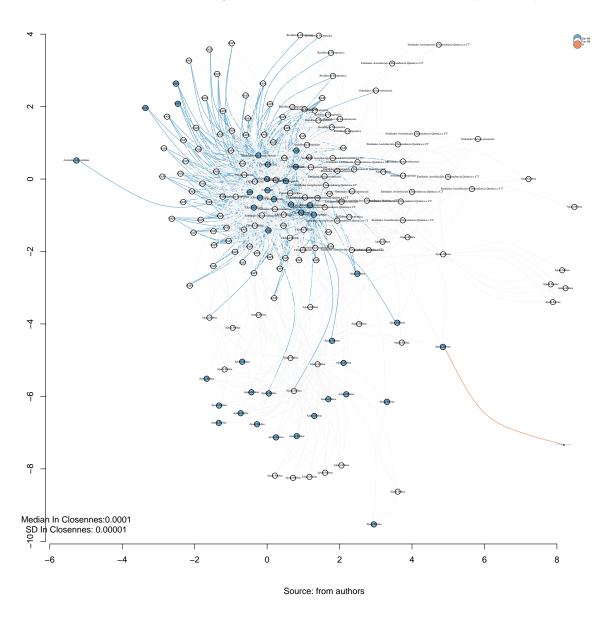
```
sd(full_no_zero_incloseness)
## [1] 9.003136e-06
```

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

```
V(full_no_zero) $incloseness <-closeness (full_no_zero, weights = E(full_no_zero) $full_no_zero, mode="in")
#Get Variable
V(full_no_zero)$full_no_zero_color_degree<-round(V(full_no_zero)$incloseness,4)
#Creating brewer pallette
vertex_full_no_zero_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
          V(full_no_zero)$full_no_zero_color_degree)), "RdBu"))(
            length(unique(V(full_no_zero)$full_no_zero_color_degree)))
#Saving as Vertex properties
V(full_no_zero)$vertex_full_no_zero_color_degree<-
  vertex_full_no_zero_color_degree[as.numeric(
  cut(V(full_no_zero)$full_no_zero_color_degree,
      breaks=length(unique(V(full_no_zero)$full_no_zero_color_degree))))]
set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(full_no_zero, es=E(full_no_zero), names=F)[,1]</pre>
# Fixing eqo
minC <- rep(-Inf, vcount(full_no_zero))</pre>
maxC <- rep(Inf, vcount(full_no_zero))</pre>
minC[1] \leftarrow maxC[1] \leftarrow 0
co <- layout_with_fr(full_no_zero, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(f
#PLotting
plot(full_no_zero,
     layout=co,
     edge.color=V(full_no_zero)$vertex_full_no_zero_color_degree[edge.start],
     edge.arrow.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="in"),
     edge.width=E(full_no_zero)$weight/mean(E(full_no_zero)$weight),
     edge.curved = TRUE,
     vertex.color=V(full_no_zero)$vertex_full_no_zero_color_degree,
     vertex.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="in")*10^5,
     vertex.frame.color="black",
     vertex.label.color="black",
     vertex.label=get.vertex.attribute(full no zero, "LABEL COR"),
     vertex.label.cex=(closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="in")+10^-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(full_no_zero)$full_no_zero_color_degree</pre>
b<-V(full_no_zero)$vertex_full_no_zero_color_degree
c<-table(a,b)
d<-as.data.frame(c)</pre>
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 - 1.1 - REDE COMPLETA (full_no_zero)", sub = "So
  text(
    x=range(co[,1])[1],
    y=range(co[,2])[1],
      labels = sprintf(
             "Median In Closennes: %.4f\nSD In Closennes: %.5f",
             median(closeness(full_no_zero, mode="in", weights = E(full_no_zero)$full_no_zero)),
             sd(closeness(full_no_zero, mode="in", weights = E(full_no_zero)$full_no_zero))
       )
```

## Network Closeness Degree Sized and Colored In – 1.1 – REDE COMPLETA (full\_no\_zero)



## 3.2.1 Closeness Non-normalized - OUT

```
summary(full_no_zero_outcloseness)
```

## Min. 1st Qu. Median Mean 3rd Qu. Max. ## 0.000029 0.001236 0.001543 0.001270 0.001600 0.002110

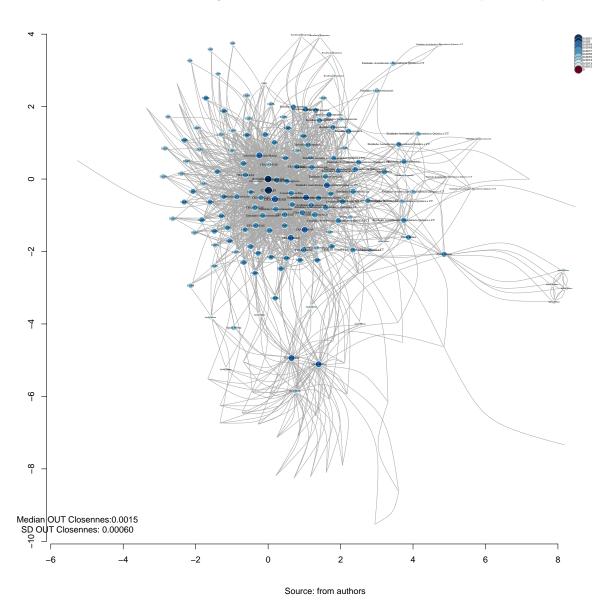
sd(full\_no\_zero\_outcloseness)

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

```
V(full_no_zero) $outcloseness <-closeness (full_no_zero, weights = E(full_no_zero) $full_no_zero, mode="out
#Get Variable
V(full no zero)$full no zero color degree<-round(V(full no zero)$outcloseness,4)
#Creating brewer pallette
vertex_full_no_zero_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
          V(full_no_zero)$full_no_zero_color_degree)), "RdBu"))(
            length(unique(V(full_no_zero)$full_no_zero_color_degree)))
#Saving as Vertex properties
V(full_no_zero)$vertex_full_no_zero_color_degree<-
  vertex_full_no_zero_color_degree[as.numeric(
  cut(V(full_no_zero)$full_no_zero_color_degree,
      breaks=length(unique(V(full_no_zero)$full_no_zero_color_degree))))]
set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(full_no_zero, es=E(full_no_zero), names=F)[,1]</pre>
# Fixing eqo
minC <- rep(-Inf, vcount(full_no_zero))
maxC <- rep(Inf, vcount(full_no_zero))</pre>
minC[1] \leftarrow maxC[1] \leftarrow 0
co <- layout_with_fr(full_no_zero, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(f
#PLotting
plot(full_no_zero,
     layout=co,
     \#edge.color=V(full\_no\_zero)\$vertex\_full\_no\_zero\_color\_degree[edge.start],
     edge.arrow.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="out"),
     edge.width=E(full_no_zero)$weight/2*mean(E(full_no_zero)$weight),
     edge.curved = TRUE,
     vertex.color=V(full_no_zero)$vertex_full_no_zero_color_degree,
     vertex.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="out")*10^4,
     vertex.frame.color="white",
     vertex.label.color="black",
     vertex.label=get.vertex.attribute(full_no_zero, "LABEL_COR"),
     vertex.label.cex=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, 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(full_no_zero)$full_no_zero_color_degree
b<-V(full_no_zero)$vertex_full_no_zero_color_degree
c<-table(a,b)</pre>
d<-as.data.frame(c)</pre>
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 - 1.1 - REDE COMPLETA (full_no_zero)", sub = "S
    x=range(co[,1])[1],
    y=range(co[,2])[1],
      labels = sprintf(
             "Median OUT Closennes: %.4f\nSD OUT Closennes: %.5f",
             median(closeness(full_no_zero, mode="out", weights = E(full_no_zero)$full_no_zero)),
             sd(closeness(full_no_zero, mode="out", weights = E(full_no_zero)$full_no_zero))
```

## Network Closeness Degree Sized and Colored OUT – 1.1 – REDE COMPLETA (full\_no\_zero)



## 3.3.1 Closeness Non-normalized - ALL

```
summary(full_no_zero_totalcloseness)
```

## Min. 1st Qu. Median Mean 3rd Qu. Max. ## 0.001529 0.002278 0.002451 0.002411 0.002564 0.003968

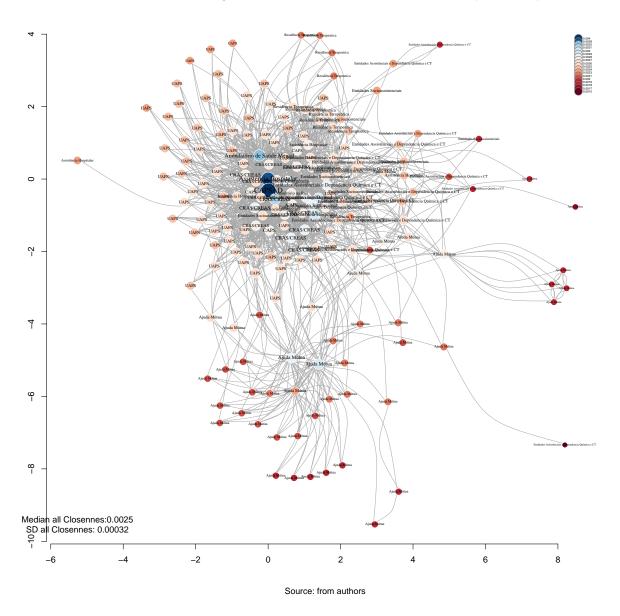
sd(full\_no\_zero\_totalcloseness)

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

```
V(full_no_zero) allcloseness <-closeness (full_no_zero, weights = E(full_no_zero) full_no_zero, mode="all
#Get Variable
V(full no zero)$full no zero color degree<-round(V(full no zero)$allcloseness,4)
#Creating brewer pallette
vertex_full_no_zero_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
          V(full_no_zero)$full_no_zero_color_degree)), "RdBu"))(
            length(unique(V(full_no_zero)$full_no_zero_color_degree)))
#Saving as Vertex properties
V(full_no_zero)$vertex_full_no_zero_color_degree<-
  vertex_full_no_zero_color_degree[as.numeric(
  cut(V(full_no_zero)$full_no_zero_color_degree,
      breaks=length(unique(V(full_no_zero)$full_no_zero_color_degree))))]
set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(full_no_zero, es=E(full_no_zero), names=F)[,1]</pre>
# Fixing eqo
minC <- rep(-Inf, vcount(full_no_zero))
maxC <- rep(Inf, vcount(full_no_zero))</pre>
minC[1] \leftarrow maxC[1] \leftarrow 0
co <- layout_with_fr(full_no_zero, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(f
#PLotting
plot(full_no_zero,
     layout=co,
     \#edge.color=V(full\_no\_zero)\$vertex\_full\_no\_zero\_color\_degree[edge.start],
     edge.arrow.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="all"),
     edge.width=E(full_no_zero)$weight/2*mean(E(full_no_zero)$weight),
     edge.curved = TRUE,
     vertex.color=V(full_no_zero)$vertex_full_no_zero_color_degree,
     vertex.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="all")*10^4,
     vertex.frame.color="white",
     vertex.label.color="black",
     vertex.label=get.vertex.attribute(full_no_zero, "LABEL_COR"),
     vertex.label.cex=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, 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(full_no_zero)$full_no_zero_color_degree
b<-V(full_no_zero)$vertex_full_no_zero_color_degree
c<-table(a,b)</pre>
d<-as.data.frame(c)</pre>
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 - 1.1 - REDE COMPLETA (full_no_zero)", sub = "S
    x=range(co[,1])[1],
    y=range(co[,2])[1],
      labels = sprintf(
             "Median all Closennes:%.4f\nSD all Closennes: %.5f",
             median(closeness(full_no_zero, mode="all", weights = E(full_no_zero)$full_no_zero)),
             sd(closeness(full_no_zero, mode="all", weights = E(full_no_zero)$full_no_zero))
```

## Network Closeness Degree Sized and Colored all – 1.1 – REDE COMPLETA (full\_no\_zero)



## 3.5 Closeness Normalized

## 3.5.1 Saving to Igraph object

```
V(full_no_zero)$incloseness_n <- closeness(full_no_zero, mode = "in",, weights = E(full_no_zero)$full_n
V(full_no_zero)$outcloseness_n <- closeness(full_no_zero, mode = "out", normalized = T, weights = E(full_no_zero)$totalcloseness_n <- closeness(full_no_zero, mode = "total", normalized = T, weights = E
```

#### 3.5.2 Saving to Environment

```
full_no_zero_incloseness_n<- closeness(full_no_zero, mode = "in", normalized = T, weights = E(full_no_z full_no_zero_outcloseness_n<- closeness(full_no_zero, mode = "out", normalized = T, weights = E(full_no_full_no_zero_totalcloseness_n<- closeness(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights = E(full_no_zero, mode = "total", normalized = T, weights =
```

#### 3.5.3 Closeness Normalized - IN

```
summary(full_no_zero_incloseness_n)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.005348 0.027540 0.027830 0.027620 0.027880 0.028910

sd(full_no_zero_incloseness_n)

## [1] 0.001675712
```

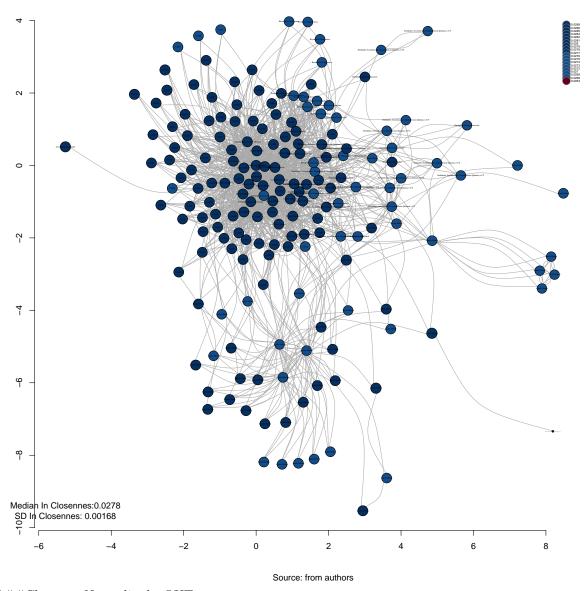
## 3.6 Network Plotting Based On Normalized Closeness - IN

```
V(full_no_zero) $incloseness_n <-closeness(full_no_zero, weights = E(full_no_zero) $full_no_zero, mode="in
#Get Variable
V(full_no_zero)$full_no_zero_color_degree<-round(V(full_no_zero)$incloseness_n,4)
#Creating brewer pallette
vertex_full_no_zero_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
          V(full_no_zero)$full_no_zero_color_degree)), "RdBu"))(
            length(unique(V(full_no_zero)$full_no_zero_color_degree)))
#Saving as Vertex properties
V(full_no_zero)$vertex_full_no_zero_color_degree<-
  vertex_full_no_zero_color_degree[as.numeric(
  cut(V(full_no_zero)$full_no_zero_color_degree,
      breaks=length(unique(V(full_no_zero)$full_no_zero_color_degree))))]
set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(full_no_zero, es=E(full_no_zero), names=F)[,1]</pre>
# Fixing ego
minC <- rep(-Inf, vcount(full_no_zero))</pre>
maxC <- rep(Inf, vcount(full_no_zero))</pre>
minC[1] \leftarrow maxC[1] \leftarrow 0
co <- layout_with_fr(full_no_zero, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(f
```

```
#PLotting
plot(full_no_zero,
     layout=co,
     \#edge.color=V(full\_no\_zero)\$vertex\_full\_no\_zero\_color\_degree[edge.start],
     edge.arrow.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="in",normaliz
     edge.width=E(full_no_zero)$weight/10*mean(E(full_no_zero)$weight),
     edge.curved = TRUE,
     vertex.color=V(full_no_zero)$vertex_full_no_zero_color_degree,
     vertex.size=(closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="in",normalized
     vertex.frame.color="black",
     vertex.label.color="black",
     vertex.label=get.vertex.attribute(full_no_zero,"LABEL_COR"),
     vertex.label.cex=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="in",normali:
     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(full_no_zero)$full_no_zero_color_degree
b<-V(full_no_zero)$vertex_full_no_zero_color_degree
c<-table(a,b)
d<-as.data.frame(c)</pre>
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 - 1.1 - REDE COMPLETA (full_no_zero)", sub = "Sou
   x=range(co[,1])[1],
   y=range(co[,2])[1],
      labels = sprintf(
             "Median In Closennes: %.4f\nSD In Closennes: %.5f",
             median(closeness(full_no_zero, mode="in", weights = E(full_no_zero)$full_no_zero, normaliz
```

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

### Network Closeness Degree Sized Normalized In - 1.1 - REDE COMPLETA (full\_no\_zero)



###ClosenessNormalized - OUT

```
summary(full_no_zero_outcloseness_n)
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.005348 0.229800 0.287000 0.236300 0.297600 0.392400
```

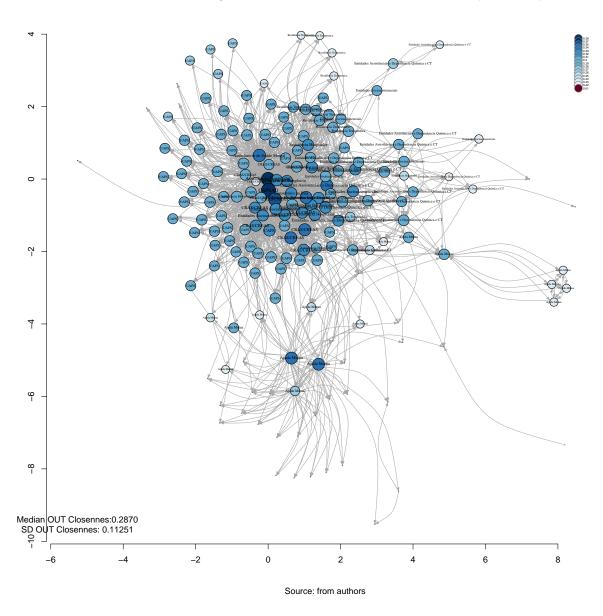
```
sd(full_no_zero_outcloseness_n)
```

## 3.7 Network Plotting Based On Normalized Closeness - OUT

```
V(full_no_zero) $outcloseness_n<-closeness(full_no_zero, weights = E(full_no_zero) $full_no_zero, mode="o
#Get Variable
V(full no zero) $full no zero color degree <-round (V(full no zero) $outcloseness n,2)
#Creating brewer pallette
vertex_full_no_zero_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
          V(full_no_zero)$full_no_zero_color_degree)), "RdBu"))(
            length(unique(V(full_no_zero)$full_no_zero_color_degree)))
#Saving as Vertex properties
V(full_no_zero)$vertex_full_no_zero_color_degree<-
  vertex_full_no_zero_color_degree[as.numeric(
  cut(V(full_no_zero)$full_no_zero_color_degree,
      breaks=length(unique(V(full_no_zero)$full_no_zero_color_degree))))]
set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(full_no_zero, es=E(full_no_zero), names=F)[,1]</pre>
# Fixing ego
minC <- rep(-Inf, vcount(full_no_zero))
maxC <- rep(Inf, vcount(full_no_zero))</pre>
minC[1] \leftarrow maxC[1] \leftarrow 0
co <- layout_with_fr(full_no_zero, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(f
#PLotting
plot(full_no_zero,
     layout=co,
     \#edge.color=V(full\_no\_zero)\$vertex\_full\_no\_zero\_color\_degree[edge.start],
     edge.arrow.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="out",normali:
     edge.width=E(full_no_zero)$weight/10*mean(E(full_no_zero)$weight),
     edge.curved = TRUE,
     vertex.color=V(full_no_zero)$vertex_full_no_zero_color_degree,
     vertex.size=(closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="out", normalized
     vertex.frame.color="black",
     vertex.label.color="black",
     vertex.label=get.vertex.attribute(full_no_zero, "LABEL_COR"),
     vertex.label.cex=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="out",normal
     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(full_no_zero)$full_no_zero_color_degree
b<-V(full_no_zero)$vertex_full_no_zero_color_degree
c<-table(a,b)</pre>
d<-as.data.frame(c)</pre>
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 - 1.1 - REDE COMPLETA (full_no_zero)", sub = "So
    x=range(co[,1])[1],
    y=range(co[,2])[1],
      labels = sprintf(
             "Median OUT Closennes: %.4f\nSD OUT Closennes: %.5f",
             median(closeness(full_no_zero, mode="out", weights = E(full_no_zero)$full_no_zero, normalis
             sd(closeness(full_no_zero, mode="out", weights = E(full_no_zero)$full_no_zero, normalized
```

## Network Closeness Degree Sized Normalized OUT - 1.1 - REDE COMPLETA (full\_no\_zero)



## 3.7.1 Closeness Normalized - ALL

```
summary(full_no_zero_totalcloseness_n)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.2844 0.4237 0.4559 0.4484 0.4769 0.7381

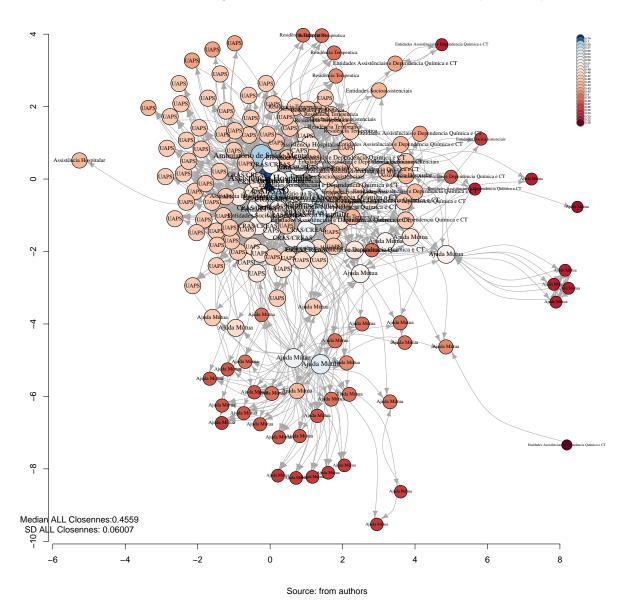
sd(full_no_zero_totalcloseness_n)
```

## 3.8 Network Plotting Based On Normalized Closeness - ALL

```
V(full_no_zero) allcloseness_n<-closeness(full_no_zero, weights = E(full_no_zero) full_no_zero, mode="a
#Get Variable
V(full no zero) $full no zero color degree <-round (V(full no zero) $allcloseness n,2)
#Creating brewer pallette
vertex_full_no_zero_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
          V(full_no_zero)$full_no_zero_color_degree)), "RdBu"))(
            length(unique(V(full_no_zero)$full_no_zero_color_degree)))
#Saving as Vertex properties
V(full_no_zero)$vertex_full_no_zero_color_degree<-
  vertex_full_no_zero_color_degree[as.numeric(
  cut(V(full_no_zero)$full_no_zero_color_degree,
      breaks=length(unique(V(full_no_zero)$full_no_zero_color_degree))))]
set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(full_no_zero, es=E(full_no_zero), names=F)[,1]</pre>
# Fixing ego
minC <- rep(-Inf, vcount(full_no_zero))
maxC <- rep(Inf, vcount(full_no_zero))</pre>
minC[1] \leftarrow maxC[1] \leftarrow 0
co <- layout_with_fr(full_no_zero, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(f
#PLotting
plot(full_no_zero,
     layout=co,
     \#edge.color=V(full\_no\_zero)\$vertex\_full\_no\_zero\_color\_degree[edge.start],
     edge.arrow.size=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="all",normali:
     edge.width=E(full_no_zero)$weight/10*mean(E(full_no_zero)$weight),
     edge.curved = TRUE,
     vertex.color=V(full_no_zero)$vertex_full_no_zero_color_degree,
     vertex.size=(closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="all",normalized
     vertex.frame.color="black",
     vertex.label.color="black",
     vertex.label=get.vertex.attribute(full_no_zero, "LABEL_COR"),
     vertex.label.cex=closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode="all",normal
     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(full_no_zero)$full_no_zero_color_degree
b<-V(full_no_zero)$vertex_full_no_zero_color_degree
c<-table(a,b)</pre>
d<-as.data.frame(c)</pre>
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 - 1.1 - REDE COMPLETA (full_no_zero)", sub = "So
    x=range(co[,1])[1],
    y=range(co[,2])[1],
      labels = sprintf(
             "Median ALL Closennes: %.4f\nSD ALL Closennes: %.5f",
             median(closeness(full_no_zero, mode="all", weights = E(full_no_zero)$full_no_zero, normali:
             sd(closeness(full_no_zero, mode="all", weights = E(full_no_zero)$full_no_zero, normalized
```

## Network Closeness Degree Sized Normalized ALL – 1.1 – REDE COMPLETA (full\_no\_zero)



## 3.9 Closeness Normalized

## 3.9.1 Saving to Igraph object

V(full\_no\_zero)\$incloseness\_n <- closeness(full\_no\_zero, weights = E(full\_no\_zero)\$full\_no\_zero, mode = V(full\_no\_zero)\$outcloseness\_n <- closeness(full\_no\_zero, weights = E(full\_no\_zero)\$full\_no\_zero, mode = V(full\_no\_zero)\$totalcloseness\_n <- closeness(full\_no\_zero, weights = E(full\_no\_zero)\$full\_no\_zero, mode = V(full\_no\_zero)\$totalcloseness\_n <- closeness(full\_no\_zero, weights = E(full\_no\_zero)\$full\_no\_zero, mode = V(full\_no\_zero)\$full\_no\_zero, mo

### 3.10 Centralization Closseness

```
V(full_no_zero)$full_no_zero_centr_closeness<- centralization.closeness(full_no_zero)$res full_no_zero_centr_closeness<- centralization.closeness(full_no_zero)$res full_no_zero_centr_closeness_all<- centralization.closeness(full_no_zero)
```

#### 3.10.1 Centralization

```
full_no_zero_centr_closeness_all$centralization
## [1] 0.15777
```

#### 3.10.2 Theoretical Max

```
full_no_zero_centr_closeness_all$theoretical_max
## [1] 185.0053
```

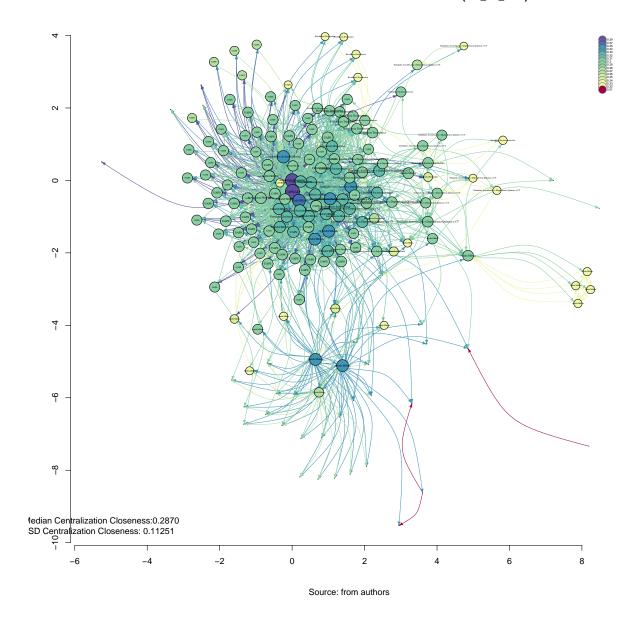
## 3.11 Network Plotting Based On Centralization Closeness

```
V(full_no_zero)$full_no_zero_centr_closeness<- centralization.closeness(full_no_zero)$res
#Get Variable
V(full_no_zero)$full_no_zero_color_degree<-round(V(full_no_zero)$full_no_zero_centr_closeness,2)
#Creating brewer pallette
vertex_full_no_zero_color_degree<-
  colorRampPalette(brewer.pal(length(unique(
          V(full_no_zero)$full_no_zero_color_degree)), "Spectral"))(
            length(unique(V(full_no_zero)$full_no_zero_color_degree)))
#Saving as Vertex properties
V(full_no_zero)$vertex_full_no_zero_color_degree<-
  vertex_full_no_zero_color_degree[as.numeric(
  cut(V(full_no_zero)$full_no_zero_color_degree,
      breaks=length(unique(V(full_no_zero)$full_no_zero_color_degree))))]
set.seed(123)
#Plotting based only on degree measures
edge.start <- ends(full_no_zero, es=E(full_no_zero), names=F)[,1]</pre>
# Fixing eqo
minC <- rep(-Inf, vcount(full_no_zero))</pre>
maxC <- rep(Inf, vcount(full_no_zero))</pre>
minC[1] <- maxC[1] <- 0
co <- layout_with_fr(full_no_zero, niter=10^4, minx=minC, maxx=maxC,miny=minC, maxy=maxC, weights = E(f
```

```
#PLotting
plot(full_no_zero,
     layout=co,
     edge.color=V(full_no_zero)$vertex_full_no_zero_color_degree[edge.start],
     edge.arrow.size=centralization.closeness(full_no_zero)$res,
     \verb|edge.width=E(full_no_zero)$| weight/10*| mean(E(full_no_zero)$| weight), |
     edge.curved = TRUE,
     vertex.color=V(full_no_zero)$vertex_full_no_zero_color_degree,
     vertex.size=centralization.closeness(full_no_zero)$res*100,
     vertex.frame.color="black",
     vertex.label.color="black",
     vertex.label=get.vertex.attribute(full_no_zero,"LABEL_COR"),
     vertex.label.cex=centralization.closeness(full_no_zero)$res,
     vertex.label.dist=0,
     rescale=F,
     xlim=range(co[,1]),
     ylim=range(co[,2])
axis(1)
axis(2)
#Solving Problems with legend rendering
a<-V(full_no_zero)$full_no_zero_color_degree
b<-V(full_no_zero)$vertex_full_no_zero_color_degree
c<-table(a,b)
d<-as.data.frame(c)</pre>
e<-subset(d, d$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 - 1.1 - REDE COMPLETA (full_no_zero)", sub = "Source: from au
  text(
    x=range(co[,1])[1],
    y=range(co[,2])[1],
      labels = sprintf(
             "Median Centralization Closeness: %.4f\nSD Centralization Closeness: %.5f",
```

```
median(centralization.closeness(full_no_zero)$res),
sd(centralization.closeness(full_no_zero)$res)
)
```

## Network Centralization Closeness – 1.1 – REDE COMPLETA (full\_no\_zero)



# 4 Closeness Dinamic Table

# 4.1 Getting Closeness Measures

```
full_no_zero_incloseness<- closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode = "in")
full_no_zero_outcloseness<- closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode = "out
full_no_zero_totalcloseness<- closeness(full_no_zero, weights = E(full_no_zero)$full_no_zero, mode = "t
full_no_zero_incloseness_n<- closeness(full_no_zero,weights = E(full_no_zero)$full_no_zero, mode = "in"
full_no_zero_outcloseness_n<- closeness(full_no_zero,weights = E(full_no_zero)$full_no_zero, mode = "out
full_no_zero_totalcloseness_n<- closeness(full_no_zero,weights = E(full_no_zero)$full_no_zero, mode = "out
full_no_zero_centr_closeness<- centralization.closeness(full_no_zero)$res %>% round(6)
```

## 4.2 Creating a datagrame of measures

```
full_no_zero_df_closseness <- data.frame(
full_no_zero_incloseness,
full_no_zero_outcloseness,
full_no_zero_totalcloseness,
full_no_zero_incloseness,
full_no_zero_incloseness_n,
full_no_zero_outcloseness_n,
full_no_zero_totalcloseness_n,
full_no_zero_totalcloseness_n,
full_no_zero_centr_closeness) %>% round(6)

#Adding type
full_no_zero_df_closseness <-cbind(full_no_zero_df_closseness, V(full_no_zero)$LABEL_COR)

#Adding names
names(full_no_zero_df_closseness) <- c("In Closeness", "Out Closeness", "Total Closeness", "In Closeness
#Ordering Variables
full_no_zero_df_closseness<-full_no_zero_df_closseness[c("Type","In Closeness", "Out Closeness", "Total</pre>
```

#### 4.3 General tabel - DT

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

entries					Search:			
Type \$	In Closeness	Out Closeness	Total Closeness	In Closeness # Normalized	Out Closeness \(\phi\) Normalized	Total Closeness   Normalized	Centralization Closeness	
All	All	All	All	All	All	All	All	
Assistência Hospitalar	0.000151	0.002	0.003774	0.028165	0.372	0.701887	0.37	
Ambulatório de Saúde Mental	0.00015	0.001848	0.003175	0.027974	0.343808	0.590476	0.34380	
CAPSAD	0.000152	0.00211	0.003968	0.028354	0.392405	0.738095	0.39240	
CRAS/CREAS	0.00015	0.001808	0.002809	0.027886	0.336347	0.522472	0.33634	
CRAS/CREAS	0.00015	0.001661	0.002778	0.027915	0.30897	0.516667	0.3089	
CRAS/CREAS	0.00015	0.001623	0.002725	0.02789	0.301948	0.506812	0.30194	
Assistência Hospitalar	0.000149	0.001639	0.002618	0.027803	0.304918	0.486911	0.30491	
Entidades Assistênciais e Dependencia Química e CT	0.00015	0.001715	0.002857	0.027869	0.319039	0.531429	0.31903	
Entidades Assistênciais e Dependencia Química e CT	0.000147	0.001675	0.002532	0.027317	0.311558	0.470886	0.31155	
Entidades Assistênciais e Dependencia Química e CT	0.000149	0.00155	0.002513	0.027807	0.288372	0.467337	0.28837	
	Assistência Hospitalar  Ambulatório de Saúde Mental  CAPSAD  CRAS/CREAS  CRAS/CREAS  CRAS/CREAS  Assistência Hospitalar  Entidades Assistênciais e Dependencia Química e CT  Entidades Assistênciais e Dependencia Química e CT	Type Closeness  All All All  Assistência Hospitalar  CAPSAD 0.000152  CRAS/CREAS 0.00015  CRAS/CREAS 0.000149	Type         In Closeness         Closeness           All         All         All           Assistência Hospitalar         0.000151         0.002           Ambulatório de Saúde Mental         0.00015         0.001848           CAPSAD         0.000152         0.00211           CRAS/CREAS         0.00015         0.001808           CRAS/CREAS         0.00015         0.001661           CRAS/CREAS         0.00015         0.001623           Assistência Hospitalar         0.000149         0.001639           Entidades Assistênciais e Dependencia Química e CT         0.000147         0.001675           Entidades Assistênciais e Dependencia Química e CT         0.000149         0.001675           Entidades Assistênciais e Dependencia Química e CT         0.000149         0.00155	Type         Closeness         Closeness         Closeness           All         All         All         All           Assistência Hospitalar         0.000151         0.002         0.003774           Ambulatório de Saúde Mental         0.00015         0.001848         0.003175           CAPSAD         0.000152         0.00211         0.003968           CRAS/CREAS         0.00015         0.001808         0.002809           CRAS/CREAS         0.00015         0.001661         0.002778           CRAS/CREAS         0.00015         0.001623         0.002725           Assistência         0.000149         0.001639         0.002618           Entidades Assistênciais e Dependencia Química e CT         0.000147         0.001675         0.002532           Entidades Assistênciais e Dependencia Química e CT         0.000149         0.00155         0.002513	Type         Closeness         Closeness         Closeness         Closeness Normalized           All         All	Type         Closeness         Closeness         Closeness         Closeness Normalized         Closeness Normalized           All         All <t< td=""><td>  Type</td></t<>	Type	

# 4.4 Aggregating data from previous table - mean

```
aggdata_mean <-aggregate(full_no_zero_df_closseness, by=list(full_no_zero_df_closseness$Type), FUN=mean
names(aggdata_mean) <- c("Group","Type","In Closeness(M)", "Out Closeness(M)", "Total Closeness(M)","In
#Removing Type variable
aggdata_mean<-aggdata_mean[,-c(2)]
```

## 4.5 Aggregating data from previous table - sd

```
aggdata_sd <-aggregate(full_no_zero_df_closseness, by=list(full_no_zero_df_closseness$Type), FUN=sd, na
names(aggdata_sd) <- c("Group","Type","In Closeness(SD)", "Out Closeness(SD)", "Total Closeness(SD)","In
#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 toghter

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

## 4.6 Plotting final table with round for Closseness



# 5 Saving objects with new variables and changes

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