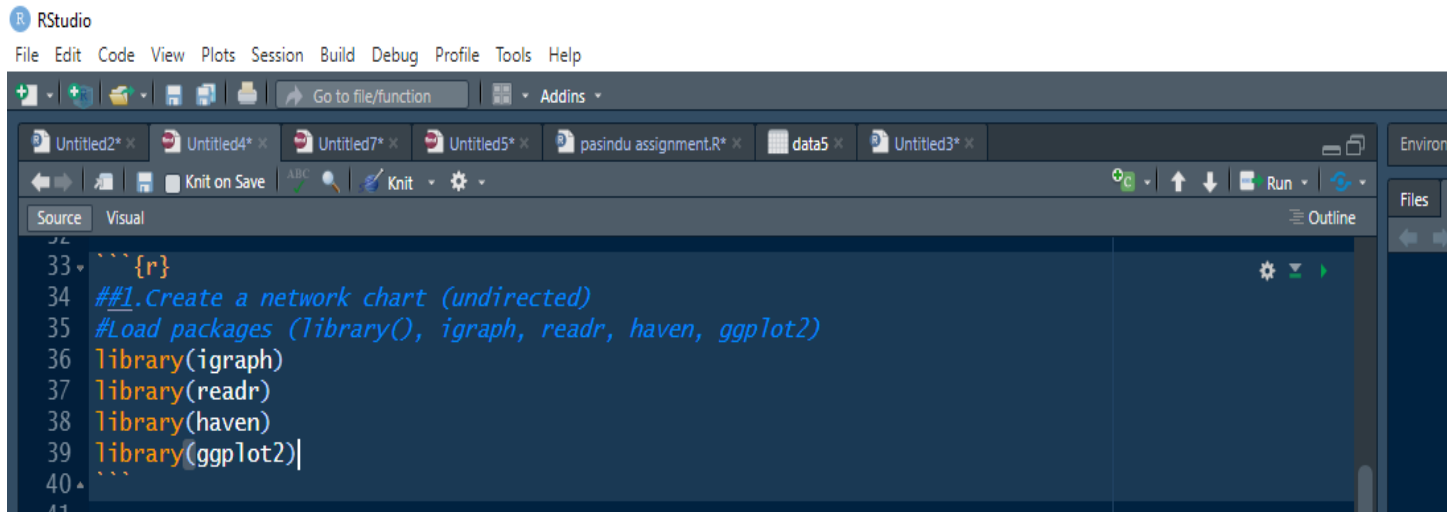


## R-studio for Network Analysis

Use my csv file. The data shows the relationships between the two people. Use **Undirect edge** type analysis.

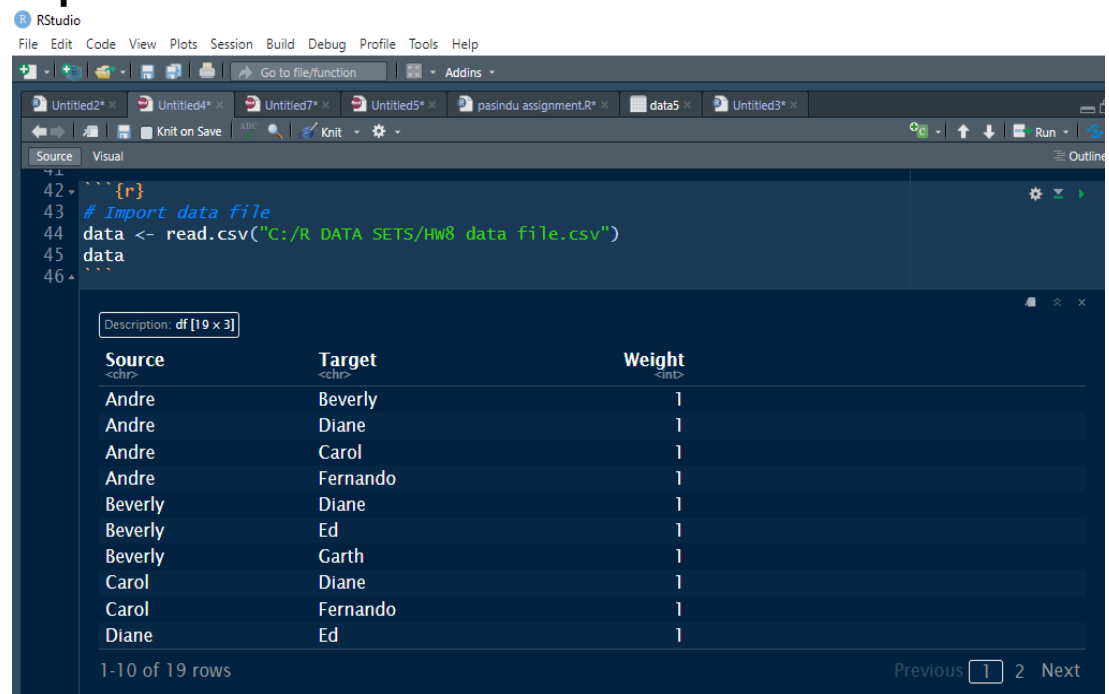
1. Create a network chart (undirected)
  - **Load packages (library(), igraph, readr, haven, ggplot2)**



The screenshot shows the RStudio interface with the Source editor open. The code in the editor is as follows:

```
33 {r}
34 ##1.Create a network chart (undirected)
35 #Load packages (library(), igraph, readr, haven, ggplot2)
36 library(igraph)
37 library(readr)
38 library(haven)
39 library(ggplot2)
40
41
```

- **Import data file**



The screenshot shows the RStudio interface with the Source editor open. The code in the editor is as follows:

```
42 {r}
43 # Import data file
44 data <- read.csv("C:/R DATA SETS/HW8 data file.csv")
45 data
46
```

Below the code editor, the Data Viewer shows a preview of the data. The description is "df [19 x 3]". The data is as follows:

Source <chr>	Target <chr>	Weight <int>
Andre	Beverly	1
Andre	Diane	1
Andre	Carol	1
Andre	Fernando	1
Beverly	Diane	1
Beverly	Ed	1
Beverly	Garth	1
Carol	Diane	1
Carol	Fernando	1
Diane	Ed	1

At the bottom of the Data Viewer, it shows "1-10 of 19 rows" and navigation buttons for "Previous", "1", "2", and "Next".

- Create a graph object (graph.data.frame) – “undirected”

```

33 {r}
34 # Create a graph object (graph.data.frame) – “undirected”
35 graph <- graph.data.frame(data, directed = FALSE)
36 graph
37
38

```

IGRAPH bc27331 UN-- 11 19 --  
+ attr: name (v/c), Weight (e/n)  
+ edges from bc27331 (vertex names):

[1] Andre	--Beverly	Andre	--Diane	Andre	--Carol	Andre	--Fernando
[5] Beverly	--Diane	Beverly	--Ed	Beverly	--Garth	Carol	--Diane
[9] Carol	--Fernando	Diane	--Ed	Diane	--Fernando	Diane	--Garth
[13] Ed	--Garth	Fernando	--Garth	Fernando	--Heather	Garth	--Heather
[17] Heather	--Ike	Ike	--Jane	Heather	--Kathy		

2. Modify the layout of the network chart.

set.seed(1234)

```

{r}
## 2.Modify the layout of the network chart.
# Set seed
set.seed(1234)

```

- fruchterman.reingold layout

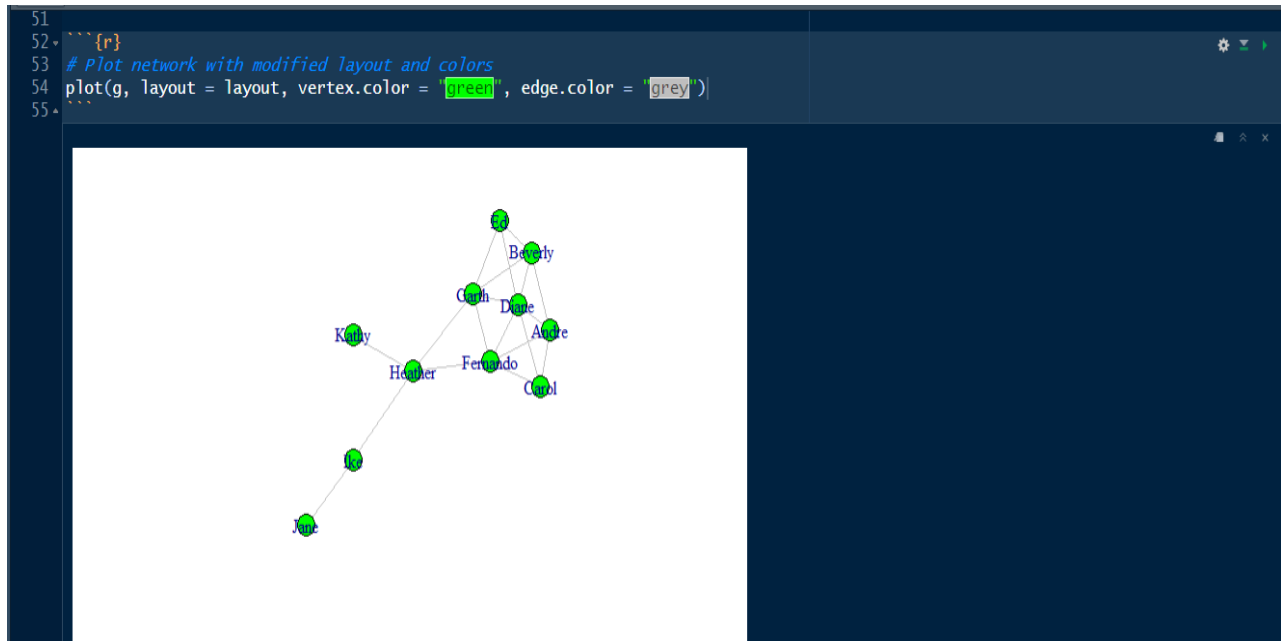
```

5 {r}
7 # Define layout of fruchterman.reingold layout
8 layout <- layout.fruchterman.reingold(graph)
9
10

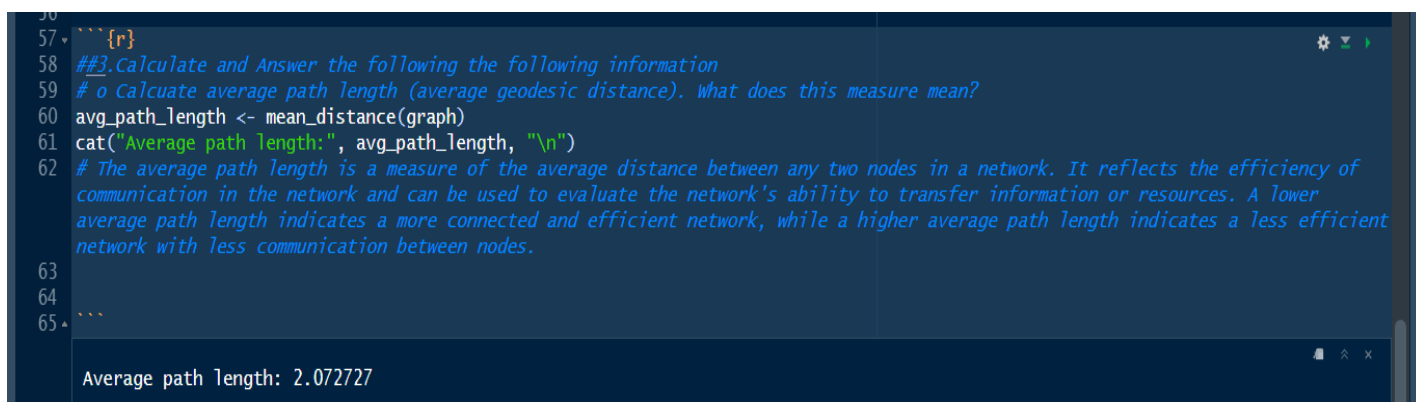
```

- vertex color – “green”
- edge color – “grey”

**Take a Screenshot of the outcome**



3. Calculate and Answer the following the following information
  - Calculate **average path length** (average geodesic distance). What does this measure mean?



- Calculate the **degree centrality**



- Create a data frame using the degree centrality (\*Hint: as.data.frame)

```
81. {r}
82. # Create data frame using degree centrality
83. degree_df <- as.data.frame(degree_centrality)
84. degree_df
85. ...
```

	degree_centrality
Andre	4
Beverly	4
Carol	3
Diane	6
Ed	3
Fernando	5
Garth	5
Heather	4
Ike	2
Jane	1

1-10 of 11 rows

- Identify the **minimum** degree, **maximum** degree, and **average degree**

```
88. {r}
89. # Identify minimum, maximum, and average degree
90. min_degree <- min(degree_centrality)
91. max_degree <- max(degree_centrality)
92. avg_degree <- mean(degree_centrality)
93. cat("Minimum degree:", min_degree, "\n")
94. cat("Maximum degree:", max_degree, "\n")
95. cat("Average degree:", avg_degree, "\n")
96.
97. # What is the difference between average path length and the average degree?
98. # Why is the result different?
99. # The average path length measures the average number of steps along the shortest paths for all possible pairs of nodes in a network, providing a measure of the network's efficiency of communication. On the other hand, the average degree measures the average number of connections that a node has in a network, providing a measure of the network's density or connectedness. These two measures capture different aspects of the network structure, and their values may not be strongly correlated. The difference in their values is due to the fact that they are measuring different aspects of the network.
100.
101.
102. ...
```

```
Minimum degree: 1
Maximum degree: 6
Average degree: 3.454545
```

- What is the difference between **average path length** and the **average degree**? Why is the result different?

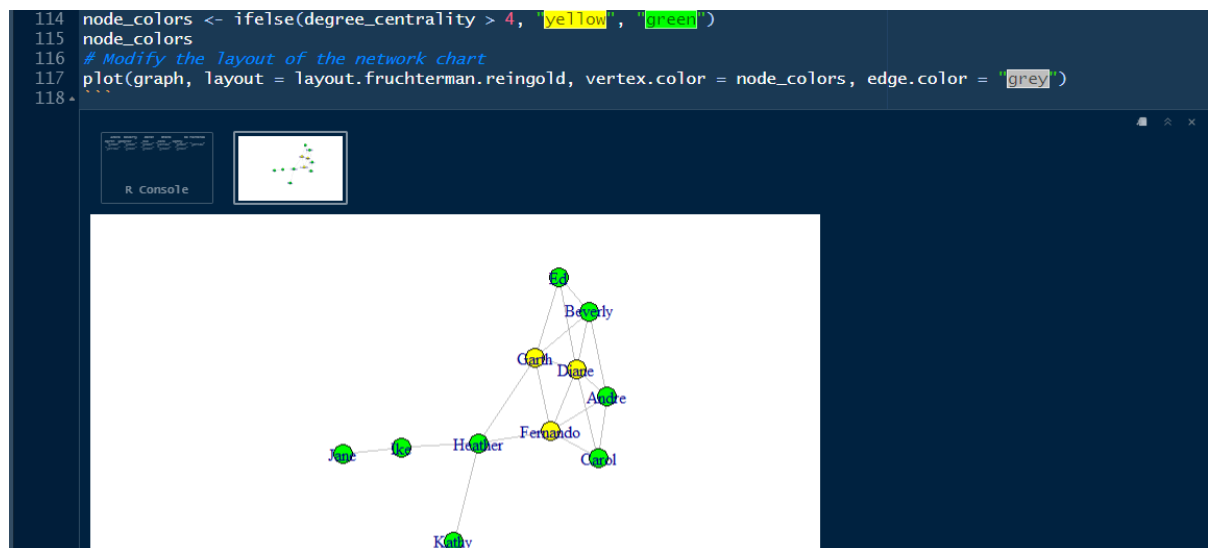
**#The average path length measures the average number of steps along the shortest paths for all possible pairs of nodes in a network, providing a measure of the network's efficiency of communication. On the other hand, the average degree measures the average number of connections that a node has in a network, providing a measure of the network's density or connectedness. These two measures capture different aspects of the network structure, and their values may not be strongly correlated. The difference in their values is due to the fact that they are measuring different aspects of the network.**

4. Modify the layout of the network chart by degree centrality.

```
##{r}
#4. Modify the layout of the network chart by degree centrality. Specifically, change the color of the vertex
with degree centrality of higher than 4 to Yellow.
# Calculate the degree centrality
degree_centrality <- degree(graph)
degree_centrality
```

Andre	Beverly	Carol	Diane	Ed	Fernando	Garth	Heather	Ike	Jane	Kathy
4	4	3	6	3	5	5	4	2	1	1

Specifically, change the color of the vertex with degree centrality of **higher than 4 to Yellow**.



5. Calculate the **Closeness Centrality**. (1.5 pt)

- Create a data frame for the closeness centrality (\*Hint:data.frame)

The screenshot shows the RStudio interface. The source editor contains the following R code:

```
##5. Calculate the Closeness Centrality. (1.5 pt)
# Create a data frame for the closeness centrality (*Hint:data.frame)
closeness <- closeness(graph)
closeness_df <- data.frame(names = v(graph)$name, closeness)
closeness_df
```

The Environment pane on the right shows the variable `high_` with values: `Named int`, `high_ "Heather"`, `high_ "Heather"`, and `high_ "Fernando"`.

The console shows the preview of the data frame `df [11 x 2]`:

	names	closeness
Andre	Andre	0.05000000
Beverly	Beverly	0.05000000
Carol	Carol	0.04761905
Diane	Diane	0.05555556
Ed	Ed	0.04761905
Fernando	Fernando	0.06250000
Garth	Garth	0.06250000
Heather	Heather	0.06250000
Ike	Ike	0.04347826
Jane	Jane	0.03125000

1-10 of 11 rows

- Who has the highest Closeness Centrality?

The screenshot shows the RStudio source editor with the following R code:

```
{r}
# Identify vertex with highest closeness centrality
highest_closeness <- v(graph)$name[which.max(closeness)]
highest_closeness
```

The console shows the output:

```
[1] "Fernando"
```

6. Calculate the **Betweenness Centrality**. (1.5 pt)

- Create a data frame for the betweenness centrality (\*Hint: as.data.frame)

The screenshot shows the RStudio interface. The source editor contains the following R code:

```
145. ##{r}
146. ##6. Calculate the Betweenness Centrality. (1.5 pt)
147. # Create a data frame for the betweenness centrality (*Hint: as.data.frame)
148. betweenness <- betweenness(graph)
149. betweenness_df <- as.data.frame(betweenness)
150. betweenness_df
151.
```

The Environment pane on the right shows the following objects:

- betw... 11 obs. o...
- betw... 11 obs. o...
- clos... 11 obs. o...
- clos... 11 obs. o...

The Data Viewer shows a data frame with 11 rows and 1 column:

	betweenness
Andre	0.8333333
Beverly	0.8333333
Carol	0.0000000
Diane	3.6666667
Ed	0.0000000
Fernando	10.8333333
Garth	10.8333333
Heather	23.0000000
Ike	9.0000000
Jane	0.0000000

1-10 of 11 rows

- Who has the highest Betweenness Centrality?

The screenshot shows the RStudio source editor with the following R code:

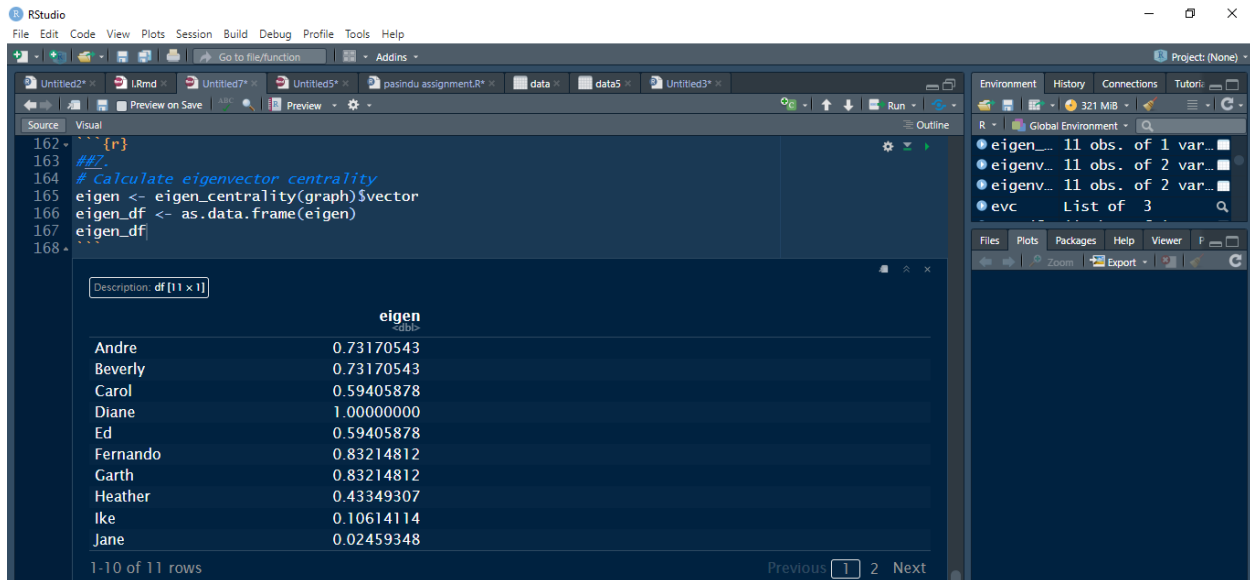
```
156. ##{r}
157. # Identify vertex with highest betweenness centrality
158. highest_betweenness <- V(graph)$name[which.max(betweenness)]
159. highest_betweenness
160.
```

The console output shows:

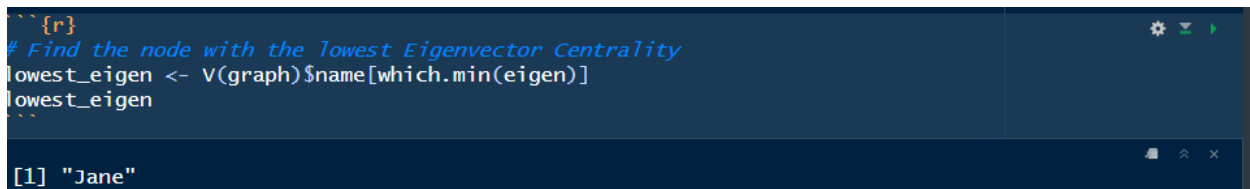
```
[1] "Heather"
```

7. Calculate the **Eigenvector Centrality**. (1.5 pt)

- Create a data frame for the betweenness centrality (\*Hint: as.data.frame)



- Who has the lowest Eigenvector Centrality?



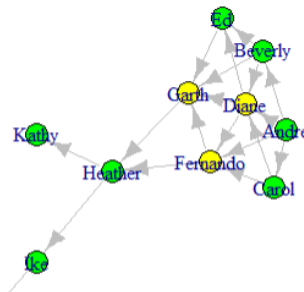
- Repeat the Q1-Q11 process. But this time, create a **DIRECTED** network chart (\*Hint: Do not need to re-write the entire thing. Only the assignment, in the beginning, needs to change!)
  - How is the result different from the **Undirected** network? Choose one of the centrality metrics to look into (e.g., degree centrality, closeness centrality, betweenness centrality, eigenvector centrality). Is the result different from the Undirected network?



```
graph <- graph.data.frame(data, directed = TRUE)
graph
set.seed(1234)
plot(graph, layout = layout_with_fr(graph),
     vertex.color = ifelse(degree(graph) > 4, "yellow", "green"),
     edge.color = "grey")
...

```

R Console



```
162 {r}
163 dc <- degree(graph, mode = "in")
164 dc_df <- data.frame(name = V(graph)$name, degree centrality = dc)
165 dc_df
166

```

Description: df [11 x 2]

	name	degree centrality
Andre	Andre	0
Beverly	Beverly	1
Carol	Carol	1
Diane	Diane	3
Ed	Ed	2
Fernando	Fernando	3
Garth	Garth	4
Heather	Heather	2
Ike	Ike	1
Jane	Jane	1

1-10 of 11 rows

Previous 1 2 Next

**#If we compare this to the degree centrality in the undirected network, we can see that the values are different. In the directed network, we have separate measures for the in-degree and out-degree of each vertex, which can result in different values than in the undirected case.**

- Based on the results, would your decision of the influencer change from Q12?

**Take a Screenshot of the outcome**

**#Based on the results, our decision on the influencer might change depending on the centrality metric we are using. For example, if we focus on in-degree centrality, we can see that vertex 1 has the highest value, which indicates that it receives the most incoming**

connections. However, if we look at betweenness centrality, vertex 5 has the highest value, which indicates that it plays a critical role in connecting different parts of the network. #Therefore, it is important to carefully choose the appropriate centrality metric depending on the research question and the characteristics of the network.