# Network Analysis and Centrality Measures

```
knitr::opts chunk$set(echo = TRUE)
library(tidyverse)
## Warning: package 'ggplot2' was built under R version 4.3.1
## Warning: package 'lubridate' was built under R version 4.3.1
## — Attaching core tidyverse packages -
                                                              — tidyverse 2.0.0 —
## ✓ dplvr 1.1.2
                        ✓ readr
                                    2.1.4
## ✓ forcats 1.0.0
                                    1.5.0

✓ stringr

## ✓ ggplot2 3.5.0

✓ tibble

                                    3.2.1
## ✓ lubridate 1.9.3

✓ tidyr

                                    1.3.0
## ✓ purrr
              1.0.1
## — Conflicts —
                                                         – tidyverse conflicts() —
## * dplyr::filter() masks stats::filter()
## * dplyr::lag() masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts
to become errors
library(igraph)
## Warning: package 'igraph' was built under R version 4.3.1
```

```
localhost: 63342/Org-Analytics/Ex2/ex2.html?\_ijt=iccuh0o0tnma3ot4l0fmd02bea\&\_ij\_reload=RELOAD\_ON\_SAVE
```

```
##
## Attaching package: 'igraph'
##
  The following objects are masked from 'package:lubridate':
##
##
##
       %--%, union
##
##
  The following objects are masked from 'package:dplyr':
##
       as_data_frame, groups, union
##
##
  The following objects are masked from 'package:purrr':
##
##
       compose, simplify
##
##
##
  The following object is masked from 'package:tidyr':
##
##
       crossing
##
## The following object is masked from 'package:tibble':
##
##
       as_data_frame
##
## The following objects are masked from 'package:stats':
##
##
       decompose, spectrum
##
## The following object is masked from 'package:base':
##
##
       union
```

#### library(kableExtra)

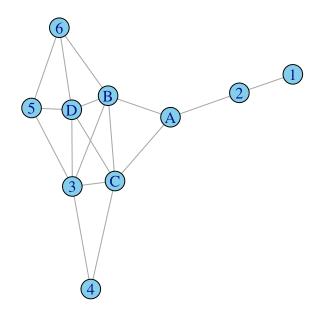
```
## Warning: package 'kableExtra' was built under R version 4.3.1
```

```
##
## Attaching package: 'kableExtra'
##
## The following object is masked from 'package:dplyr':
##
## group_rows
```

```
# Create a blank matrix
adjacency matrix <- matrix(0, nrow=10, ncol=10)
# Define the names for the rows and columns for clarity
rownames(adjacency_matrix) <- c("5", "6", "D", "3", "4", "B", "C", "A", "2", "1")
colnames(adjacency_matrix) <- c("5", "6", "D", "3", "4", "B", "C", "A", "2", "1")
# Connections for 5
adjacency_matrix["5", c("D", "6", "3")] <- 1
# Connections for 6
adjacency_matrix["6", c("D", "B", "5")] <- 1
# Connections for D
adjacency_matrix["D", c("5", "6", "3", "B", "C")] <- 1
# Connections for 3
adjacency_matrix["3", c("D", "4", "C", "B", "5")] <- 1
# Connections for 4
adjacency_matrix["4", c("3", "C")] <- 1
# Connections for B
adjacency_matrix["B", c("6", "D", "C", "A", "3")] <- 1
# Connections for C
adjacency_matrix["C", c("3", "4", "B", "D", "A")] <- 1
# Connections for A
adjacency_matrix["A", c("C", "B", "2")] <- 1
# Connections for 2
adjacency_matrix["2", c("A", "1")] <- 1</pre>
# Connections for 1
adjacency_matrix["1", "2"] <- 1</pre>
# Since the graph is undirected, we mirror the matrix along the diagonal
adjacency_matrix <- adjacency_matrix + t(adjacency_matrix)</pre>
# convert where the value is 2 to 1
adjacency_matrix[adjacency_matrix == 2] <- 1</pre>
#sort the matrix
adjacency matrix <- adjacency matrix[order(rownames(adjacency matrix)), order(colnames(a
djacency_matrix))]
# Print out the matrix
adjacency matrix
```

```
## 1 2 3 4 5 6 A B C D
## 1 0 1 0 0 0 0 0 0 0 0 0
## 2 1 0 0 0 0 0 1 0 0 0
## 3 0 0 0 1 1 0 0 1 1 1
## 4 0 0 1 0 0 0 0 0 1 0
## 5 0 0 1 0 0 1 0 0 0 1
## 6 0 0 0 0 1 0 0 1 0 1
## A 0 1 0 0 0 0 0 1 1 0
## B 0 0 1 0 0 1 1 0 1
## C 0 0 1 1 0 0 1 1 0 1
## D 0 0 1 0 1 1 0 1 1
```

```
library(igraph)
graph <- graph_from_adjacency_matrix(adjacency_matrix, mode = "undirected")
# Plot the graph
plot(graph, layout = layout_nicely(graph), vertex.color = "skyblue", edge.arrow.size = 0.5)</pre>
```



```
# Calculate degree centrality for all nodes
degree_centrality <- degree(graph, mode = "all")

# Calculate closeness centrality for all nodes
closeness_centrality <- closeness(graph, mode = "all")

# Calculate betweenness centrality for all nodes
betweenness_centrality <- betweenness(graph, directed = FALSE, normalized = TRUE)

centrality_measures_all <- data.frame(
   Degree = degree_centrality,
   Closeness = closeness_centrality,
   Betweenness = betweenness_centrality
)

centrality_measures_all</pre>
```

```
##
    Degree Closeness Betweenness
## 1
         1 0.03333333 0.00000000
## 2
         2 0.04545455 0.22222222
## 3
         5 0.06250000 0.12870370
         2 0.05000000 0.00000000
## 4
## 5
         3 0.04761905 0.01481481
## 6
         3 0.05263158 0.02592593
         3 0.06250000 0.38888889
## A
         5 0.07142857 0.25092593
## B
## C
         5 0.07142857 0.23888889
## D
         5 0.06250000 0.09074074
```

```
# Extract the centralities for seats A-D
seat_choices <- c('A', 'B', 'C', 'D')
centrality_measures <- data.frame(
   Degree = degree_centrality[seat_choices],
   Closeness = closeness_centrality[seat_choices],
   Betweenness = betweenness_centrality[seat_choices]
)
rownames(centrality_measures) <- seat_choices</pre>
```

#### Centrality Measures for Seats A-D

	Degree	Closeness	Betweenness
Α	3	0.06250	0.388889
В	5	0.07143	0.250926
С	5	0.07143	0.238889
D	5	0.06250	0.090741

## Seat A

- **Key Feature**: Highest betweenness centrality, lowest degree centrality.
- **Benefit**: Sitting in Seat A positions you as a key bridge in the network, connecting different coworker groups. This is beneficial for meeting diverse individuals -> can offer broad insights into the company culture and cross-departmental projects.
- **Drawback**: The lower degree centrality means you might have fewer direct connections, potentially making it harder to quickly form a close-knit group of friends.

#### Seat B

- Key Feature: High degree and closeness centrality.
- Benefit: access to a larger direct network and easier communication with others due to the high closeness centrality, facilitating exchange of information and support.
- **Drawback**: the focus here is less on being the sole connector and more on being part of a cohesive network, which might **limit exclusive networking**.

### Seat C

• Key Feature: Similar to Seat B.

## Seat D

- Key Feature: High degree and closeness centrality but lowest betweenness centrality.
- Benefit: Similar to B and C
- **Drawback**: The lowest betweenness centrality = less ideal for those looking to play a bridging role between unconnected coworker groups (less networkhub role).

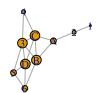
In Summary: - For broad networking across diverse groups: Seat A - For quickly establishing a strong, central presence in your immediate network: Seats B, C, and D

What if... - if you want to make the strongest bond with one person in the bus, where should you sit, assuming every seat is available? - 1 or 2 - they have the least number of connections, they are far from the rest of the network, and they are not in the middle of the network.

If you choose 1, then 2 is your only bet and 2 has more "opportunities". So if you want to make the strongest connection, you would sit in seat 2 and only talk to the person in seat 1.

```
degree centrality <- degree(graph, mode = "all")</pre>
closeness_centrality <- closeness(graph, mode = "all")</pre>
betweenness_centrality <- betweenness(graph, directed = FALSE)</pre>
# Set up the layout of the graph just once to use in all plots
layout <- layout_with_fr(graph)</pre>
# Set up the plotting area
par(mfrow = c(2, 2))
# Degree Centrality Plot
plot(graph, layout = layout,
     vertex.label = V(graph)$name,
     vertex.size = degree centrality * 5,
     vertex.label.cex = 0.8,
     edge.arrow.size = 0.5,
     main = "Degree Centrality")
# Closeness Centrality Plot
plot(graph, layout = layout,
     vertex.label = V(graph)$name,
     vertex.size = closeness_centrality * 500, # Scale factor to make the sizes visibl
e, adjust as needed
     vertex.label.cex = 0.8,
     edge.arrow.size = 0.5,
     main = "Closeness Centrality")
# Betweenness Centrality Plot
plot(graph, layout = layout,
     vertex.label = V(graph)$name,
     vertex.size = betweenness_centrality / max(betweenness_centrality) * 50, # Normali
ze and scale, adjust as needed
     vertex.label.cex = 0.8,
     edge.arrow.size = 0.5,
     main = "Betweenness Centrality")
# Resetting to default single plotting layout
par(mfrow = c(1, 1))
```

## **Degree Centrality**



# **Closeness Centrality**



# **Betweenness Centrality**



par(mai = c(0, 0, 0.5, 0.5))