# CSI 445/660 – Part 3 (Networks and their Surrounding Contexts)

#### **<u>Ref</u>**: Chapter 4 of [Easley & Kleinberg].

### Homophily:

- A basic principle: "We tend to be similar to our friends".
- Governs the structure of social networks.
- Has a long history:
  - Socrates: "People love those who are like themselves".
  - Plato: "Similarity begets friendship".
  - Well known proverb: "Birds of a feather flock together".
- Provides an illustration of how the surrounding context drives the formation of networks.

# Triadic Closure



- Having a common friend is one reason for triadic closure.
- Homophily provides another reason.
- Suppose B and C are majors in the same department.
- They may become friends even though there is no common friend. (This is an effect of the surrounding context).

#### Measuring Homophily:

- A characteristic must be specified.
- **Examples:** Age, gender, ethnicity.
- How can we check whether a given network exhibits homophily with respect to a specified characteristic?

# Measuring Homophily (continued)



- Friendship network of some children in an elementary school.
- Circles denote girls and squares denote boys.
- We want to check whether this network exhibits gender homophily.
- Extreme case of homophily: The network does not have any "cross-gender edge" (i.e., an edge joining a boy and a girl). This is not typical.
- One can develop a numerical measure of homophily with respect to a characteristic.
- This will be illustrated using a characteristic (namely, gender) which has two possible values.

Description of the Method: See Handout 3.1.

# Measuring Homophily (continued)

**Homophily Test:** Consider a network H with  $N_B$  boys and  $N_G$  girls. Let  $p = N_B/(N_B + N_G)$  and  $q = 1 - p = N_G/(N_B + N_G)$ . If the fraction of cross edges in H is significantly below 2pq, then there is evidence for gender homophily.

#### Example:



- Here,  $N_B = 6$  and  $N_G = 3$ .
- Total number of edges = 18.
- No. of cross edges = 5.
- So, fraction of cross edges = 5/18.

$$p = N_B/(N_B + N_G) = 6/9 = 2/3.$$

• 
$$q = 1 - p = 1/3$$
.

• 
$$2pq = 4/9 = 8/18$$
.

Since the actual fraction of cross edges (5/18) is less than the fraction 2pq, we conclude that the network exhibits some degree of homophily.

# Mechanisms Underlying Homophily

- Homophily is observed behavior.
- Sociologists want to understand the mechanisms that lead to homophily.
- Two known mechanisms are **selection** and **socialization**.

#### Selection:

- Applies to **immutable** characteristics (such as ethnicity or race).
- People "select" friends with similar characteristics.

#### Socialization or Social Influence:

- Applies to mutable characteristics (e.g. behaviors, interests, beliefs, opinions).
- People may modify their characteristics to align with the behaviors of their friends.

### Selection and Socialization

- Socialization may be viewed as the **reverse** of selection.
- Reason:
  - With selection, individual characteristics drive the formation of links.
  - With socialization, links in a network shape people's (mutable) characteristics.
- In general, there is also some interplay between the two mechanisms.

#### Longitudinal Studies to Understand Link Formation:

- From a single snapshot of a network, it is generally difficult to determine the reason for the formation of links.
- Longitudinal studies, where links and behaviors are tracked over a period of time, are needed.

# A Famous Longitudinal Study: Summary

- Published in 2007 by Nicholas Christakis (Yale University) and James Fowler (UC San Diego).
- Longitudinal study (part of Framingham Heart Study) over a 32 year period (1971 to 2003) involving 12,067 people.
- **Focus:** Obesity status.
- **Observation:** Normal weight and overweight people formed clusters in the network consistent with homophily.
- The main cause of homophily was social influence; changes in the obesity status of one's friends had a significant effect on the person.
- The authors go on to suggest that obesity is a form of contagion that spreads through a social network. (This suggestion has been questioned by other researchers.)

### Affiliation Networks

- **So far:** Surrounding context not part of the network.
- The idea of affiliation networks allows the surrounding context to be part of the network.
- Introduce activities or focal points as nodes in the network, leading to a more general form of the network.
- Examples of focal points: Hobbies, interests.

Affiliation Network Example:



- Dark circles: People.
- Squares: Focal points.

# Affiliation Networks (continued)

#### An Affiliation Network from Previous Discussion:



**Note:** An edge between a person x and a company y indicates that x serves (or served) on the Board of Directors for y.

Focus: Formation of edges between people due to focal points.

#### Some Graph Theoretic Definitions:



- An example of a **bipartite** graph.
- There are two sets of nodes.
- Each edge joins a node from one set to a node in the other set. (No edge joins a pair of nodes in the same set.)

**Observation:** Each cycle in a bipartite graph must contain an **even** number of nodes and edges.

### Projected Networks of Affiliation Networks

#### **Projected Network:**

- Network on the nodes representing people.
- There is an edge between two people if they both have edges to at least one common focal point.

**Example 1:** An affiliation network and the corresponding projected network.



**Example 2:** Another affiliation network and the corresponding projected network.



### Link Formation in Affiliation Networks

#### Three forms of Closure Processes:

- Triadic closure: Due to a common friend or homophily.
- Focal closure: A new edge may form because of a common focal point (effect of homophily).



Membership closure: A new edge may form between a person and a focal point (also an effect of homophily).



**Question:** Can we study the link formation issue in a quantitative fashion?

#### Illustration – Study of Triadic Closure:

- Study done by Kossinets and Watts [2006].
- **Caveat:** Study uses online data; conclusions from the study may not be applicable to settings based on human interactions.

#### Basic questions:

- How does the likelihood of the formation of a link increase when two people have one friend in common (compared to when they they have no common friend)?
- 2 How does the likelihood increase when two people have two or more friends in common?

# Study by Kossinets & Watts (continued)



**Note:** We would expect the likelihood to increase as the number of common friends increases.

## Study by Kossinets & Watts (continued)

Description of Methodology: See Handout 3.2.

Example to Illustrate the Methodology:



#### Note:

- In N<sub>1</sub>, pairs of nodes have no common neighbor or one common neighbor.
- So, according to the methodology, we must construct the sets  $S_0$  and  $S_1$ .

# Illustrative Example (continued)



Network  $N_1$ 

Network N2

 $S_0$  = Set of pairs (x, y) such that x and y have **no** common neighbor in  $N_1$  and the edge  $\{x, y\}$  is not in  $N_1$ 

$$= \{(A,E), (B,E), (C,E), (D,E)\}$$

Thus,  $|S_0| = 4$ ,  $|Q_0| = 1$  and  $T(0) = |Q_0|/|S_0| = 1/4$ .

# Illustrative Example (continued)



Network  $N_1$ 

Network N2

 $S_1$  = Set of pairs (x, y) such that x and y have **one** common neighbor in  $N_1$  and the edge  $\{x, y\}$  is not in  $N_1$ 

$$= \{(A,B), (A,C), (B,C)\}$$

 $Q_1 = Subset of S_1 such that for each pair (x, y) in Q_1$ the edge {x, y} is in N<sub>2</sub> = {(A,B)}

Thus,  $|S_1| = 3$ ,  $|Q_1| = 1$  and  $T(1) = |Q_1|/|S_1| = 1/3$ .

### Study by Kossinets & Watts

#### **Details About the Data Set:**

- Data from email communication between students at a US university. No. of students  $\approx$  22,600.
- Observation period: One year.
- Each student is a node; the edge {x, y} is added when they exchanged email.
- By considering multiple pairs of snapshots of the network, they constructed an average value of *T*(*k*) for each value of *k*.

#### **Results:**

- T(0) (the likelihood of link formation with no common friends) is close to 0.
- Probability of link formation increases with the number of common friends (k).
- Having two common friends increases the likelihood by a factor of more than 2 compared to having one common friend.

### Study by Kossinets & Watts: Triadic Closure



Figure 4.9: Quantifying the effects of triadic closure in an e-mail dataset [259]. The curve determined from the data is shown in the solid black line; the dotted curves show a comparison to probabilities computed according to two simple baseline models in which common friends provide independent probabilities of link formation.

### Comparison with a Baseline Model

**Assumption:** There is a (small) value p such that for each pair of people x and y, each common friend causes the link  $\{x, y\}$  **independently** with probability p.

#### Model Derivation:

- Suppose x and y have  $k \ge 1$  friends in common.
- The probability that they **don't** form a link is  $(1-p)^k$ .
- So, the probability  $T_b(k)$  that they **do** form a link is given by  $T_b(k) = 1 (1-p)^k$ .

#### Notes:

- The plot in Slide 3-21 shows the actual curve sandwiched between  $T_b(k)$  and  $T_b(k-1)$ .
- The value of *p* is chosen so that the model provides a good alignment with the actual curve.
- The plot suggests that the baseline model is reasonable for low values of *k*.

# Focal Closure – Brief Discussion



- Goal: To understand how the likelihood of the link {A, B} depends on the number of common foci.
- Kossinets & Watts used classes as the foci.
- They computed the empirical estimates of the probability values using the methodology discussed in Handout 3.2. (The results are shown in Slide 3-24.)
- When the number of common courses is small, the likelihood of link formation increases.
- A subsequent increase in the number of common courses has a "diminishing returns" effect.

### Study by Kossinets & Watts: Focal Closure



Figure 4.10: Quantifying the effects of focal closure in an e-mail dataset [259]. Again, the curve determined from the data is shown in the solid black line, while the dotted curve provides a comparison to a simple baseline.

# Schelling's Model of Spatial Segregation



■ Thomas Schelling (1921 – )

- Professor Emeritus, University of MD & New England Complex Systems Institute.
- Winner of the 2005 Nobel Prize in Economics (for contributions to Game Theory).

- An easily seen effect of homophily: Racially homogeneous neighborhoods (see Slide 3-26).
- In 1972, Schelling suggested a spatial model to explain this.
  - The model shows that global patterns of segregation can arise due to homophily operating at a local level.
  - These mechanisms operate even when no single individual wants segregation.

# Schelling's Model ... (continued)



(a) Chicago, 1940

(b) Chicago, 1960

Figure 4.14: The tendency of people to live in racially homogeneous neighborhoods produces spatial patterns of segregation that are apparent both in everyday life and when superimposed on a map — as here, in these maps of Chicago from 1940 and 1960 [302]. In blocks colored yellow and orange the percentage of African-Americans is below 25, while in blocks colored brown and black the percentage is above 75.

#### Details of the Model (Game):

- Grid representation for a city, with each cell representing a section of the city.
- There are two types of people (agents), denoted by X and O.
- This agent classification is based on an immutable characteristic.
- Some grid cells have agents while others are empty.



- Each cell may have up to 8 neighbors.
- Boundary cells have fewer neighbors.

# Schelling's Model ... (continued)

**Constraint:** Each agent wants to have **at least** *t* other agents of the same type as its neighbors.

- The parameter t is called the threshold. (For example, t may be chosen as 3.)
- An agent with less than t neighbors of the same type is "unsatisfied" and wants to move to another cell where the threshold is satisfied.

#### **Dynamics of Movement:**

- Movement happens in rounds.
- Actions carried out in each each round:
  - Unsatisfied agents are considered in some (predetermined) order.

Actions in each round (continued):

- Each unsatisfied agent is moved to a cell where the threshold is satisfied. (If there are many possible cells, one is chosen randomly. If there is no such cell, the agent may be moved to a random cell or left where it is.)
- Such movements may cause other agents to become "unsatisfied".
- Rounds are repeated until all agents are satisfied. (The game may never end.)

#### An Example of a Configuration:

X1*	X2*				
ХЗ	O1*		O2		
X4	X5	O3	O4	O5*	
X6*	O6			X7	X8
	07	O8	X9*	X10	X11
		O9	O10	011*	

#### Notes:

- Agents are numbered so that their movements can be readily followed.
- Threshold value t = 3.
- An asterisk is used to indicate an unsatisfied agent.

#### A Subsequent Configuration:

ХЗ	X6	O1	O2		
X4	X5	ОЗ	O4		
	O6	X2	X1	X7	X8
011	07	O8	X9	X10	X11
	O5	O9	O10*		

#### Notes:

- Small examples are useful in understanding the model.
- Several applets are available on the web to try larger examples.

# Schelling's Model ... (continued)

#### **Results from an Applet:** (Threshold = 4)



(c) After 350 steps

(d) After 800 steps

Figure 4.19: Four intermediate points in a simulation of the Schelling model with a threshold t of 4, on a 150-by-150 grid with 10,000 agents of each type. As the rounds of movement progress, large homogeneous regions on the grid grow at the expense of smaller, narrower regions.

### Observations Regarding Schelling's Model

- Segregation takes place (at the global level) even though no agent is actively seeking it. (Each agent is willing to have a minority of neighbors of its type.)
- Segregation is not built into the model; there are patterns where there is not much segregation. (However, empirical evidence suggests that such patterns are hard to reach from random initial configurations.)
- In real life, segregation effect is amplified by genuine desire on the part of a small fraction of people who want to avoid other types of people.
- Schelling's work also suggests that immutable characteristics (e.g. race, ethnicity) may be highly correlated with certain mutable characteristics (e.g. decision about where to live).