# Topics of Complex Social Networks: Domination, Influence and Assortativity

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# Python Code at github:

 $\verb|https://github.com/mboudour/GraphMultilayerity/tree/master/vartopics|$ 

## Slides at slideshare:

http://www.slideshare.net/MosesBoudourides/topics-of-complex-socialnetworks-domination-influence-and-assortativity

# Table of Contents

- A. Dominating Sets
- ♣ B. Network Influence and Scalar Attribute Assortativity
- C. Multilayer Networks and Enumerative Attribute Assortativity

# A. Dominating Sets

# Dominating Sets in a Graph

#### Definition

Let G = (V, E) be a (simple undirected) graph. A set  $S \subseteq V$  of vertices is called a **dominating set** (or **externally stable**) if every vertex  $v \in V$  is either an element of S or is adjacent to an element of S.

#### Remark

A set  $S \subseteq V$  is a dominating set if and only if:

- for every  $v \in V \setminus S$ ,  $|N(v) \cap S| \ge 1$ , i.e.,  $V \setminus S$  is enclaveless;
- $\bullet \ \mathsf{N}[S] = V;$
- for every  $v \in V \setminus S$ ,  $d(v, S) \leq 1$ .

Above N(v) is the open neighborhood of vertex v, i.e.,  $N(v) = \{w \in V : (u, w) \in E\}$ , N[v] is the closed neighborhood of vertex v, i.e.,  $N[v] = N(v) \cup \{v\}$ , d(v, x) is the geodesic distance between vertices v and x and  $d(v, S) = \min\{d(v, s) : s \in S\}$  is the distance between vertex v and the set of vertices S.

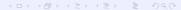
#### Standard Reference

Haynes, Teresa W., Hedetniemi, Stephen T., & Slater, Peter J. (1998). *Fundamentals of Domination in Graphs*. New York: Marcel Dekker.

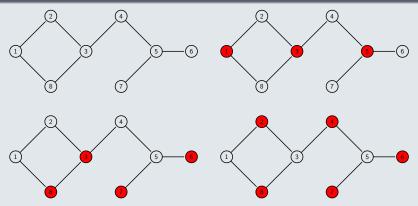
#### Definition

Let S be a set of vertices in graph G.

- *S* is called **independent** (or **internally stable**) if no two vertices in *S* are adjacent.
- If S is a dominating set, then
  - S is called **minimal dominating set** if no proper subset  $S' \subset S$  is a dominating set;
  - S is called a minimum dominating set if the cardinality of S is minimum among the cardinalities of any other dominating set;
  - The cardinality of a minimum dominating set is called the domination number of graph G and is denoted by  $\gamma(G)$ ;
  - *S* is called **independent dominating set** if *S* is both an independent and a dominating set.



## Example



The sets  $\{1,3,5\}$ ,  $\{3,6,7,8\}$  and  $\{2,4,6,7,8\}$  (red circles) are all minimal dominating sets. However, only the first is a minimum dominating set. Apparently  $\gamma=3$  for this graph.

# Theorem (Ore, 1962)

A dominating set S is a minimal dominating set if and only if, for each vertex  $u \in S$ , one of the following two conditions holds:

- ① u is such that  $N(u) \subseteq V \setminus S$ , i.e., u is an isolate of S,
- ② there exists a vertex  $v \in V \setminus S$  such that  $N(v) \cap S = \{u\}$ , i.e., v is a *private neighbor* of u.

#### Theorem (Ore, 1962)

Every connected graph G of order  $n \ge 2$  has a dominating set S.

# Complexity of the Dominating Set Problem

## Theorem (Johnson, 1974)

The dominating set problem is  $\mathcal{NP}$ -complete.

# Algorithmic Computation of Dominating Sets

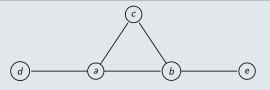
- As the dominating set problem is NP-complete, there are certain efficient algorithms (typically based on linear and integer programming) for its approximate solution.
- Here, we are algorithms that have been already implemented in Python's Sage. However, these implementations return only one of the minimum or independing (minimum) dominating sets.

# An Algebraic Computation of All Minimal Dominating Sets

## A Brute Force Heuristics (Berge)

- Given two elements x and y of a set, let
  - "x + y" denote *logical summation* of x and y (i.e., "x or y")
  - and " $x \cdot y$ " denote their logical multiplication (i.e., "x and y").
- Thus, for each vertex v,  $N[v] = v + u_1 + ... + u_k$ , where  $u_1, ..., u_k$  are all vertices adjacent to v.
- In this way,  $Poly(G) = \prod_{v \in V} N[v]$  becomes a polynomial in graph vertices.
- Notice that each monomial of Poly(G) can be simplified by removing numerical coefficients and repeated (more than once) vertices in it.
- Therefore, each simplified monomial of Poly(G) including r vertices is a minimal dominating set of cardinality r and a minimum dominating set corresponds to the minimal r.

#### Example



As N[a] = a + b + c + d, N[b] = a + b + c + e, N[c] = a + b + c, N[d] = a + d, N[e] = b + e, we find:

Poly(G) = 
$$N[a] N[b] N[c] N[d] N[e] =$$
  
=  $(a+b+c+d)(a+b+c+e)(a+b+c)(a+d)(b+e) =$   
=  $(a+b+c)^3(a+d)(b+e) +$   
 $+(a+b+c)^2(d+e)(a+d)(b+e) =$   
=  $ab+ae+bd+$   
 $+abc+abd+abe+ace+ade+bcd+bde+cde+$   
 $+abcd+abde+abce+acde+bcde+$ 

#### Therefore,

- $\{\{a,b\},\{a,e\},\{b,d\}\}$  is the collection of minimum dominating sets (each one of cardinality 2 equal to the domination number of this graph),
- $\{\{a,b,c\},\{a,b,d\},\{a,b,e\},\{a,c,e\},\{a,d,e\},\{b,c,d\},$  $\{b,d,e\},\{c,d,e\}\}$  is the collection of minimal dominating sets of cardinality 3,
- $\{\{a,b,c,d\},\{a,b,d,e\},\{a,b,c,e\},\{a,c,d,e\},\{a,d,e\},$  $\{b,c,d,e\}\}$  is the collection of minimal dominating sets of cardinality 4 and
- $\{a, b, c, d, e\}$  is the minimal dominating set of cardinality 5.

# Egocentric Subgraphs Induced by a Dominating Set in a Graph

#### Definition

Let G = (V, E) be a (simple undirected) graph.

- Let  $U \subset V$  be a set of vertices in a graph G and let  $u \in U$ .
  - The subgraph induced by U (in G) is called an egocentric (sub)graph (in G) if

$$U = N[u],$$

i.e., if all vertices of U are dominated by u;

- vertex u is the **ego** of the egocentric (sub)graph N[u];
- vertices  $w \in N(u)$  are called **alters** of ego u.
- Let  $S \subset V$  be a dominating set in G and let  $v \in S$ .
  - The subgraph induced by N[v] (in G) is called a **dominating** egocentric (sub)graph;
  - vertex v is the dominating ego;
  - vertices  $w \in N(v)$  are the **dominated alters** by v.
  - Graph G is called multiple egocentric or |S|-egocentric graph corresponding to the dominating set S.

#### Remark

- Given a (simple undirected) graph, typically, there is a multiplicity of dominating sets for that graph.
- Therefore, any egocentric decomposition of a graph depends on the dominating set that was considered in the generation of the constituent egocentric subgraphs.

# Private and Public Alters

#### Definition

Let S be a dominating set in graph G = (V, E) ( $|S| \ge 2$ ) and let  $v \in S$  an ego.

- An alter  $w \in N(v)$  ( $w \notin S$ ) is called **private alter** if  $N(w) \subset N[v]$ .
- An alter  $w \in N(v)$  ( $w \notin S$ ) is called **public alter** if  $N(w) \setminus N[v] \neq \emptyset$ .

#### Remark

- Two adjacent egos u, v are not considered alters to each other! Nevertheless, they define an ego—to—ego edge (bridge).
- A private alter is always adjacent to a single ego.
- A public alter is adjacent either to at least two egos or to a single ego and to another alter which is adjacent to a different ego.

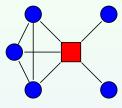


Figure : Five private alters (blue circles) adjacent to an ego (red square).



Figure : *Left*: one public alter (green circle) shared by two egos (red squares). *Right*: two public alters (green circles), each one adjacent to a different ego (red square).

# Dominating and Dominated Bridges

#### Definition

Let S be a dominating set in a graph G = (V, E)  $(|S| \ge 2)$ .

- If two egos  $v_1, v_2 \in S$  are adjacent, then edge  $(v_1, v_2) \in E$  is called **dominating edge** or **bridge of egos**.
- If two private alters  $w_1 \in N(v_1)$ ,  $w_2 \in N(v_2)$  are adjacent (possibly  $v_1 = v_2$ ), then edge  $(w_1, w_2) \in E$  is called **dominated edge among private alters** or **bridge of private alters**.
- If two public alters  $w_1 \in N(v_1), w_2 \in N(v_2)$  are adjacent (possibly  $v_1 = v_2$ ), then edge  $(w_1, w_2) \in E$  is called **dominated edge among public alters** or **bridge of public alters**.
- If a private alter  $w_1 \in N(v_1)$  and a public alter  $w_2 \in N(v_2)$  are adjacent (possibly  $v_1 = v_2$ ), then edge  $(w_1, w_2)$  is called dominated edge among private—public alters or bridge of private—to—public alters.

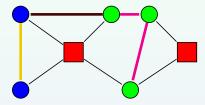


Figure: One edge (bridge) among private alters (yellow line), two edges (bridges) among public alters (magenta lines) and one edge (bridge) among private—to—public alters (brown lines).

#### Notation

• The set of all private alters in G is denoted as:

$$\textcolor{red}{V_{\mathsf{private}}} = \{ w \in V \smallsetminus S \colon \exists v \in S \text{ s.t. } w \in N(v) \smallsetminus S \text{ and } N(w) \subset N[v] \}.$$

The set of all public alters in G is denoted as:

$$V_{\text{public}} = \{ w \in V \setminus S \colon N(w) \setminus N[v] \neq \emptyset, \forall v \in S \}.$$

The set of all dominating bridges (among egos) in G is denoted as:

$$E_{\text{ego}} = \{(v_1, v_2) \in E: v_1, v_2 \in S\}.$$

• The set of all dominated bridges among private alters in G is denoted as:

$$E_{private} = \{(v_1, v_2) \in E: v_1, v_2 \in V_{private}\}.$$

• The set of all dominated bridges among public alters in G is denoted as:

$$E_{\text{public}} = \{(v_1, v_2) \in E: v_1, v_2 \in V_{\text{public}}\}.$$

• The set of all dominated bridges among private—public alters in G is denoted as:

$$E_{\text{private}} = \text{public} = \{(v_1, v_2) \in E: v_1 \in V_{\text{private}}, v_2 \in V_{\text{public}}\}.$$

#### Definition

Let  $G \setminus S$  be the graph remaining from G = (V, E) after removing a dominating set S (together with all edges which are incident to egos in S). Sometimes,  $G \setminus S$  is called **alter (sub)graph** (or **alter-to-alter (sub)graph**) (of G). In other words,  $G \setminus S$  is the subgraph induced by  $V \setminus S$  and the set of edges of  $G \setminus S$  is the set

$$E_{\mathsf{private}} \cup E_{\mathsf{public}} \cup E_{\mathsf{private} - \mathsf{public}}.$$

#### Proposition

If S is a dominating set of a graph G = (V, E), then

$$|S| + |V_{private}| + |V_{public}| = |V|,$$

$$\sum_{v \in \mathcal{E}} \mathsf{degree}(v) + |E_{\mathsf{private}}| + |E_{\mathsf{public}}| + |E_{\mathsf{private}}| + |E_{\mathsf{private}}| = |E|.$$

#### The Network of the 15 Florentine Families: All Dominating Egocentric

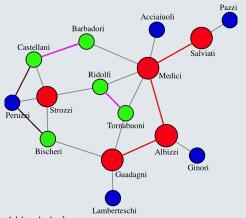
#### Subgraphs

Minimum Dominating Set (S)	IDS <sup>1</sup>	$ V_{private} $	$ V_{public} $	$\sum_{v \in S} degree(v)$	$ E_{private} $	$ E_{\text{public}} $	$ E_{private-public} $	$ E_{\rm ego} $
$\{A,C,K,J,M\}$		3	7	14	0	5	2	1
$\{C, E, F, I, J\}$	~	2	8	15	0	5	0	0
$\{C, F, I, J, M\}$	~	1	9	12	0	8	0	0
$\{C,I,K,J,M\}$		2	8	13	0	8	0	1
$\{C, D, F, I, M\}$		2	8	14	0	7	0	1
$\{C, E, D, F, L\}$		3	7	17	0	5	0	2
$\{A,C,E,K,J\}$		4	6	17	0	2	2	1
${A, C, F, J, M}$	_	2	8	13	0	5	2	0
$\{C, E, I, K, J\}$		3	7	16	0	5	0	1
$\{A,C,D,F,M\}$		3	7	15	0	4	2	1
$\{A,C,E,D,K\}$		5	5	19	0	2	2	3
$\{A, C, D, K, M\}$		4	6	16	0	4	2	2
$\{C, E, K, J, L\}$		3	7	16	0	5	0	1
$\{C, E, F, J, L\}$	~	2	8	15	0	5	0	0
$\{C, E, D, F, I\}$		3	7	17	0	5	0	2
$\{A,C,E,D,F\}$		4	6	18	0	2	2	2
$\{C, D, I, K, M\}$		3	7	15	0	7	0	2
$\{C, E, D, I, K\}$		4	6	18	0	5	0	3
$\{A,C,E,F,J\}$	~	3	7	16	0	2	2	0
$\{C, E, D, K, L\}$		4	6	18	0	5	0	3

The 15 Florentine Families: A = Strozzi, B = Tornabuoni, C = Medici, D = Albizzi, E = Guadagni, F = Pazzi, G = Acciaiuoli, H = Bischeri, I = Peruzzi, J = Ginori, K = Salviati, L = Castellani, M = Lamberteschi, N = Ridolfi, O = Barbadori.

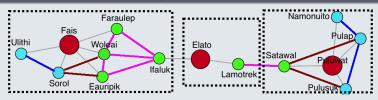
<sup>&</sup>lt;sup>1</sup>Independent Dominating Set (IDS).

# Example (Florentine Families Network)



```
\begin{split} |S| &= 5 \text{ (5 egos, red big circles);} \\ |V_{\text{private}}| &= 5 \text{ (5 private alters, blue circles);} \\ |V_{\text{public}}| &= 5 \text{ (5 public alters, green circles);} \\ |E_{\text{private}}| &= 0 \text{ (no edges among private alters);} \\ |E_{\text{public}}| &= 2 \text{ (2 edges among public alters, magenta lines);} \\ |E_{\text{private}}| &= p_{\text{ublic}}| &= 2 \text{ (2 edges among private-to-public alters, brown lines).} \\ |E_{\text{ego}}| &= 3 \text{ (3 among egos, red lines).} \end{split}
```

# Example (The Voyaging Network among 14 Western Carolines Islands, Hage & Harary, 1991)



|S| = 3 (3 egos, red big circles);

 $|V_{\text{private}}| = 5$  (5 private alters, blue circles);

 $|V_{\text{public}}| = 6$  (6 public alters, green circles);

 $E_{private} = 3$  (3 edges among private alters, blue lines);

 $E_{\text{public}} = 6$  (6 edges among public alters, magenta lines);

 $E_{private - public} = 4$  (4 edges among private-to-public alters, brown lines).

(Three communities enclosed in dotted rectangles.)

Minimum Dominating Set $(S)$	IDS	$ V_{private} $	$ V_{\text{public}} $	$\sum_{v \in S} degree(v)$	$ E_{private} $	$ E_{\text{public}} $	$ E_{private-public} $	$ E_{\rm ego} $
$\{G,I,M\}$	~	5	6	11	3	6	4	0
$\{C,I,M\}$	~	5	6	11	3	6	4	0

The Dominant Western Carolines Islands: C = Puluwat, G = Pulap, I = Fais, M = Elato.

# Community Partitions in a Graph

#### Definition (Newman & Girvan, 2004)

Let G = (V, E) be a graph.

- A **clustering** of vertices of G is a partition of the set of vertices V into a (finite) family  $C \subset 2^V$  of subsets of vertices, often called **modules**, such that  $\bigcup_{C \in C} C = V$  and  $C \cap C' = \emptyset$ , for each  $C, C' \in C, C \neq C'$ .
- A clustering C may be assessed by a quality function Q = Q(C), called **modularity**, which is defined as:

$$Q = \sum_{C \in \mathcal{C}} \left[ \frac{|E(C)|}{|E|} - \left( \frac{2|E(C)| + \sum_{C' \in \mathcal{C}, C \neq C'} |E(C, C')|}{2|E|} \right)^2 \right]$$
$$= \sum_{C \in \mathcal{C}} \left[ \frac{|E(C)|}{|E|} - \left( \frac{\sum_{v \in C} degree(v)}{2|E|} \right)^2 \right],$$

where E(C) is the number of edges inside module C and E(C,C') is the number of edges among modules C and C'. Essentially, modularity compares the number of edges inside a given module with the expected value for a randomized graph of the same size and same degree sequence.

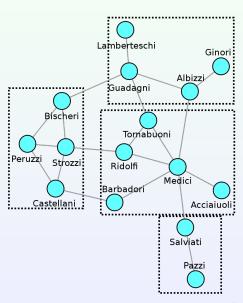
 A clustering that maximizes modularity is usually called community partition and the corresponding modules are called communities. Theorem (Brandes, Delling, Gaertler, Görke, Hoefer, Nikoloski & Wagner, 2008)

Modularity is strongly  $\mathcal{NP}$ -complete.

# Algorithmic Computation of Dominating Sets

Here, we are using the community detection algorithm (through modularity maximization) of the **Louvain method** (Blondel, Guillaume, Lambiotte & Lefebvre, 2008) as implemented in Python by **Thomas Aynaud**.

# Communities in the Network of Florentine Families

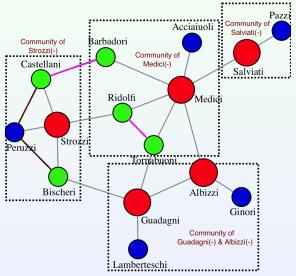


# The Florentine Families Network: All Dominating Egocentric Subgraphs and their Community Partitions

Minimum Dominating Set $(S)$	γ	C	Community Partition $(C)$
${A, C, K, J, M}$	5	4	${A(-), C(-), J + M, K(-)}$
$\{C, E, F, I, J\}$	5	4	$\{I, C(-), E(-) + J, F\}$
$\{C, F, I, J, M\}$	5	4	$\{I,C(-),J+M,F\}$
$\{C,I,K,J,M\}$	5	4	$\{I, C(-), J + M, K(-)\}$
$\{C,D,F,I,M\}$	5	4	$\{I, C(-), D(-) + M, F\}$
$\{C, E, D, F, L\}$	5	4	$\{L(-), C(-), E(-) + D(-), F\}$
$\{A,C,E,K,J\}$	5	4	$\{A(-), C(-), E(-) + J, K(-)\}$
$\{A,C,F,J,M\}$	5	4	$\{A(-), C(-), J+M, F\}$
$\{C, E, I, K, J\}$	5	4	$\{I, C(-), E(-) + J, K(-)\}$
$\{A,C,D,F,M\}$	5	4	$\{A(-), C(-), D(-) + M, F\}$
$\{A,C,E,D,K\}$	5	4	$\{A(-), C(-), E(-) + D(-), K(-)\}$
$\{A,C,D,K,M\}$	5	4	$\{A(-), C(-), D(-) + M, K(-)\}$
$\{C, E, K, J, L\}$	5	4	$\{L(-), C(-), E(-) + J, K(-)\}$
$\{C, E, F, J, L\}$	5	4	$\{L(-), C(-), E(-) + J, F\}$
$\{C, E, D, F, I\}$	5	4	$\{I, C(-), E(-) + D(-), F\}$
$\{A,C,E,D,F\}$	5	4	$\{A(-), C(-), E(-) + D(-), F\}$
$\{C, D, I, K, M\}$	5	4	$\{I, C(-), D(-) + M, K(-)\}$
$\{C, E, D, I, K\}$	5	4	$\{I, C(-), E(-) + D(-), K(-)\}$
$\{A,C,E,F,J\}$	5	4	$\{A(-), C(-), E(-) + J, F\}$
$\{C, E, D, K, L\}$	5	4	$\{L(-), C(-), E(-) + D(-), K(-)\}$

The 15 Florentine Families: A = Strozzi, B = Tornabuoni, C = Medici, D = Albizzi, E = Guadagni, F = Pazzi, G = Acciaiuoli, H = Bischeri, I = Peruzzi, J = Ginori, K = Salviati, L = Castellani, M = Lamberteschi, N = Ridolfi, O = Barbadori.

# The Network of Florentine Families: Egos in Communities



Five egos distributed inside four communities.

Egos are colored red, private alters blue and public alters green.

Dotted rectangles embrace vertices lying inside the same communities.

# B. Network Influence and Scalar Attribute Assortativity

# The Friedkin-Johnsen Model of Social Influence

- Let G = (V, E) be a (simple undirected) graph.
- Vertices represent persons.
- For  $i \in V$  and at each time t, person i holds an opinion (or attitude)  $x_i^t$ , where  $x_i^t$  is a considered to be a **scalar attribute** of graph vertices that takes values in the interval [0,1].
- ullet Each person interacts with all her graph neighbors and updates her opinion at the subsequent time t+1 as follows:

$$x_i^{t+1} = \sigma_i \sum_{j \in N(i) \neq \varnothing} \frac{1}{k_i} A_{ij} x_i^t + (1 - \sigma_i) x_i^t,$$

where  $\mathbf{A}=A_{ij}$  is the adjacency matrix of the graph,  $k_i$  is the degree of i, N(i) is the set of neighbors of i and  $\sigma_i \in [0,1]$  is the **susceptilibity** of i to the influence of her neighbors. Let  $\mathbf{S}$  be the diagonal matrix such that  $S_{ii}=\frac{\sigma_i}{k_i}$ , for all  $i\in V$ . Note that when  $k_i=0$ , then  $A_{ij}=0$ , for all  $j\in V$  and, thus,  $\mathbf{S}$  is well defined, if G is assumed to be free of isolated vertices.

#### Proposition

If G is a connected graph, then, for any person  $i \in V$  and any initial opinion  $x_i^0 \in [0,1]$ , there exists a **consensus** opinion  $x^\infty \in [0,1]$  such that

$$\lim_{t\to\infty} x_i^t = x^{\infty}.$$

In fact,  $x^{\infty}$  is given by

$$\mathbf{x}^{\infty} = (\mathbf{I} - \mathbf{S}\mathbf{A})^{-1}(\mathbf{I} - \mathbf{S})\mathbf{x}^{0},$$

where I is the unit matrix (1's on the diagonal and 0's elsewhere).

# Contrary Influence

- The previous mechanism of social influence converges to a consesus, because at each time step each person's opinion is compromized with the opinions of her neighbors.
- However, one can also think of a mechanism of negative influence, when at each time step each person's opinion tends to diverge from the opinions of her neighbors in the following way:

$$\begin{aligned} x_i^{t+1} &= \sigma_i D(x_i^t) + (1-\sigma_i) x_i^t, \\ \text{where, denoting } y_i^t &= \sum_{j \in N(i) \neq \varnothing} \frac{1}{k_i} A_{ij} x_i^t, \\ D(x_i^t) &= \left\{ \begin{array}{ll} \max\{2x_i^t - y_i^t, 0\}, & \text{when } x_i^t \leq y_i^t \\ \min\{2x_i^t - y_i^t, 1\}, & \text{when } x_i^t > y_i^t. \end{array} \right. \end{aligned}$$

## Proposition

For the model of negative influence, if G is a connected graph, then, for any person  $i \in V$  and any initial opinion  $x_i^0 \in [0, 1]$ ,

$$\lim_{t\to\infty} x_i^t = 0 \text{ or } 1.$$

# Scalar Attribute Assortativity Coefficient

- Let G = (V, E) be a graph with adjacency matrix  $A_{ij}$ ,  $k_i$  be the degree of  $i \in V$  and |E| = m.
- Let each vertex *i* possess a scalar attribute  $x_i$  ( $\in \mathbb{R}$ ).
- Then one can define (cf. Mark Newman, 2003), the (normalized) scalar attribute assortativity as follows:

$$r = \frac{\sum_{i,j \in V} (A_{ij} - \frac{k_i k_j}{2m}) x_i x_j}{\sum_{i,j \in V} (k_i \delta_{ij} - \frac{k_i k_j}{2m}) x_i x_j},$$

where  $\delta_{ij}$  is the Kronecker delta. This is an example of a (Pearson) correlation coefficient with a covariance in its numerator and a variance in the denominator.

• Clearly,  $r \in [-1,1]$  and, as we will discuss later in the case of enumerative attributes, r is a measure of assortative (or disassortative) mixing according to the scalar attribute  $x_i$ , with vertices having similar values of this attribute being more (or less) likely to be connected by an edge.

# C. Multilayer Networks and Enumerative Attribute Assortativity

# **Graph Partitions**

Let

$$G=(V,E)$$

be a (simple undirected) graph with (finite) set of vectors V and (finite) set of edges E.

- The terms "graph" and "(social) network" are used here interchangeably.
- If  $S_1, S_2, \ldots, S_s \subset V$  (for a positive integer s) are s nonempty (sub)sets of vertices of G, the (finite) family of subsets

$$\mathcal{S} = \mathcal{S}(G) = \{S_1, S_2, \dots, S_s\},\$$

forms a **partition** (or s-**partition**) in G, when, for all  $i, j = 1, 2, \dots, s, i \neq j$ ,

$$S_i \cap S_j = \emptyset$$
 and

$$V = \bigcup_{i=1}^{s} S_i.$$

•  $S_1, S_2, \ldots, S_s$  are called **groups** (of vertices) of partition S.

- For every graph, there exists (at least one) partition of its vertices.
- Two trivial graph partitions are:
  - ullet the |V|-partition into singletons of vertices and

$$S_{point} = S_{point}(G) = \{\{v\} : v \in V\},$$

the 1-partition into the whole vertex set

$$S_{\text{total}} = S_{\text{total}}(G) = \{V\}.$$

 If G is a bipartite graph with (vertex) parts U and V, the bipartition of G is denoted as:

$$S_{\text{bipartition}} = S_{\text{bipartition}}(G) = \{U, V\}.$$

### **Examples of Nontrivial Graph Partitions**

### Structural (endogenous) partitions:

- Connected components
- Communities
- Chromatic partitions (bipartitions, multipartitions)
- Degree partitions
- Time slicing of temporal networks
- Domination bipartitions (ego- and alter-nets)
- Core–Periphery bipartitions

### Ad hoc (exogenous) partitions:

- Vertex attributes (or labels or colors)
- Layers (or levels)



### Multilayer Partitions and Multilayer Graphs

• A multilayer partition of a graph G is a partition

$$\mathcal{L} = \mathcal{L}(G) = \{L_1, L_2, \dots, L_\ell\}$$

into  $\ell \geq 2$  groups of vertices  $L_1, L_2, \dots, L_{\ell}$ , which are called **layers**.

 A graph with a multilayer partition is called a multilayer graph.

### Ordering Partitions

- Let  $\mathcal{P} = \{P_1, P_2, \dots, P_p\}$  and  $\mathcal{Q} = \{Q_1, Q_2, \dots, Q_q\}$  be two (different) partitions of vertices of graph G = (V, E).
- Partition  $\mathcal{P}$  is called **thicker** than partition  $\mathcal{Q}$  (and  $\mathcal{Q}$  is called **thinner** than  $\mathcal{P}$ ), whenever, for every  $j=1,2,\cdots,q$ , there exists a  $i=1,2,\cdots,p$  such that

$$Q_j \subset P_i$$
.

• Partitions  $\mathcal{P}$  and  $\mathcal{Q}$  are called (enumeratively) equivalent, whenever, for all  $i=1,2,\cdots,p$  and  $j=1,2,\cdots,q$ ,

$$P_i \cap Q_j \neq \emptyset$$
.

ullet Obviously, if partitions  ${\mathcal P}$  and  ${\mathcal Q}$  are equivalent, then

$$pq \leq |V|$$
.



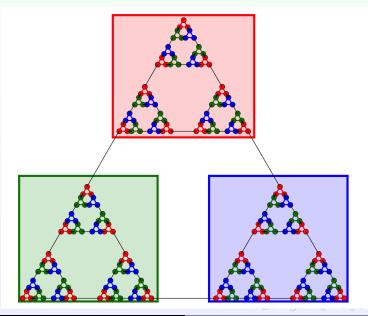
# Self–Similarity of Partitions

• A partition  $\mathcal{P}$  of vertices of graph G (such that  $|V| \geq p^2$ ) is called (**enumeratively**) **self–similar**, whenever, for any  $i = 1, 2, \dots, p$ , there is a p–(sub)partition

$$\mathcal{P}^i = \{P_1^i, P_2^i, \dots, P_p^i\}$$

- of the (induced) subgraph  $P_i$ .
- In this case, graph G is called (enumeratively) self–similar with respect to partition  $\mathcal{P}$ .

### A Sierpinski (Cantor-type) Graph with k = 2.5 and depth 4



# Graph Partitions as Enumerative Attribute Assignments

 An assignment of enumerative attributes (or discrete attribute assignment) to vertices of graph G is a mapping

$$A: V \to \{1, 2, \cdots, \alpha\}, \text{ for some } \alpha \leq |V|,$$

through which the vertices of G are classified according to the values they take in the set  $\{1, 2, \dots, \alpha\}$ .

- Every p-partition  $\mathcal{P} = \{P_1, P_2, \dots, P_p\}$  of vertices of G corresponds to a p-assignment  $\mathcal{A}_{\mathcal{P}}$  of enumerative attributes to vertices of G distributing them in the groups of partition  $\mathcal{P}$ .
- Conversely, every assignment of enumerative attributes to vertices of G taking values in the (finite) set  $\{1,2,\cdots,\alpha\}$  corresponds to a partition  $\mathcal{P}^{\alpha}=\{P_{1}^{\alpha},P_{2}^{\alpha},\ldots,P_{p}^{\alpha}\}$  of the vertices of G such that, for any  $k=1,2,\cdots,\alpha$ ,

$$P_k^{\alpha} = \{ v \in V : \mathcal{A}(v) = k \} = \mathcal{A}^{-1}(k).$$

 Thus, vertex partitions and enumerative vertex attribute assignments are coincident.

- Let  $\mathcal{P}$  be a partition of G into p groups of vertices and  $\mathcal{A}$  be an assignment of  $\alpha$  discrete attributes to vertices of G.
- Partition  $\mathcal{P}$  is called **compatible** with attribute assignment  $\mathcal{A}$ , whenever partition  $\mathcal{P}$  is thinner than partition  $\mathcal{P}^{\alpha}$ , i.e., whenever, for every  $k=1,2,\cdots,\alpha$ , there exists (at least one)  $i=1,2,\cdots,p$  such that

$$P_i = \mathcal{A}^{-1}(k).$$

• Apparently, if partition  $\mathcal{P}$  is compatible with attribute assignment  $\mathcal{A}$ , then

$$p \geq \alpha$$
.

 Trivially, every discrete attributes assignment (or every partition) is compatible to itself.

### Assortativity of a Partition

- Let  $\mathcal{P} = \{P_1, P_2, \dots, P_p\}$  be a vertex partition of graph G.
- Identifying  $\mathcal{P}$  to a p-assignment  $\mathcal{A}_{\mathcal{P}}$  of enumerative attributes to the vertices of G, one can define (cf. Mark Newman, 2003), the (normalized) enumerative attribute assortativity (or discrete assortativity) coefficient of partition  $\mathcal{P}$  as follows:

$$r_{\mathcal{P}} = r_{\mathcal{P}}(\mathcal{A}_{\mathcal{P}}) = \frac{\operatorname{tr} \mathbf{M}_{\mathcal{P}} - ||\mathbf{M}_{\mathcal{P}}^2||}{1 - ||\mathbf{M}_{\mathcal{P}}^2||},$$

where  $\mathbf{M}_{\mathcal{P}}$  is the  $p \times p$  (normalized) mixing matrix of partition  $\mathcal{P}$ . Equivalently:

$$r_{\mathcal{P}} = \frac{\sum_{i,j \in V} (A_{ij} - \frac{k_i k_j}{2m}) \delta(\mathcal{A}_{\mathcal{P}}(i), \mathcal{A}_{\mathcal{P}}(j))}{2m - \sum_{i,j \in V} (\frac{k_i k_j}{2m}) \delta(\mathcal{A}_{\mathcal{P}}(i), \mathcal{A}_{\mathcal{P}}(j))},$$

where  $\{A_{ij}\}$  is the adjacency matrix of graph G, m is the total number of edges of G,  $k_i$  is the degree of vertex i and  $\delta(x,y)$  is the Kronecker delta.

• In general,

$$-1 \leq r_{\mathcal{P}} \leq 1$$
,

#### where

- $r_P = 0$  signifies that there is no assortative mixing of graph vertices with respect to their assignment to the p groups of partition  $\mathcal{P}$ , i.e., graph G is configured as a perfectly mixed network (the random null model).
- $r_{\mathcal{P}}=1$  signifies that there is a perfect assortative mixing of graph vertices with respect to their assignment to the p groups of partition  $\mathcal{P}$ , i.e., the connectivity pattern of these groups is perfectly homophilic.
- When  $r_{\mathcal{P}}$  atains a minimum value, which lies in general in the range [-1,0), this signifies that there is a perfect disassortative mixing of graph vertices with respect to their assignment to the p groups of partition  $\mathcal{P}$ , i.e., the connectivity pattern of these groups is perfectly heterophilic.

### Assortative Mixing Among Partitions

- Let  $\mathcal{P} = \{P_1, P_2, \dots, P_p\}$  and  $\mathcal{Q} = \{Q_1, Q_2, \dots, Q_q\}$  be two (different) partitions of vertices of graph G.
- Then, for any  $i=1,2,\cdots,p, j=1,2,\cdots,q, P_i\cap Q_j\neq\varnothing$  and the intersection of  $\mathcal P$  and  $\mathcal Q$

$$\mathcal{P} \cap \mathcal{Q} = \{P_i \cap Q_j \colon i = 1, 2, \cdots, p, j = 1, 2, \cdots, q\}$$

is also a vertex partition of G.

- Notice that partition  $\mathcal{P} \cap \mathcal{Q}$  is compatible with any one of the following three discrete attribute assignments on G:
  - $\mathcal{A}_{\mathcal{P}}$ , in which, for any group  $P_i \cap Q_j$  of  $\mathcal{P} \cap \mathcal{Q}$ , all vertices of  $P_i \cap Q_j$  are assigned a value in the set  $\{1, 2, \dots, p\}$ ,
  - $\mathcal{A}_{\mathcal{Q}}$ , in which, for any group  $P_i \cap Q_j$  of  $\mathcal{Q} \cap \mathcal{Q}$ , all vertices of  $P_i \cap Q_j$  are assigned a value in the set  $\{1, 2, \dots, q\}$  and
  - $\mathcal{A}_{\mathcal{P}\cap\mathcal{Q}}$ , in which, for any group  $P_i\cap Q_j$  of  $\mathcal{P}\cap\mathcal{Q}$ , all vertices of  $P_i\cap Q_j$  are assigned a value in the set  $\{1,2,\cdots,r\}$ , for some  $r\leq pq$ .

- Thus, we may define a discrete assortativity coefficient of partition  $\mathcal{P} \cap \mathcal{Q}$  according to each one of the three compatible attribute assignments.
- Here we will focus on the third case and we define the discrete assortativity coefficient of the joint partition for P and Q as follows:

$$r_{\mathcal{PQ}}=r_{\mathcal{P}\cap\mathcal{Q}}(\mathcal{A}_{\mathcal{P}\cap\mathcal{Q}}).$$

Apparently now,

$$r_{\mathcal{PQ}} = r_{\mathcal{QP}}$$
.

# Special Cases

• If  $\mathcal{Q} = \mathcal{S}_{\mathsf{point}}$ , then  $r_{\mathcal{P}\mathcal{S}_{\mathsf{point}}}$  is the attribute assortativity coefficient  $r_{\mathcal{P}}$  of graph G equipped with the vertex attributes corresponding to partition  $\mathcal{P}$ , i.e.,

$$r_{\mathcal{P}}S_{\text{point}} = r_{\mathcal{P}}.$$

• If  $\mathcal{P} = \mathcal{S}_{\text{total}}$ , then, for any partition  $\mathcal{P}$ ,

$$r_{\mathcal{P}\mathcal{S}_{\mathsf{total}}} = 1.$$

• If G is bipartite and  $\mathcal{P} = \mathcal{S}_{bipartition}$ , then, for any partition  $\mathcal{P}$ ,

$$r_{\mathcal{PS}_{\mathsf{bipartition}}} \in [-1, 0).$$

