

Network Terminology & Representation

What is a social network?

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

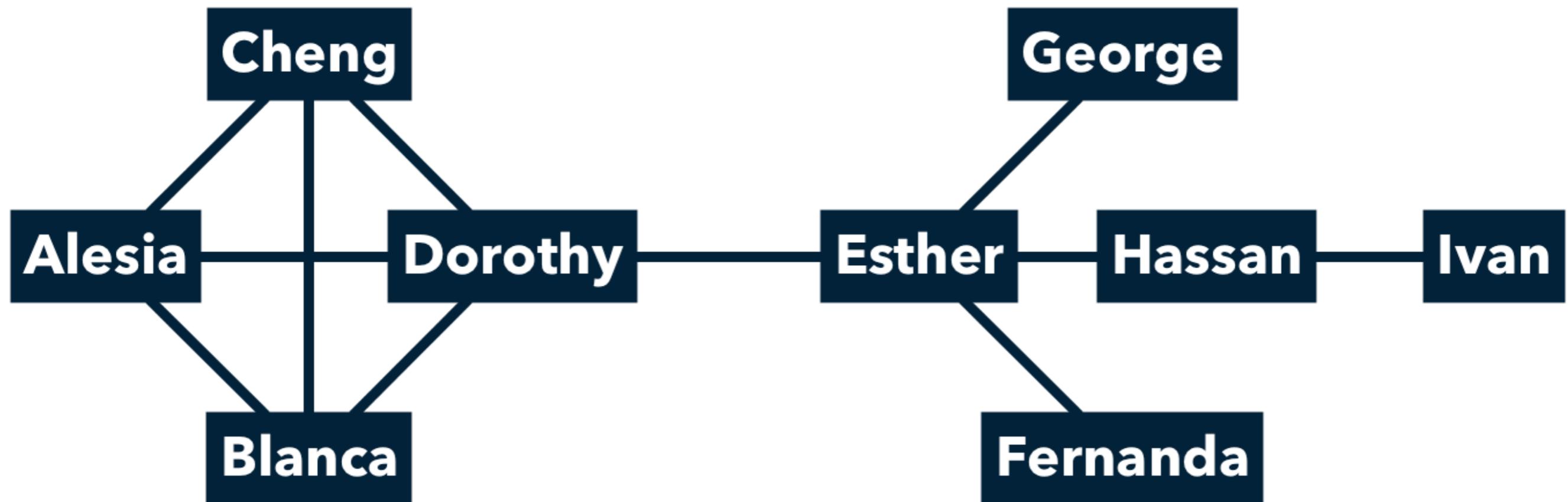


And a set of “relations”
(i.e. friendship, payment, ...)



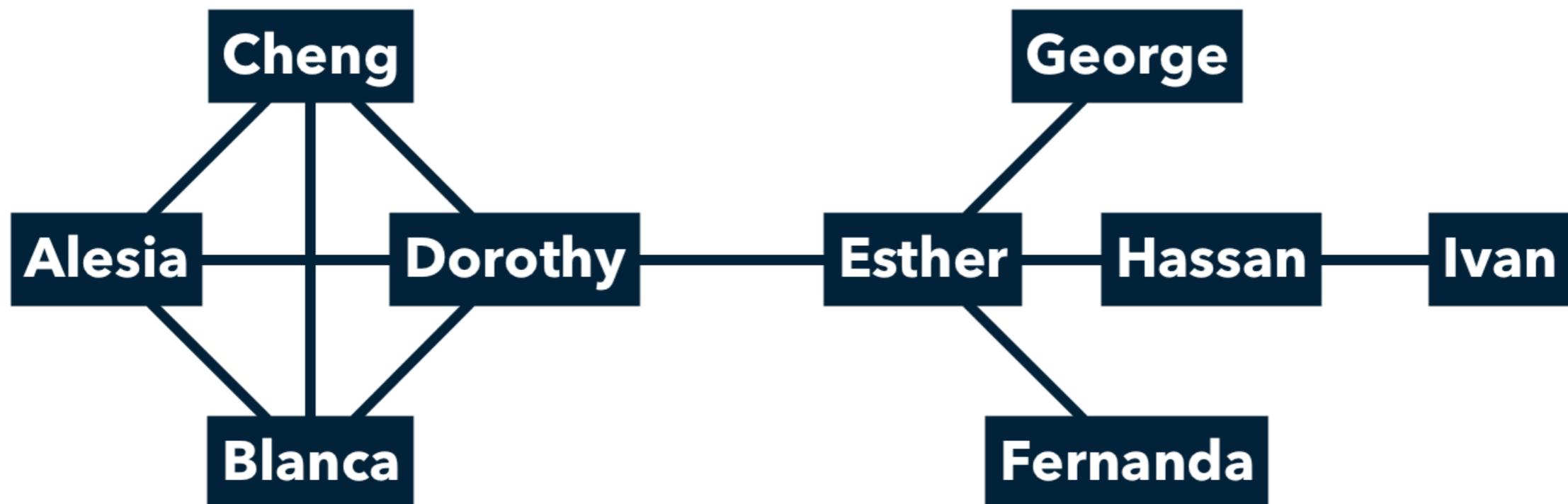
What is a social network?

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

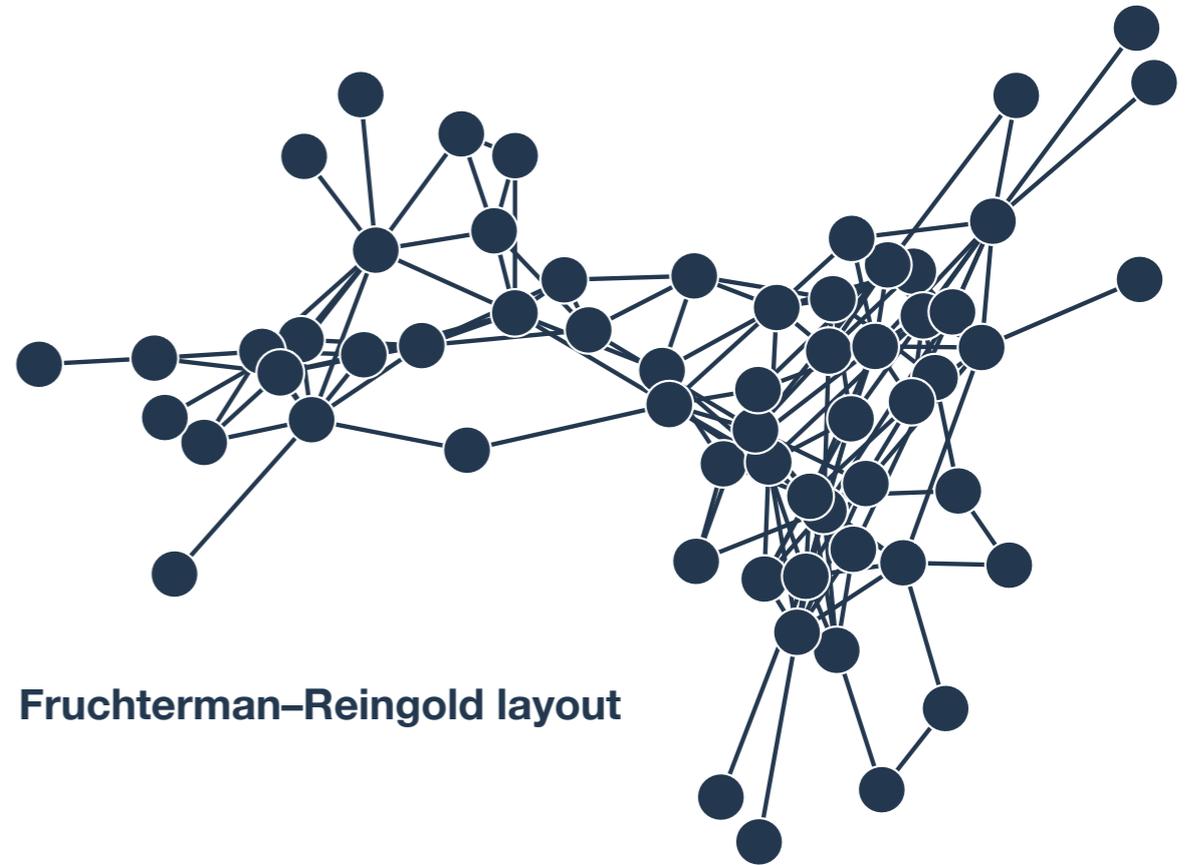
- ∴ Easy to understand!
- ∴ Circles connected by lines don't require much explanation

Descriptive

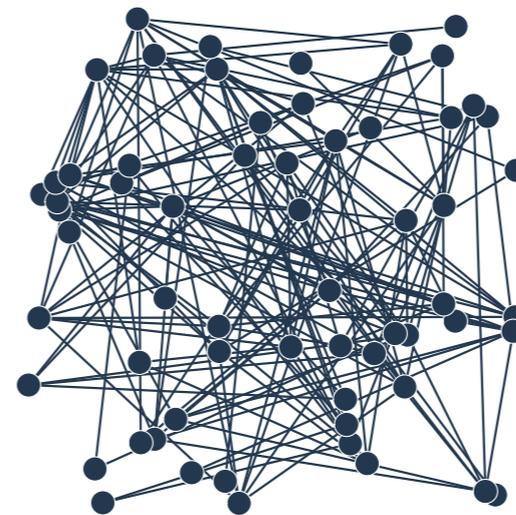
- ∴ Easily gives an idea of the size of a network, overall density of relations, etc.
- ∴ Can suggest important structure

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

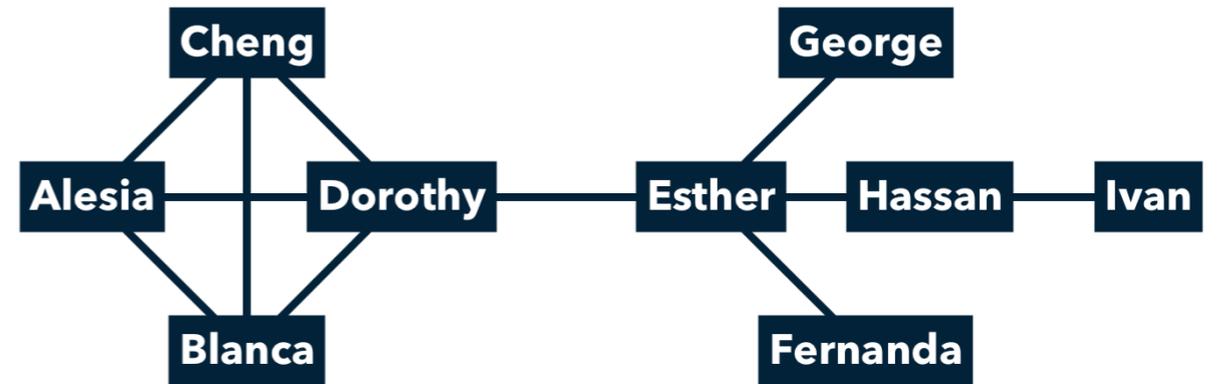
- ∴ Tool borrowed from formal graph theory
- ∴ Allows for analysis (and theorization!) using the branch of mathematics called linear algebra

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

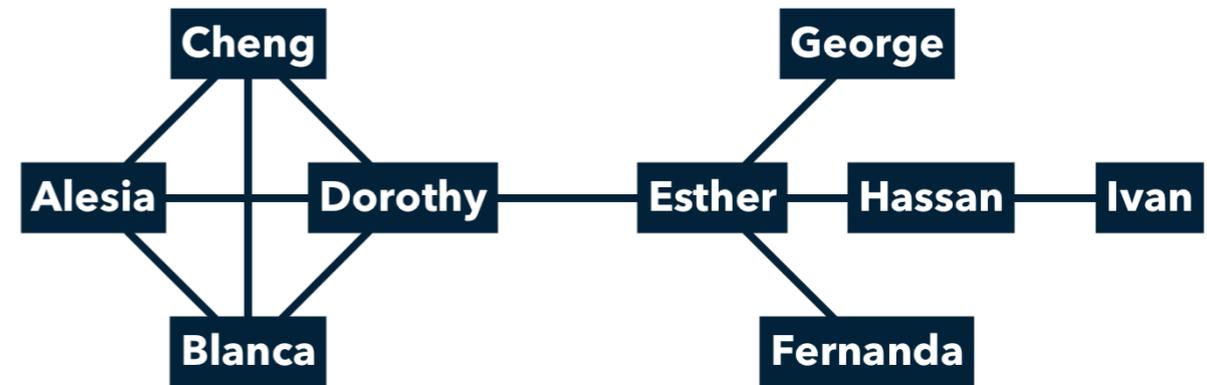
- ∴ Can look overwhelming for those without a background in math or computer science



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

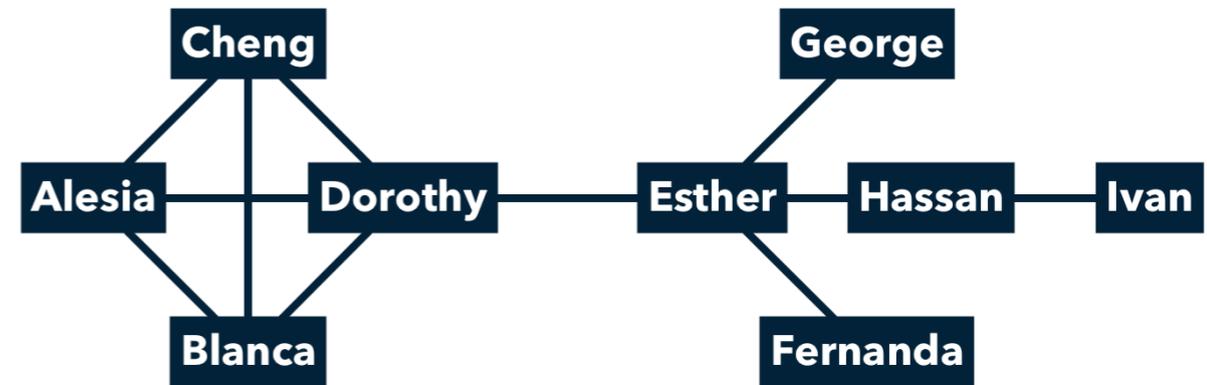
Reading adjacency matrices



	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

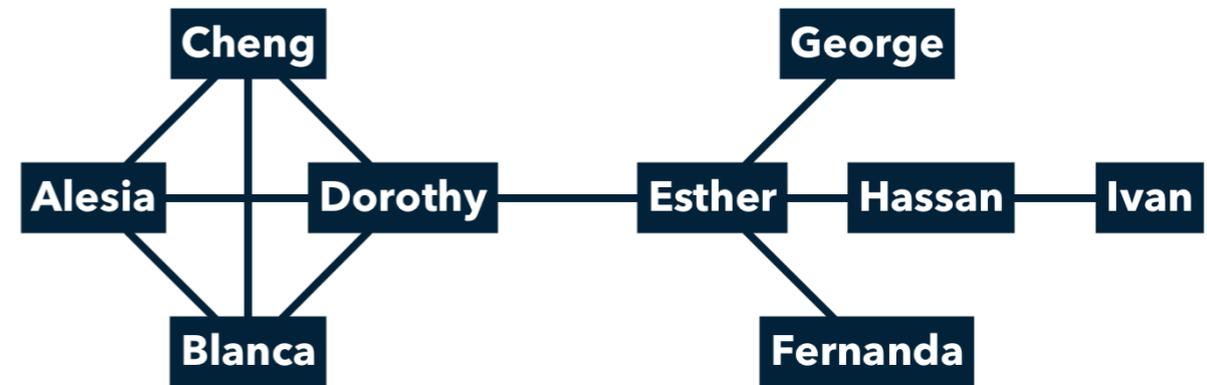


	A	B	C	D	E	F	G	H	I
Alesia	0	1	1	1	0	0	0	0	0
Blanca	1	0	1	1	0	0	0	0	0
Cheng	1	1	0	1	0	0	0	0	0
Dorothy	1	1	1	0	1	0	0	0	0
Esther	0	0	0	1	0	1	1	1	0
Fernanda	0	0	0	0	1	0	0	0	0
George	0	0	0	0	1	0	0	0	0
Hassan	0	0	0	0	1	0	0	0	1
Ivan	0	0	0	0	0	0	0	1	0

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

Network representations

Reading adjacency matrices

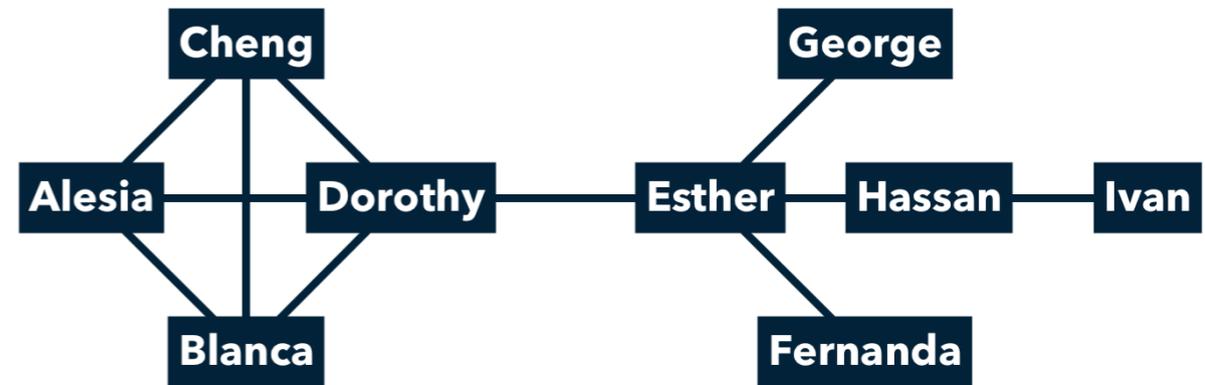


	A	B	C	D	E	F	G	H	I
Alesia	0	1	1	1	0	0	0	0	0
Blanca	1	0	1	1	0	0	0	0	0
Cheng	1	1	0	1	0	0	0	0	0
Dorothy	1	1	1	0	1	0	0	0	0
Esther	0	0	0	1	0	1	1	1	0
Fernanda	0	0	0	0	1	0	0	0	0
George	0	0	0	0	1	0	0	0	0
Hassan	0	0	0	0	1	0	0	0	1
Ivan	0	0	0	0	0	0	0	1	0

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 worksheet

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.....	X	X	X	X	X	X	X	X
2. Miss Laura Mandeville.....	X	X	X	X	X	X	X
3. Miss Theresa Anderson.....	X	X	X	X	X	X	X	X
4. Miss Brenda Rogers.....	X	X	X	X	X	X	X
5. Miss Charlotte McDowd.....	X	X	X	X
6. Miss Frances Anderson.....	X	X	X	X
7. Miss Eleanor Nye.....	X	X	X	X
8. Miss Pearl Ogleshorpe.....	X	X	X
9. Miss Ruth DeSand.....	X	X	X	X
10. Miss Verne Sanderson.....	X	X	X	X
11. Miss Myra Liddell.....	X	X	X	X
12. Miss Katherine Rogers.....	X	X	X	X	X	X
13. Mrs. Sylvia Avondale.....	X	X	X	X	X	X	X
14. Mrs. Nora Fayette.....	X	X	X	X	X	X	X	X
15. Mrs. Helen Lloyd.....	X	X	X	X	X
16. Mrs. Dorothy Murchison.....	X	X
17. Mrs. Olivia Carleton.....	X	X
18. Mrs. Flora Price.....	X	X

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 symmetric)

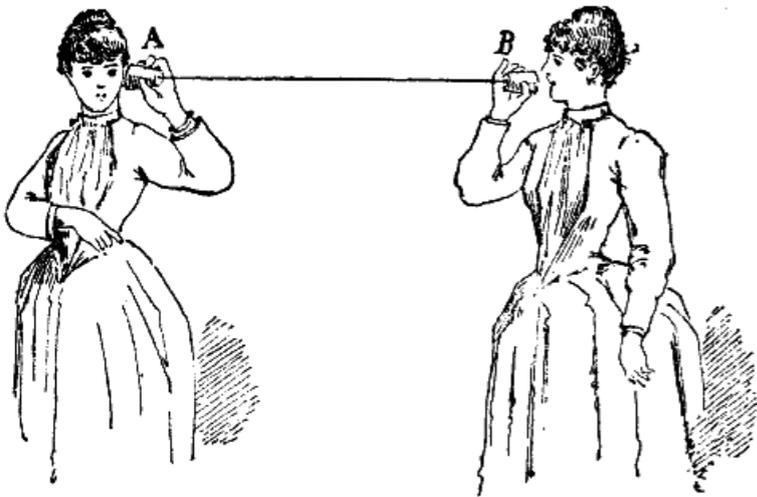
| *Valued vs binary*
(weights and other attributes)

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, norms, etc.
- ∴ Rarely stated, but implicit in the large majority of network analysis
- ∴ E.g. Strength of weak ties (Granovetter) and structural holes (Burt)



Networks reflect *bonds*

- ∴ A long-running (but somewhat less common) theorization holds that network ties define us (our interests, capabilities, identities)
- ∴ 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 and how we see ourselves

Discussion

Image credit



Screenshot from [I Love Lucy \(1951\)](#), via [anthonybalduccisjournal.wordpress.com](#)

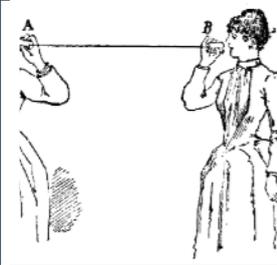


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