

**Sept. 14**

- 1. Administrative**
- 2. Network terminology and representations**
- 3. Network theory**

## Lab 1 due Sept 28

- ⋮ First lab is due next moday
- ⋮ Linked from the syllabus  
(<https://soci424.netlify.app/labs/lab1.html>)

## Help sessions

- ⋮ Help sessions scheduled will be  
*Thursdays from 10:30-11:30am*
- ⋮ Scheduled work sessions?

# Network Terminology & Representation

# What is a social network?

**A set of “actors”**  
(i.e. people, orgs, ...)

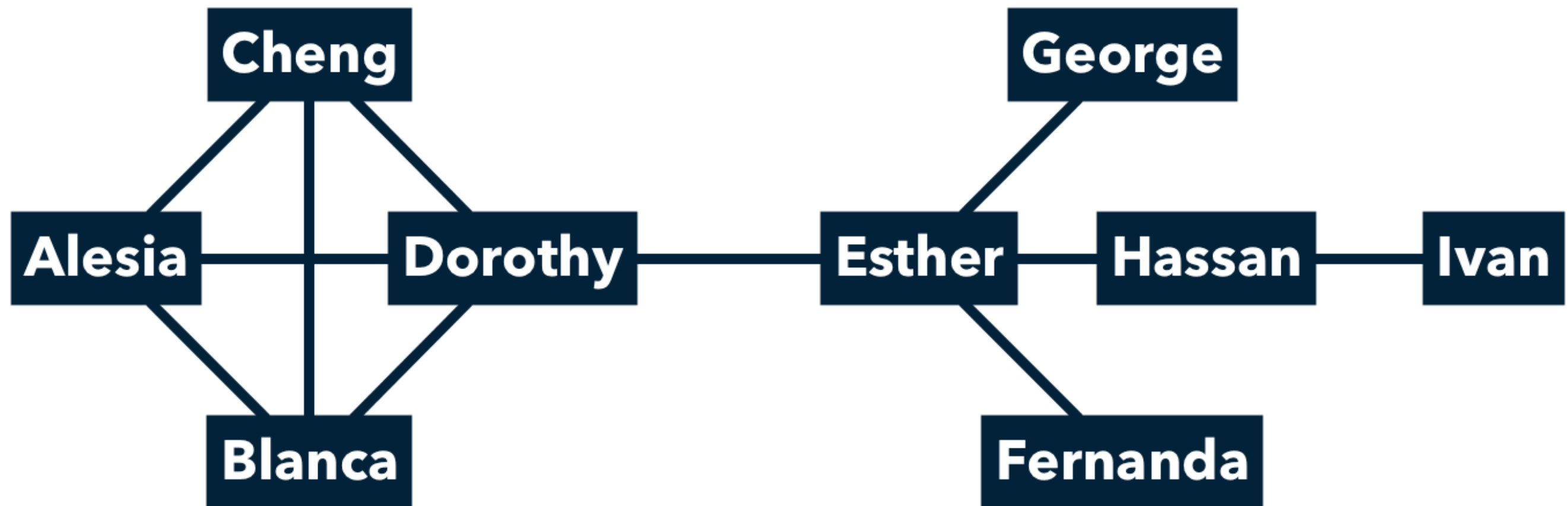


**And a set of “ties”**  
(i.e. friendship, payment, ...)



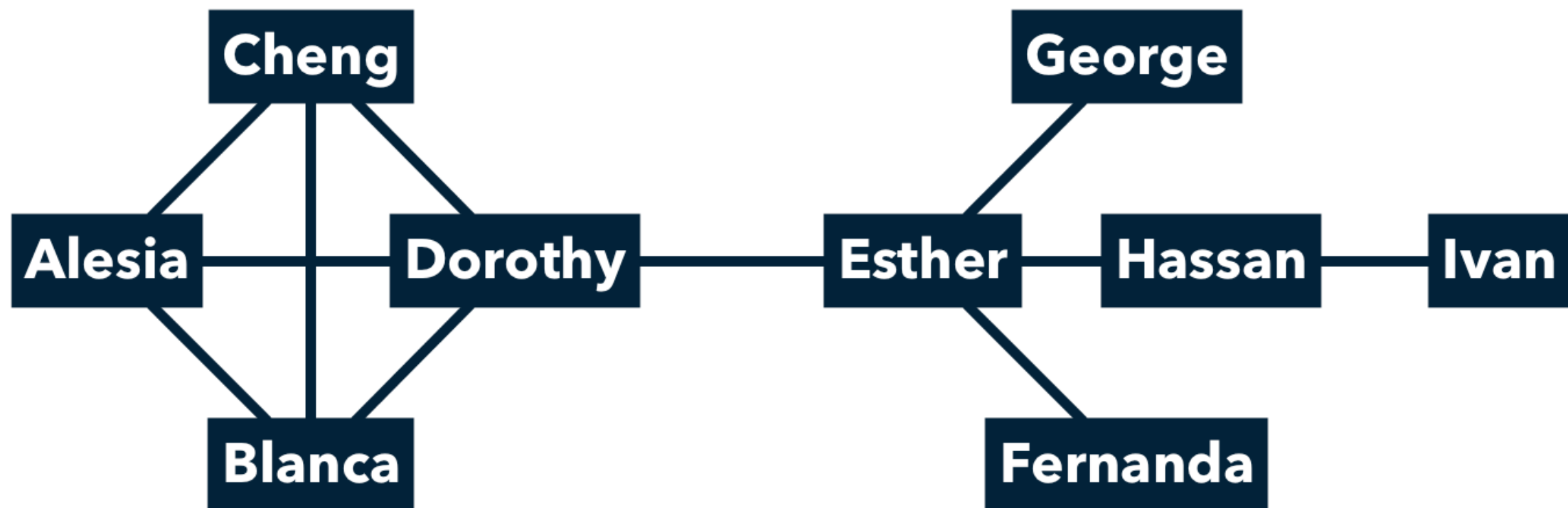
# What is a social network?

Putting these together gives us a “network” picture



# Network mini-glossary

- | **Node / vertex / actor:** A single person, organization, etc.
- | **Edge / tie / relation / arc:** A link between two nodes
- | **Ego:** A focal node
- | **Alter:** Anyone connected to ego
- | **Path:** A chain of nodes connected by edges (usually: no repeats)
- | **Cluster:** A subset of nodes that are tightly tied to each other



# Network representations

## Graph visualizations

### Intuitive

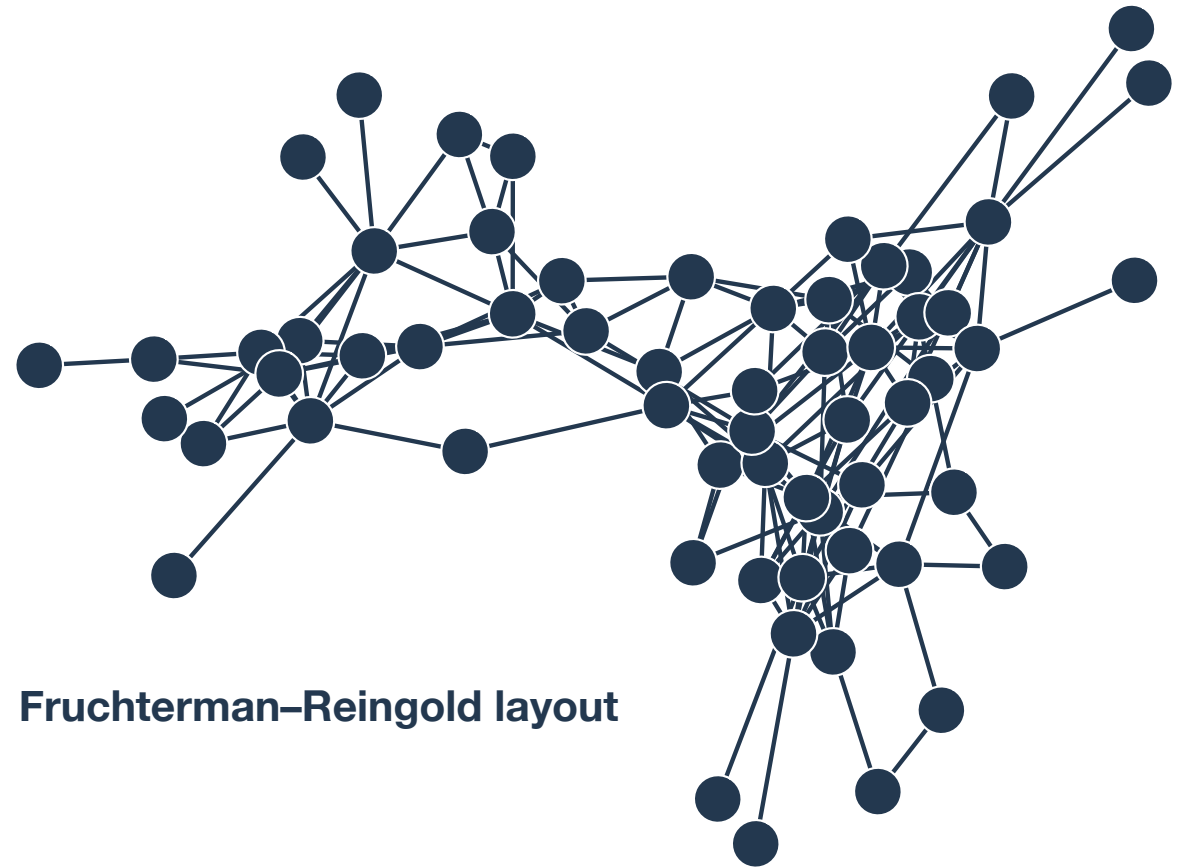
- ⋮ Easy to understand!
- ⋮ Circles connected by lines don't require much explanation

### Descriptive

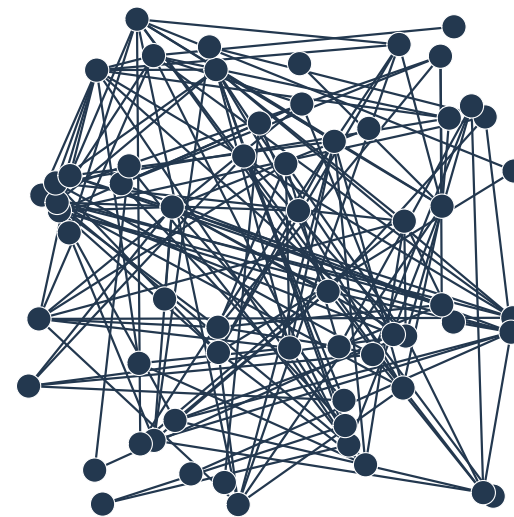
- ⋮ Easily gives an idea of the size of a network, overall density of relations, etc.
- ⋮ Can suggest important structure

### Can be deceptive!

- ⋮ Graph visualizations use a large number of heuristics to get a picture that “looks good.”
- ⋮ Different heuristics and different runs of the same heuristic can tell diverging stories



Fruchterman-Reingold layout



Random layout



Multidimensional scaling

# Network representations

## Adjacency matrices

### Mathematically convenient

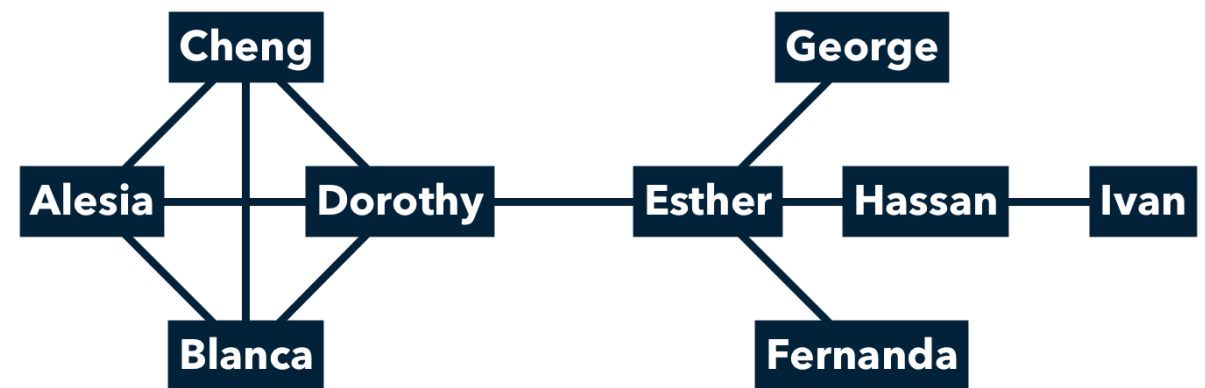
- ⋮ Tool borrowed from formal graph theory
- ⋮ Allows for analysis using the branch of mathematics called linear algebra

### Computationally convenient

- ⋮ Computers are very good at working with adjacency matrices (unless they get very big)
- ⋮ Easy to perform simple measurements and manipulations

### Looks intimidating

- ⋮ Can look overwhelming for those without a background in math or computer science

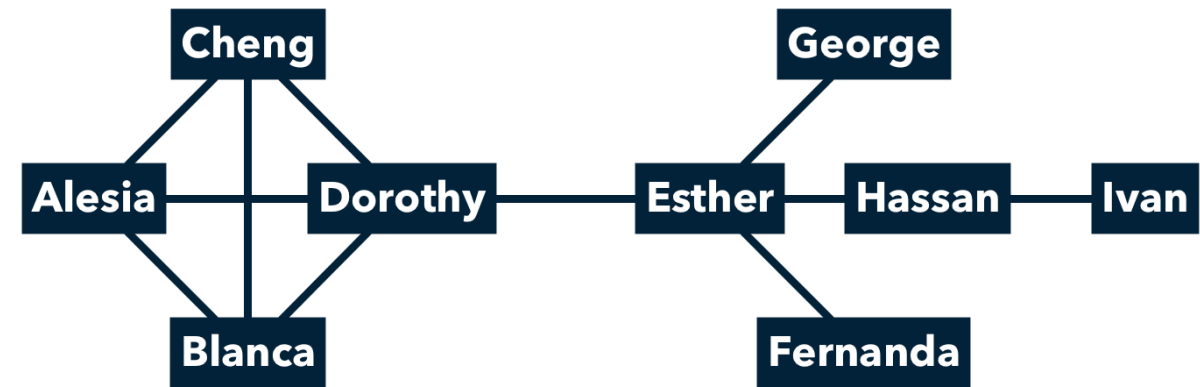


0	1	1	1	0	0	0	0	0
1	0	1	1	0	0	0	0	0
1	1	0	1	0	0	0	0	0
1	1	1	0	1	0	0	0	0
0	0	0	1	0	1	1	1	0
0	0	0	0	1	0	0	0	0
0	0	0	0	1	0	0	0	0
0	0	0	0	1	0	0	0	1
0	0	0	0	0	0	0	1	0



# Network representations

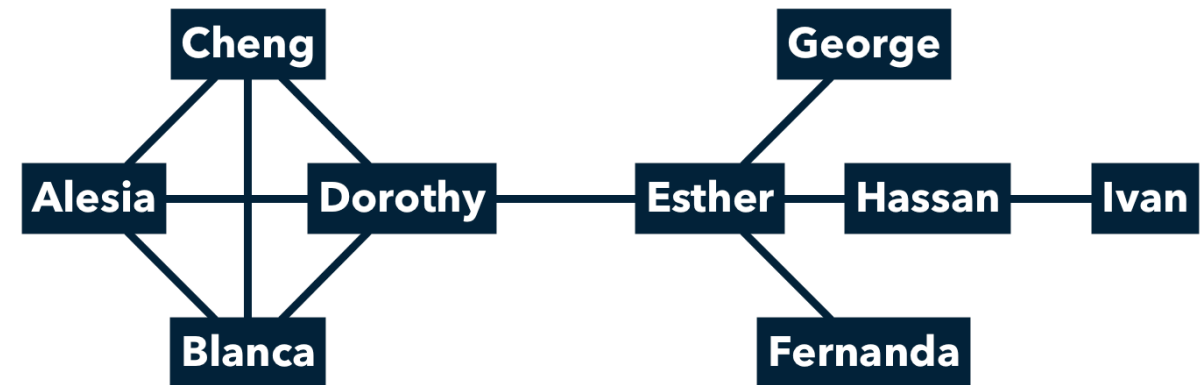
## Reading adjacency matrices



	A	B	C	D	E	F	G	H	I
Alesia	0	1	1	1	0	0	0	0	0
Blanca	1	0	1	1	0	0	0	0	0
Cheng	1	1	0	1	0	0	0	0	0
Dorothy	1	1	1	0	1	0	0	0	0
Esther	0	0	0	1	0	1	1	1	0
Fernanda	0	0	0	0	1	0	0	0	0
George	0	0	0	0	1	0	0	0	0
Hassan	0	0	0	0	1	0	0	0	1
Ivan	0	0	0	0	0	0	0	1	0

# Network representations

## Reading adjacency matrices

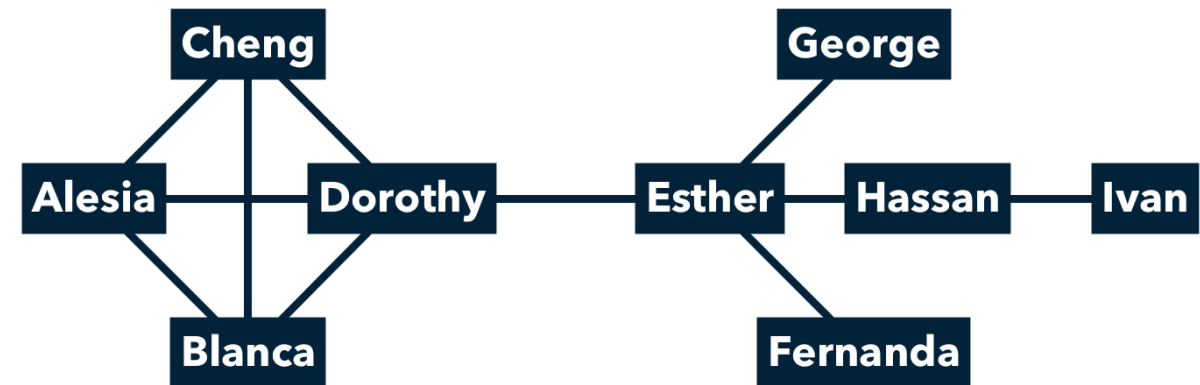


	A	B	C	D	E	F	G	H	I
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Dorothy	1	1	1	0	1	0	0	0	0
Esther	0	0	0	1	0	1	1	1	0
Fernanda	0	0	0	0	1	0	0	0	0
George	0	0	0	0	1	0	0	0	0
Hassan	0	0	0	0	1	0	0	0	1
Ivan	0	0	0	0	0	0	0	1	0

Esther is friends with Dorothy, Fernanda, George, and Hassan

# Network representations

## Reading adjacency matrices

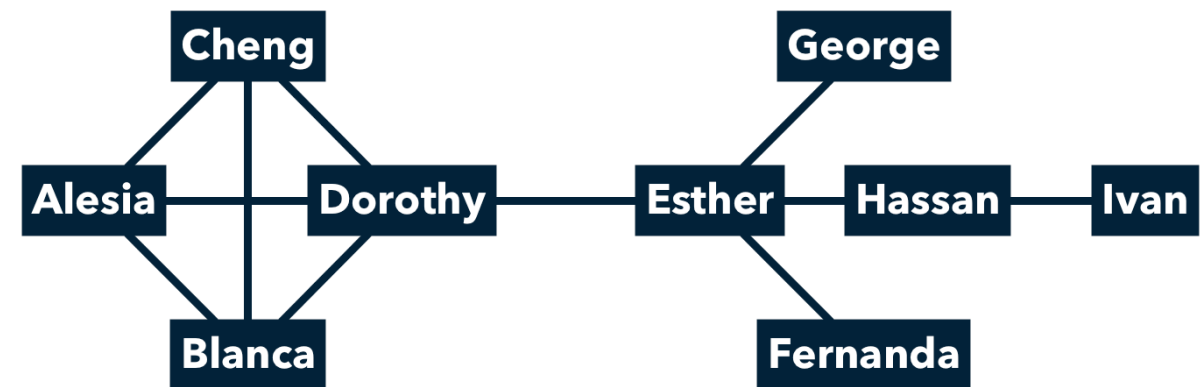


	A	B	C	D	E	F	G	H	I
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Blanca	1	0	1	1	0	0	0	0	0
Cheng	1	1	0	1	0	0	0	0	0
Dorothy	1	1	1	0	1	0	0	0	0
Esther	0	0	0	1	0	1	1	1	0
Fernanda	0	0	0	0	1	0	0	0	0
George	0	0	0	0	1	0	0	0	0
Hassan	0	0	0	0	1	0	0	0	1
Ivan	0	0	0	0	0	0	0	1	0

The *diagonal* of the matrix shows self-relationships (often all zeros)

# Network representations

## Reading adjacency matrices



	A	B	C	D	E	F	G	H	I	
Alesia	0	1	1	1	0	0	0	0	0	3
Blanca	1	0	1	1	0	0	0	0	0	3
Cheng	1	1	0	1	0	0	0	0	0	3
Dorothy	1	1	1	0	1	0	0	0	0	4
Esther	0	0	0	1	0	1	1	1	0	4
Fernanda	0	0	0	0	1	0	0	0	0	1
George	0	0	0	0	1	0	0	0	0	1
Hassan	0	0	0	0	1	0	0	0	1	2
Ivan	0	0	0	0	0	0	0	1	0	1

Easy to see how many friends each person has with row or column sums

# Network representations

Adjacency matrices are closely related to affiliation matrices like the one from this week's lab

NAMES OF PARTICIPANTS OF GROUP I	CODE NUMBERS AND DATES OF SOCIAL EVENTS REPORTED IN <i>Old City Herald</i>													
	(1) 6/27	(2) 3/2	(3) 4/12	(4) 9/26	(5) 2/25	(6) 5/19	(7) 3/15	(8) 9/16	(9) 4/8	(10) 6/10	(11) 2/23	(12) 4/7	(13) 11/21	(14) 8/3
1. Mrs. Evelyn Jefferson.....	×	×	×	×	×	×	....	×	×	....	....	....	....	....
2. Miss Laura Mandeville.....	×	×	×	....	×	×	×	×	....	....	....	....	....	....
3. Miss Theresa Anderson.....	....	×	×	×	×	×	×	×	×	....	....	....	....	....
4. Miss Brenda Rogers.....	×	....	×	×	×	×	×	×	....	....	....	....	....	....
5. Miss Charlotte McDowd.....	....	....	×	×	×	....	×	....	....	....	....	....	....	....
6. Miss Frances Anderson.....	....	....	×	....	×	×	....	×	....	....	....	....	....	....
7. Miss Eleanor Nye.....	....	....	....	....	×	×	×	×	....	....	....	....	....	....
8. Miss Pearl Oglethorpe.....	....	....	....	....	....	×	....	×	×	....	....	....	....	....
9. Miss Ruth DeSand.....	....	....	....	....	×	....	×	×	×	....	....	....	....	....
10. Miss Verne Sanderson.....	....	....	....	....	....	....	×	×	×	....	....	×	....	....
11. Miss Myra Liddell.....	....	....	....	....	....	....	....	×	×	×	....	×	....	....
12. Miss Katherine Rogers.....	....	....	....	....	....	....	....	×	×	×	....	×	×	×
13. Mrs. Sylvia Avondale.....	....	....	....	....	....	....	×	×	×	×	....	×	×	×
14. Mrs. Nora Fayette.....	....	....	....	....	....	×	×	....	×	×	×	×	×	×
15. Mrs. Helen Lloyd.....	....	....	....	....	....	....	×	×	....	×	×	×	....	....
16. Mrs. Dorothy Murchison.....	....	....	....	....	....	....	....	×	×	....	....	....	....	....
17. Mrs. Olivia Carleton.....	....	....	....	....	....	....	....	....	×	....	×	....	....	....
18. Mrs. Flora Price.....	....	....	....	....	....	....	....	....	×	....	×	....	....	....

FIG. 3.—Frequency of interparticipation of a group of women in Old City, 1936—Group I.

# Network Theory

# What is a network tie?

## What counts as a tie?

- | At its broadest, a tie is any kind of relation between actors
- | Many network scholars focus on *social* ties (*relationships* rather than just *relations*)



## Tie characteristics

- | *Events vs states*
- | *Directed vs undirected*  
(asymmetric vs antisymmetric vs semetric)
- | *Valued vs binary*  
(weights and other attributes)



# Network representations

## Borgatti and Halgin (2011) on *network theory*

### Two consistent traits of network theories

- ∴ Focus on *structure* and *position* as causal elements
- ∴ Implicit theories of what a network *does*

### Networks allow *flow*

- ∴ One view of what networks do is act as pipes that transmit information, money, contagions, behavior, etc.
- ∴ Rarely stated, but implicit in the vast majority of network analysis
- ∴ Strength of weak ties (Granovetter) and structural holes (Burt)

### Networks form *bonds*

- ∴ A long-running but less common theorization holds that network ties define us
- ∴ E.g. managers are defined by relationships of authority over others
- ∴ E.g. being followed by a celebrity on social media can grant status
- ∴ Networks are prisms (Podolny) that affect how we are seen

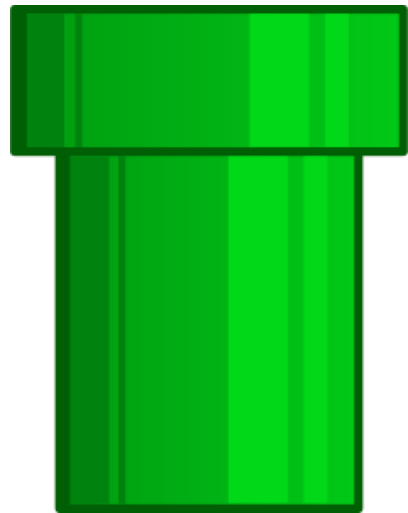


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