

Graph Theory and Complex Networks: An Introduction

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Chapter 09: Social networks

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Introduction

Observation

Sociologists have always been interested in social structures:

- formation of groups
- influence relationships
- ties of families and friends
- (dis)likings in groups of people

Observation

Graphs form a natural way for modeling social structures

- Sociograms and blockmodeling
- Basic concepts: balance, cohesiveness, affiliation networks
- Equivalence

Example: The influence of the Medici's

Observation

The **Strozzi family** was richer and had more representatives in the local legislature. Yet the Medici's power surpassed that of the Strozzi's.

Reconsider the **betweenness centrality**:

$$c_B(u) = \sum_{x \neq y \neq u} \frac{|S(x, u, y)|}{|S(x, y)|}$$

with

- $S(x, u, y)$ is collection of shortest (x, y) paths containing u
- $S(x, y)$ is set of shortest paths between vertices x and y .

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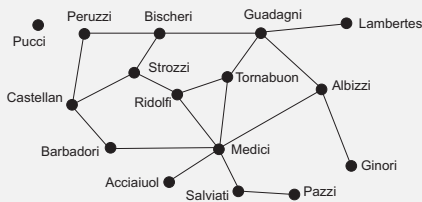
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Example: The influence of the Medici's

Normalization

Normalize $c_B(u)$ by the maximum possible pairs of families that u can connect: $\binom{n-1}{2}$

$c_B(\text{Medici}) = 0.522$ whereas $c_B(\text{Strozzi}) = 0.103$



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Starters: sociograms

History

Already early in the 1930s Jacob Moreno introduced graph-like representations for social structures and suggested that they could be used for discovering new features.

Sociograms in the classroom

In order to get an impression of how a class operates, teachers can ask their pupils to list the three classmates they (dis)like the most.

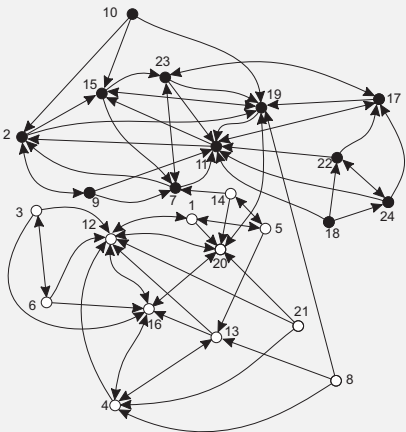
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Example classroom sociogram

Sex	ID	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
F	1	■				+				-	-		+								+	+			
M	2	-	■							+	-				+						+		-	-	
F	3		■	■				+	-					+			+								
F	4			■	■									+											
F	5	+			■	■								+	+						-	-			
F	6	-	+										+												
M	7		+					■	■				+								-	-		+	
F	8			+				■	■				+								-	+			-
M	9		+					■	■			+													-
M	10		+					■	■			+									+				-
M	11		+					■	■			+									+				-
F	12	+						■	■			+									+				-
F	13				+							+	■	■							+				-
F	14				+	-	+					+	■	■							+				-
M	15					+						+	■	■							+				-
F	16				+							+	■	■							+				-
M	17											+	■	■							+				+
M	18											+	■	■							+				+
M	19											+	■	■							+				+
F	20											+	■	■							+				+
F	21	-	-	+								+	■	■							+				+
M	22											+	■	■							+				+
M	23	-										+	■	■							+				+
M	24											+	■	■							+				+
	+	2	4	1	4	2	1	4	0	1	0	8	8	3	1	4	6	3	0	7	6	0	2	3	2
	-	4	2	0	1	0	4	4	0	4	9	1	1	1	2	3	1	2	0	7	6	10	4	3	3

Classroom example - positive nominations

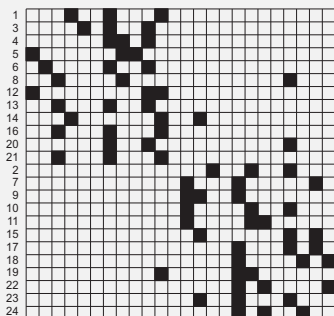


- Clear distinction between boys ("●") and girls ("○")
- Relation between 19 and 20 is important
- There are a few "isolated" children (8 & 10)

Issue
Can we discover these properties **mathematically**?

Blockmodeling

Essence: reorder the rows and columns in the adjacency matrix in order to discover **subgroups**. Can be done automatically (and is then called **clustering**).



Concentrate on SCC (largest Strongly Connected Component)

Eccentricity

Recall: Eccentricity u is maximal minimal distance to other vertices

Child:	1	2	4	5	7	9	11	12
Ecc.:	5	6	6	4	7	7	7	5
Child:	13	14	15	16	17	19	20	23
Ecc.:	6	3	6	5	6	5	4	6

Observations

Child #14 is one of the few nominating a boy *and* a girl. She also seems to be "in the middle."

Concentrate on SCC

Closeness

Recall: $c_C(u) = \frac{1}{\sum_{v \in V(G)} d(u,v)}$

Child:	1	2	4	5	7	9	11	12
Close:	.23	.21	.18	.25	.18	.18	.18	.22
Child:	13	14	15	16	17	19	20	23
Close:	.18	.30	.21	.21	.21	.25	.25	.21

Observation

The closeness confirms that child #14 is close to **everyone**.

Concentrate on SCC

Betweenness

Child:	1	2	4	5	7	9	11	12
Betw.:	.140	.153	.050	.105	.083	.007	.155	.220
Child:	13	14	15	16	17	19	20	23
Betw.:	.016	.054	.083	.140	.017	.466	.469	.029

Observation

The picture has changed dramatically: child #14 may be close, but her importance should be questioned.

Metrics already discussed

Definition (Vertex centrality)

G is (strongly) connected. The **vertex centrality**:

$$c_E(u) = 1 / \max\{d(u, v) | v \in V(G)\}$$

Definition (Closeness)

G is (strongly) connected. The **closeness**: $c_C(u) = 1 / \sum_{v \in V(G)} d(u, v)$

Definition (Betweenness)

G is simple and (strongly) connected. $S(x, y)$ is set of shortest paths between x and y . $S(x, u, y) \subseteq S(x, y)$ paths that pass through u .

Betweenness centrality: $c_B(u) = \sum_{x \neq y \neq u} \frac{|S(x, u, y)|}{|S(x, y)|}$.

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Prestige

Definition (Degree prestige)

Let D be a directed graph. The **degree prestige** $p_{deg}(v)$ of a vertex $v \in V(D)$ is defined as its indegree $\delta^-(v)$.

Definition (Proximity prestige)

Let D be a directed graph with n vertices. The **influence domain** $R^-(v)$ is the set of vertices from where v can be reached through a directed path, that is, $R^-(v) = \{u \in V(D) | \exists (u, v)\text{-path}\}$. The **proximity prestige**:

$$p_{prox}(v) = \frac{|R^-(v)| / (n-1)}{\sum_{u \in R^-(v)} d(u, v) / |R^-(v)|}$$

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Ranked prestige

Definition

Consider a simple directed graph D with vertex set $\{1, 2, \dots, n\}$ with adjacency matrix \mathbf{A} . The **ranked prestige** of a vertex k is:

$$p_{rank}(k) = \sum_{i=1, i \neq k}^n \mathbf{A}[i, k] \cdot p_{rank}(i)$$

Simple example

ID	A	B	C
A	—	0.5	0.4
B	0.1	—	0.6
C	0.9	0.5	—

$$\begin{aligned} p_{rank}(A) &= 0.5 \cdot p_{rank}(B) + 0.4 \cdot p_{rank}(C) \\ p_{rank}(B) &= 0.1 \cdot p_{rank}(A) + 0.6 \cdot p_{rank}(C) \\ p_{rank}(C) &= 0.9 \cdot p_{rank}(A) + 0.5 \cdot p_{rank}(B) \end{aligned}$$

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Computing ranked prestige

Some simple rewriting

$$\begin{aligned} p_{rank}(A) &= 0.5 \cdot p_{rank}(B) + 0.4 \cdot p_{rank}(C) & x &= 0.5 \cdot y + 0.4 \cdot z & (1) \\ p_{rank}(B) &= 0.1 \cdot p_{rank}(A) + 0.6 \cdot p_{rank}(C) & y &= 0.1 \cdot x + 0.6 \cdot z & (2) \\ p_{rank}(C) &= 0.9 \cdot p_{rank}(A) + 0.5 \cdot p_{rank}(B) & z &= 0.9 \cdot x + 0.5 \cdot y & (3) \end{aligned}$$

Some simple substitutions

- 1 Substitute (2) into (3)
- 2 Substitute (3) into (2)
- 3 Require that $\sqrt{x^2 + y^2 + z^2} = 1$

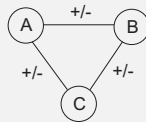
Results

$$x = 0.52 \quad y = 0.48 \quad z = 0.71$$

Structural balance

Basic idea

Consider **triads**: potential relationships between triples of social entities, and label every relationship as positive or negative. We then consider **balanced** triads.



A-B	B-C	A-C	B/I	Description
+	+	+	B	Everyone likes each other
+	+	-	I	Dislike A-C stresses relation B has with either of them
+	-	+	I	Dislike B-C stresses relation A has with either of them
+	-	-	B	A and B like each other, and both dislike C
-	+	+	I	Dislike A-B stresses relation C has with either of them
-	+	-	B	B and C like each other, and both dislike A
-	-	+	B	A and C like each other, and both dislike B
-	-	-	I	Nobody likes each other

Structural balance: signed graphs

Definition

A **signed graph** is a simple graph G in which each edge e is labeled with either a positive (“+”) or negative (“-”) sign, $sign(e)$.

Definition

The **product of two signs** s_1 and s_2 is again a sign, denoted as $s_1 \cdot s_2$. It is negative if and only if *exactly one* of s_1 and s_2 is negative. The **sign of a trail** T is the product of the signs of its edges:
 $sign(T) = \prod_{e \in E(T)} sign(e)$.

Definition

An undirected signed graph is **balanced** when all its cycles are positive.

Balanced networks: special characterization

Theorem

An undirected signed *complete* graph G is balanced if and only if $V(G)$ can be partitioned into two disjoint subsets V_0 and V_1 such that each negative-signed edge is incident to a vertex from V_0 and one from V_1 , and each positive-signed edge is incident to vertices from the same set.

More formally

Let $E^-(G)$ be the edges with negative sign, and $E^+(G)$ the ones with positive sign. Then, $E^-(G) = \{\langle x, y \rangle \mid x \in V_0, y \in V_1\}$ and $E^+(G) = \{\langle x, y \rangle \mid x, y \in V_0 \text{ or } x, y \in V_1\}$.

Proof: V can be properly partitioned $\Rightarrow G$ is balanced

Every cycle in G contains an even number of edges from $E^-(G)$. All other edges have positive sign. G must be balanced.

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Proof

Proof: G is balanced $\Rightarrow V$ can be partitioned

- Let $u \in V(G)$ and let $N^+(u) = \{v \in N(u) \mid \text{sign}(\langle u, v \rangle) = "+" \}$
- Set $V_0 \leftarrow \{u\} \cup N^+(u)$ and $V_1 \leftarrow V(G) \setminus V_0$.
- Consider $v_0, w_0 \in V_0$, other than u . Note: $\langle u, v_0 \rangle$ and $\langle u, w_0 \rangle$ are positive signed \Rightarrow also $\langle v_0, w_0 \rangle$ is positive signed.
- Consider $v_1, w_1 \in V_1$. The triangle with vertices u, v_1, w_1 must be positive; $\langle u, v_1 \rangle$ and $\langle u, w_1 \rangle$ are negative signed $\Rightarrow \langle v_1, w_1 \rangle$ must be positive signed.
- Consider $\langle v_0, v_1 \rangle$, $\text{sign}(\langle u, v_0 \rangle)$ is positive, $\text{sign}(\langle u, v_1 \rangle)$ negative $\Rightarrow \langle v_0, v_1 \rangle$ must be negative signed.

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Balanced networks: path characterization

Theorem

Consider an undirected signed graph G and two distinct vertices $u, v \in V(G)$. G is balanced if and only if all (u, v) -paths have the same sign.

Proof: G is balanced \Rightarrow all (u, v) -paths have the same sign

- Let P and Q be two distinct (u, v) -paths.
- Let $E' = (E(P) \cup E(Q)) \setminus (E(P) \cap E(Q))$.
- $G[E']$ consists of edge-disjoint positive-signed cycles.
- For each cycle C : $E(C) = E(\hat{P}) \cup E(\hat{Q})$ with \hat{P} (\hat{Q}) a subpath of P (Q).
- $\text{sign}(C) = \text{sign}(\hat{P}) \cdot \text{sign}(\hat{Q})$ is positive \Rightarrow signs of \hat{P} and \hat{Q} must be the same.

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Balanced networks: path characterization

Proof: all (u, v) -paths have the same sign $\Rightarrow G$ is balanced**Note:**

- u and v have been chosen arbitrarily
- Every cycle C can be constructed as the union of two edge-disjoint paths P and Q

Consequence: for all C : $\text{sign}(C) = \text{sign}(P) \cdot \text{sign}(Q)$ must be positive $\Rightarrow G$ is balanced.

Balanced networks: general characterization

Theorem

An undirected signed graph G is balanced if and only if $V(G)$ can be partitioned into two disjoint subsets V_0 and V_1 such that

$$E^-(G) = \{\langle x, y \rangle \mid x \in V_0, y \in V_1\} \text{ and}$$

$$E^+(G) = \{\langle x, y \rangle \mid x, y \in V_0 \text{ or } x, y \in V_1\}.$$

Proof: V can be properly partitioned $\Rightarrow G$ is balanced

- Add $e = \langle u, v \rangle$ to G , with u, v nonadjacent
- u and v in same subset $\Rightarrow \text{sign}(e)$ becomes positive, otherwise negative.
- Continue until reaching complete signed graph G^* .
- We know G^* is balanced $\Rightarrow G$ is balanced.

Balanced networks: general characterization

Proof: G is balanced $\Rightarrow V$ can be properly partitioned

- Assume G is connected. Prove by induction on number of edges m .
- Trivially OK for $m = 1$. Assume correct for $m > 1$ edges.
- Consider nonadjacent vertices u and v : all (u, v) -paths have the same sign. Add $e = \langle u, v \rangle$.
- New cycle C will consist of e and a (u, v) -path P from G .
- $\text{sign}(C) = \text{sign}(e) \cdot \text{sign}(P)$, and $\text{sign}(C) = \text{sign}(P) \Rightarrow \text{sign}(e) = \text{sign}(P) \Rightarrow C$ must be positive.
- Continue until reaching complete graph G^* , and subsequently partition $V(G^*)$.

Checking for balance

Algorithm (Balanced graphs)

Consider an undirected signed graph G . $N^+(v)$ is the set of vertices adjacent to v through a positive-signed edge. $N^-(v)$ is analogous. Let I be the set of inspected vertices so far.

- 1 Select an arbitrary vertex $u \in V(G)$ and set $V_0 \leftarrow \{u\}$ and $V_1 \leftarrow \emptyset$. Set $I \leftarrow \emptyset$.
- 2 Select arbitrary vertex $v \in (V_0 \cup V_1) \setminus I$. Assume $v \in V_i$.
 - For all $w \in N^+(v)$: $V_i \leftarrow V_i \cup \{w\}$.
 - For all $w \in N^-(v)$: $V_{(i+1) \bmod 2} \leftarrow V_{(i+1) \bmod 2} \cup \{w\}$.
 - Also, $I \leftarrow I \cup \{v\}$.
- 3 If $V_0 \cap V_1 \neq \emptyset$ stop: G is not balanced. Otherwise, if $I = V(G)$ stop: G is balanced. Otherwise, repeat the previous step.

Affiliation networks

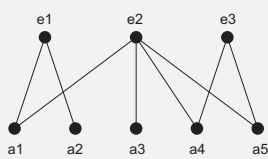
Basic idea

Social structures are assumed to consist of **actors** and **events**. Actors are tied to each other through their participation in an event. Two events are bound through the actors that participate in both events \Rightarrow **two-mode networks**.

Observation

Affiliation networks are naturally represented as **bipartite graphs**: Let V_A represent the actors and V_E the events. Edge (v_a, v_e) if actor a participates in event e .

Affiliation networks & adjacency submatrix



	e1	e2	e3
a1	1	1	0
a2	1	0	0
a3	0	1	0
a4	0	1	1
a5	0	1	1

Special tables

Note

$AE[i, j] = 1$ if and only if actor i participated in event j

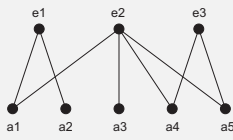
Part 1

$$NE[i, j] = \sum_{k=1}^{n_E} AE[i, k] \cdot AE[j, k]$$

Part 2

$$NA[i, j] = \sum_{k=1}^{n_A} AE[k, i] \cdot AE[k, j]$$

Counting joint participations



NE	a1	a2	a3	a4	a5	NA	e1	e2	e3
a1	2	1	1	1	1	e1	2	1	0
a2	1	1	0	0	0	e2	1	4	2
a3	1	0	1	1	1	e3	0	2	2
a4	1	0	1	2	2				
a5	1	0	1	2	2				

THE END