

Network analysis; an introduction (with igraph in R)

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Outline of brief introduction to Network Analysis

- ① What is relational view and network analysis?

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- ⑤ A real life example from science studies!

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- ⑤ A real life example from science studies!
- ⑥ Where to next?!


What is this?!

Now what?!

ID	Name	Age	Political view	Education
1	Tom	24	left	NA
2	Sara	22	right	BA
3	Bill	30	neutral	MA
4	Margaret	31	NA	PhD

A poll/survey results?

Variables/attributes



ID	Name	Age	Political view	Education
1	Tom	24	left	NA
2	Sara	22	right	BA
3	Bill	30	neutral	MA
4	Margaret	31	NA	PhD

A variable by observation table

		Variables/attributes			
Respondents/ Observations	ID	Name	Age	Political view	Education
	1	Tom	24	left	NA
	2	Sara	22	right	BA
	3	Bill	30	neutral	MA
	4	Margaret	31	NA	PhD

What if respondents know each other?!

The diagram shows a data table with five columns and four rows. A bracket above the columns is labeled 'Variables/attributes'. A bracket to the left of the rows is labeled 'Respondents/ Observations'. In the 'Name' column, the entries are 'Tom', 'Sara', 'Bill', and 'Margaret'. Red double-headed vertical arrows are placed between 'Sara' and 'Bill', and between 'Bill' and 'Margaret', indicating a relationship or awareness between these respondents.

Variables/attributes				
ID	Name	Age	Political view	Education
1	Tom	24	left	NA
2	Sara	22	right	BA
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Respondents/
Observations

Different contexts of familiarity

- Family, college, gym, ...

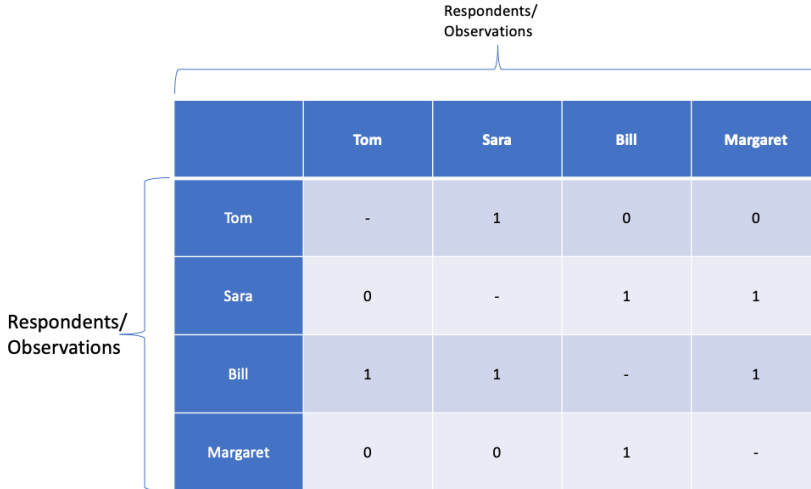
		Variables/attributes				
		ID	Name	Age	Political view	Education
Respondents/ Observations	1	Tom	24	left	NA	
	2	Sara	22	right	BA	
	3	Bill	30	neutral	MA	
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Stories behind ties!

- Independence of observations? (in many cases it is violated!)

		Respondents/ Observations			
		Tom	Sara	Bill	Margaret
Respondents/ Observations	Tom	-	Instagram celebrity	Is not sure remembers	NA
	Sara	NA	-	Brother	Brother's crush
	Bill	(Former) same gym member	Sister	-	MA classmate / friend / smartest batchmate
	Margaret	NA	NA	MA classmate	-

Adjacency (familiarity) matrix



The diagram shows an adjacency matrix with blue headers and light blue data cells. A bracket above the columns is labeled 'Respondents/ Observations'. A bracket to the left of the rows is also labeled 'Respondents/ Observations'.

	Respondents/ Observations			
	Tom	Sara	Bill	Margaret
Tom	-	1	0	0
Sara	0	-	1	1
Bill	1	1	-	1
Margaret	0	0	1	-

Read Edge List as CSV (to construct a network)

```
edge_list2_use <- read_csv("../1_data/humans_ties.csv")  
kable(edge_list2_use)
```

source	target	weight	label
Tom	Sara	0.5	Acquaintance
Sara	Bill	1.0	Sibling
Sara	Margaret	0.5	Acquaintance
Bill	Tom	0.5	Acquaintance
Bill	Sara	1.0	Sibling
Bill	Margaret	1.0	Friend
Margaret	Bill	0.5	Acquaintance

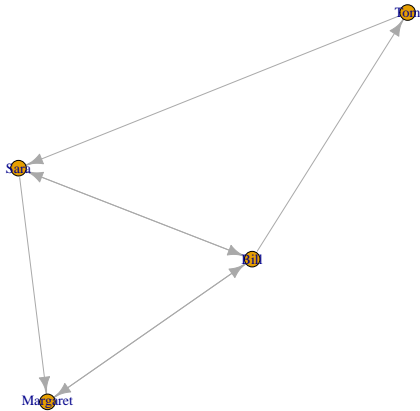
Convert it to a (network) graph object

```
gg = graph_from_data_frame(d = edge_list2_use, directed = TRUE)
print(gg)
```

```
## IGRAPH bd091fb DNW- 4 7 --
## + attr: name (v/c), weight (e/n), label (e/c)
## + edges from bd091fb (vertex names):
## [1] Tom      ->Sara      Sara      ->Bill      Sara      ->Margaret Bill      ->Tom
## [5] Bill     ->Sara      Bill      ->Margaret Margaret->Bill
```

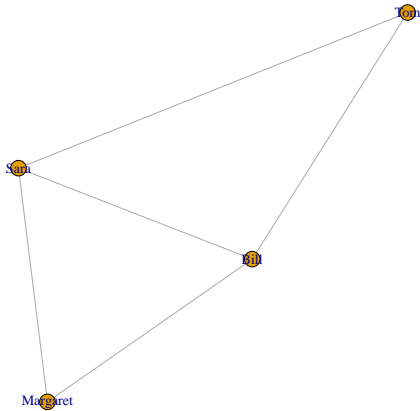
Plot the graph with a layout (directed)

```
set.seed(2225235)
gg_layout = layout.fruchterman.reingold(graph = gg)
plot(gg, layout = gg_layout, edge.label = NA, vertex.size=8)
```



Plot the graph with a layout (un-directed)

```
gg_undirected = graph_from_data_frame(d = edge_list2_use, directed = F)
gg_undirected = simplify(graph = gg_undirected, remove_multiple = T)
plot(gg_undirected, layout = gg_layout, edge.label = NA, vertex.size=8)
```



Add a new attribute to nodes?

```
print(V(gg))
```

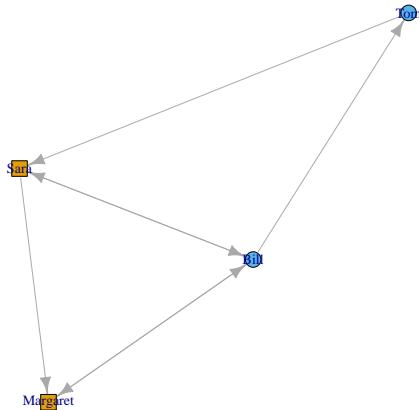
```
## + 4/4 vertices, named, from bd091fb:  
## [1] Tom      Sara      Bill      Margaret
```

```
V(gg)$gender <- c('male', 'female', 'male', 'female')  
V(gg)$shape <- c('circle', 'square', 'circle', 'square')  
print(gg)
```

```
## IGRAPH bd091fb DNW- 4 7 --  
## + attr: name (v/c), gender (v/c), shape (v/c), weight (e/n), label  
## | (e/c)  
## + edges from bd091fb (vertex names):  
## [1] Tom      ->Sara      Sara      ->Bill      Sara      ->Margaret Bill      ->Tom  
## [5] Bill      ->Sara      Bill      ->Margaret Margaret->Bill
```

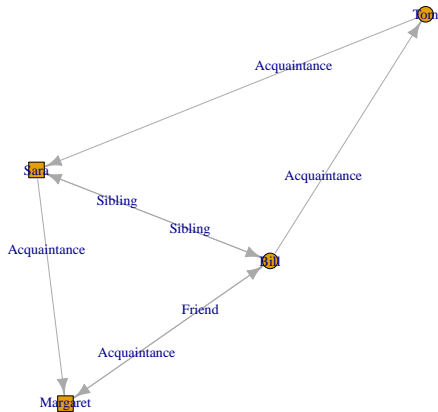
Color and shape of nodes based on gender

```
plot(gg, edge.label = NA, vertex.color = factor(V(gg)$gender),  
     vertex.shape = V(gg)$shape, layout = gg_layout, vertex.size=8)
```



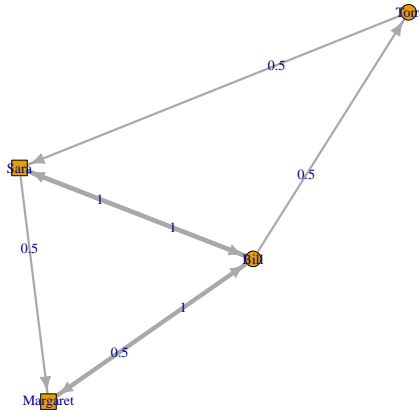
Name ties based on types

```
plot(gg, edge.label = E(gg)$label, layout = gg_layout, vertex.size=8)
```



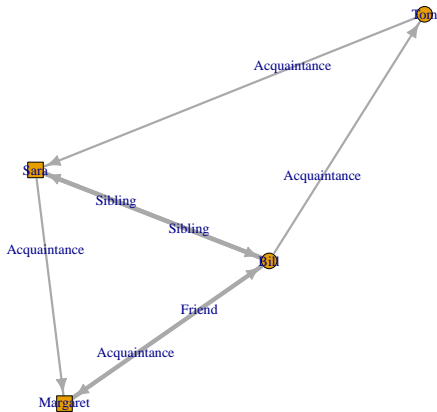
Weight ties based on importance

```
plot(gg, edge.width = E(gg)$weight*5, edge.label = E(gg)$weight, layout = gg_layout, vertex.size=8)
```



Mixture of weight/label

```
plot(gg, edge.label = E(gg)$label, edge.width = E(gg)$weight*5, layout = gg_layout, vertex.size=8)
```

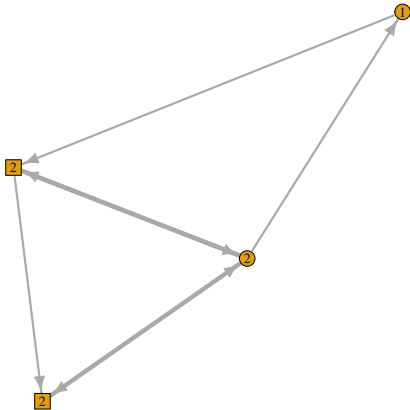


A glimpse to more serious analysis

- After simple visualization (if possible), a five number summary!
 - ① **Size:** V , E (N of vertices/nodes and ties/edges, respectively)
 - ② **Density** (ratio of ties to possible ties, 1 = fully connected)
 - ③ **Components** & (dis)connectivity (more connection inside groups, less among them)
 - ④ **Diameter** (how compact the network is?)
 - ⑤ **Clustering Coefficient** (transitivity and triangles)
- **Centrality** in network (different measures of importance in structure)
 - Degree, Closeness, Betweenness, Eigenvector, ...

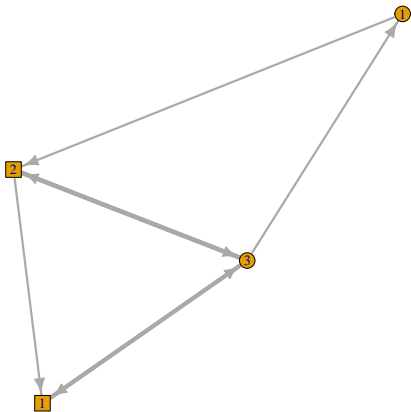
In-degree of a node (incoming ties)

```
plot(gg, edge.label = NA, edge.width = E(gg)$weight*5,  
     vertex.label = degree(gg, mode = 'in'), layout = gg_layout, vertex.size=8)
```



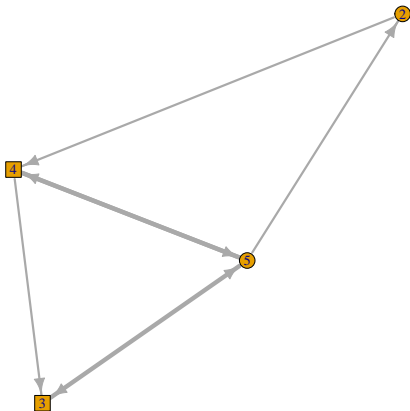
Out-degree of a node (outgoing ties)

```
plot(gg, edge.label = NA, edge.width = E(gg)$weight*5,  
     vertex.label = degree(gg, mode = 'out'), layout = gg_layout, vertex.size=8)
```



Degree of a node (both incoming/outgoing ties)

```
plot(gg, edge.label = NA, edge.width = E(gg)$weight*5,  
     vertex.label = degree(gg, mode = 'all'), layout = gg_layout, vertex.size=8)
```

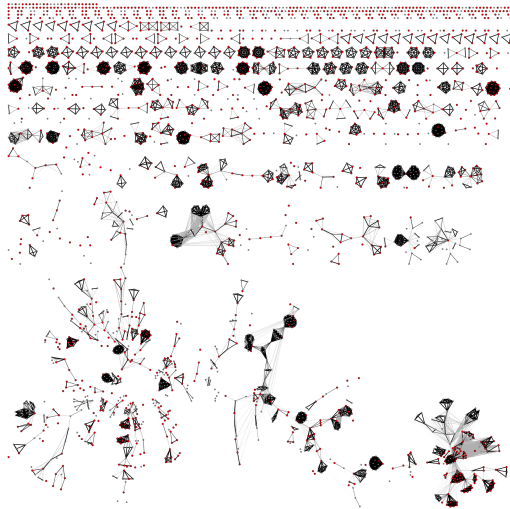


Sociological theories (& SNA conceptualization)¹

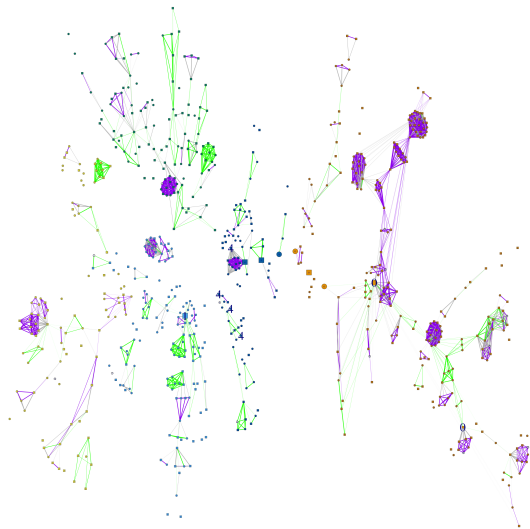
- **Matteo effect**, winner takes all?
 - Highly prolific scientists attract higher collaborations from other scientists?
 - Attaching preferably to a few **star scientists**/leaders?
- **Fragmentation** of ideas, sociology as a interstitial science?
 - Methodologists bridging the islands?
- [Sociological] **small world** of disconnected islands?
- **Core** of leaders and **periphery** of followers?

¹Akbaritabar, A., Traag, V. A., Caimo, A., & Squazzoni, F. (2020). Italian Sociologists: A Community of Disconnected Groups. *Scientometrics*. <https://doi.org/10.1007/s11192-020-03555-w>

Coauthorship of Italian sociologists



Communities in the giant component



What can we learn from these communities? (1/2)²

Table 2: Gender composition and internationality of members of the communities detected from the giant component (Percentages are calculated by rows separately for gender and country)

Community	# members	Gender			Country			
		Female	Male	Missing Gender	Europe	Italy	Other	Missing Country
0	254	43%	54%	3%	54%	29%	11%	5%
1	142	50%	49%	1%	36%	55%	6%	3%
2	122	38%	61%	1%	37%	56%	3%	4%
3	103	45%	54%	1%	41%	44%	5%	11%
4	91	47%	49%	3%	32%	57%	9%	2%

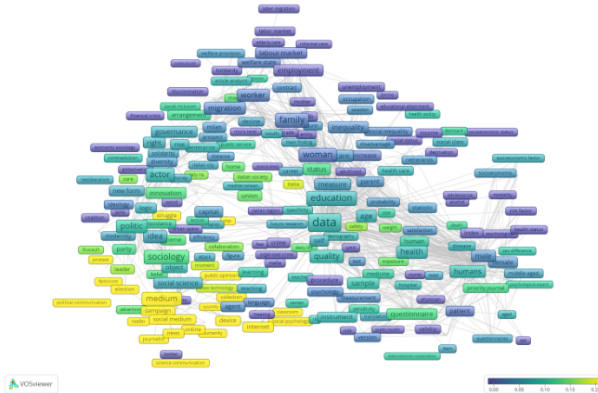
Table 3: Sectors composition of members of the communities detected from the giant component (Percentages are calculated by rows)

Community	# members	Scientific Disciplinary Sectors						
		postdoc	SPS/07	SPS/08	SPS/09	SPS/10	SPS/11	Missing Sector
0	254	2%	1%	5%	0	0%	0%	91%
1	142	2%	6%	3%	8%	1%	1%	78%
2	122	5%	10%	1%	7%	0	1%	76%
3	103	2%	4%	2%	12%	1%	0	80%
4	91	1%	7%	7%	0	1%	2%	82%

²Akbaritabar, A., Traag, V. A., Caimo, A., & Squazzoni, F. (2020). Italian Sociologists: A Community of Disconnected Groups. *Scientometrics*. <https://doi.org/10.1007/s11192-020-03555-w>

What can we learn from these communities? (2/2)³

- 65% foreigners
- Medium, science communication, social medium, internet, political communication & public opinion



³ Akbaritabar, A., Traag, V. A., Caimo, A., & Squazzoni, F. (2020). Italian Sociologists: A Community of Disconnected Groups. *Scientometrics*. <https://doi.org/10.1007/s11192-020-03555-w>

Where to next?!

- **Awesome network analysis list:** <https://github.com/briatte/awesome-network-analysis>

