

**Degree distribution:**  $P(k)$  **Avg. degree:**  $\bar{k} = \langle k \rangle = \frac{1}{N} \sum_{i=1}^N k_i = \frac{2E}{N}$

**Path length:**  $h$  In directed networks we define an **in-degree** and **out-degree**.

**Clustering coefficient:**  $C$  The (total) degree of a node is the sum of in- and out-degrees.

**Connected components:**  $s$

The **maximum number of edges** in an undirected graph on  $N$  nodes is  $E_{\max} = \binom{N}{2} = \frac{N(N-1)}{2}$

**Degree distribution  $P(k)$ :** Prob. a randomly chosen node has  $k$  neighbors.  $N_k = \#$  nodes with degree  $k$

Normalized histogram:  $P(k) = N_k / N \rightarrow$  plot

An **undirected** graph with the number of edges  $E = E_{\max}$  is called a **complete graph** and its **average degree is  $N-1$**

**Distance (shortest path, geodesic)** between a pair of nodes is defined as the **number of edges** along the **shortest path** connecting the nodes

\* If the two nodes are **not connected**, the distance is usually defined as **infinite (or zero)**

In **directed graphs**, paths need to follow the **direction of the arrows**

**Consequence:** Distance is **not symmetric**:  $h_{B,C} \neq h_{C,B}$

**Diameter:** The **maximum (shortest path) distance** between **any pair** of nodes in a graph

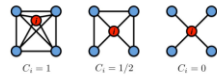
**Clustering coefficient (for undirected graphs)**

How connected are  $i$ 's neighbors to each other?

Node  $i$  with degree  $k_i$

$C_i \in [0, 1]$

$C_i = \frac{2e_i}{k_i(k_i - 1)}$  where  $e_i$  is the number of edges between the neighbors of node  $i$



Note  $k_i(k_i - 1)$  : max number of edges between  $k_i$  neighbors

Clustering coeff is undefined (or = 0) for node degree 0 or 1

**Average clustering coefficient:**  $C = \frac{1}{N} \sum_i C_i$

**Bridge edge:** If we erase the edge, the graph becomes disconnected

**Articulation node:** If we erase the node, the graph becomes disconnected

$G_p$ : undirected graph on  $n$  nodes where each edge  $(u, v)$  appears i.i.d. with probability  $p$

**Largest component = Giant component**

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**Triadic closure = High clustering coefficient**

**How to find connected components:**

- Start from random node and perform Breadth First Search (BFS)
- Label the nodes that BFS visits
- If all nodes are visited, the network is connected
- Otherwise find an unvisited node and repeat BFS

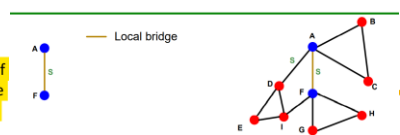
**Bridge** describes an edge's role where deleting this edge would **cause its corresponding vertices falling in different components**;

An edge is a **local bridge** if its **end points** have **no friends in common**

(Alternatively, distance perspective?)

**Span** of a local bridge: the **distance its endpoints** would be from each other if the edge were deleted

If a node A in a network satisfies the STC property and is involved in at least two strong ties, then any local bridge it is involved in must be a weak tie



If a node A has edges to nodes B and C, then the B-C edge is especially likely to form if A's edges to B and C are both strong ties.

More formally, as Granovetter suggested:

A node A violates the **Strong Triadic Closure property** if it has strong ties to two other nodes B and C, and there is no edge at all (either a strong and weak tie) between B and C.

**Probabilistic relational classifier**

Repeat for each node  $i$  and label  $c$

$$P(Y_i = c) = \frac{1}{|N(i)|} \sum_{j \in N(i)} W(i, j) P(Y_j = c)$$

- $W(i, j)$  is the edge strength from  $i$  to  $j$
- $N_i$  is the number of neighbors of  $i$

**Structural balance property:** For every set of three nodes, if we consider the three edges connecting them, either all three of these edges are labeled +, or else exactly one of them is labeled +.

**Homophily test:** If the fraction of **cross-attributes edges** is **significantly less than  $2pq$** , then there is evidence for homophily.

**Balance theorem:** If a labeled complete graph is balanced, then either all pairs of nodes are friends, or else the nodes can be divided into two groups, X and Y, such that every pair of nodes in X like each other, every pair of nodes in Y like each other, and everyone in X is the enemy of everyone in Y.

	Stag	Hare	q-mix
Stag	4, 4	0, 3	4q, 4q+3(1-q)
Hare	3, 0	3, 3	3q+3(1-q), 3(1-q)
p-mix	4p+3(1-p), 4p	3(1-p), 3p+3(1-p)	计算q用player1的等式

A choice of strategies – one by each player – is **Pareto-optimal** if there is no other choice of strategies in which all players receive payoffs at least as high, and at least one player receives a strictly higher payoff.

**Pareto-socially optimal** 和是 Pareto efficient, dominant

**Social optimality:** A choice of strategies – one by each player – is a social welfare maximizer (or socially optimal) if it **maximizes the sum of the player's payoffs**

Suppose  $n$  numbers are drawn independently from the uniform distribution on the interval  $[0, 1]$  and then sorted from smallest to largest. The expected value of the number in the  $k$ th position on this list is  $\frac{k}{n+1}$ .

Now, if the seller runs a second-price auction, and the bidders follow their dominant strategies and bid truthfully, the seller's expected revenue will be the expectation of the second-highest value. Since this will be the value in position  $n-1$  in the sorted order of the  $n$  random values from smallest to largest, the expected value is  $(n-1)/(n+1)$ , by the formula just described. On the other hand, if the seller runs a first-price auction, then in equilibrium we expect the winning bidder to submit a bid that is  $(n-1)/n$  times her true value. Her true value has an expectation of  $n/(n+1)$  (since it is the largest of  $n$  numbers drawn independently from the unit interval), and so the seller's expected revenue is

$$\left(\frac{n-1}{n}\right) \left(\frac{n}{n+1}\right) = \frac{n-1}{n+1}$$

The two auctions provide exactly the same expected revenue to the seller!

**Price of Anarchy (POA):** the ratio between the system performance with **strategic players** and the **best-possible** system performance

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**Perfect Matching:** When there are an equal number of nodes on each side of a bipartite graph, a perfect matching is an assignment of nodes on the left to nodes on the right, in such a way that

- each node is connected by an edge to the node it is assigned to, and
- no two nodes on the left are assigned to the same node on the right.

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**Revenue equivalence theorem:** given certain conditions, any mechanism that results in the same outcomes (i.e. allocates items to the same bidders) also has the same expected revenue.

**Stationary strategy:** the same split in every period in which they are scheduled to propose

- The split  $(a_1, b_1)$  that A will offer whenever he is scheduled to propose a split;
- the split  $(a_2, b_2)$  that B will offer whenever she is scheduled to propose a split; and
- reservation amounts  $a'$  and  $b'$ , constituting the minimum offers that A and B respectively will accept from the other.

A will offer B the least he can get B to accept his offer

$b_1 = b'$

B will offer A the least he can get A to accept his offer:

$a_2 = a'$

B's reservation amount is the indifference amount between accepting A's offer and rejecting A's offer

Accept:  $b_1$

Reject:  $py(1-p)b_2$

Indifference:  $b_1 = py(1-p)b_2$  (1)

Similarly,  $a_2 = px(1-p)a_1$  (2)

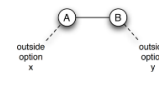
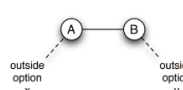
Solving (1), (2),

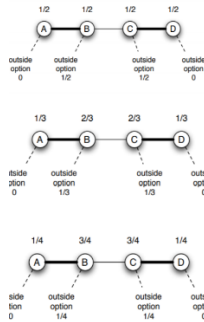
$a_1 = \frac{(1-p)x+1-y}{2-p}$ ,  $b_1 = 1-a_1 = \frac{y+(1-p)(1-x)}{2-p}$

**Nash Bargaining Solution:** When A and B negotiate over splitting a dollar, with an outside option of  $x$  for A and an outside option of  $y$  for B (and  $x + y \leq 1$ ), the Nash bargaining outcome is

- $x + \frac{1}{2} \frac{x+1-y}{2}$  to A, and
- $y + \frac{1}{2} \frac{y+1-x}{2}$  to B

**Surplus:**  $s = 1 - x - y$

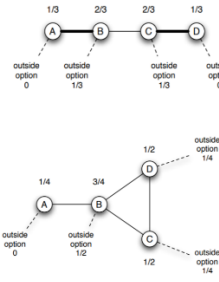




In any network with a stable outcome, there is a balanced outcome

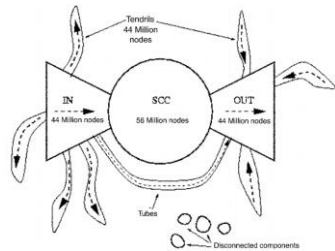
Balance can be used as a way to refine stable outcomes to align with experimental results.  
Nash bargaining solution: captures the weak power advantages, and some fairness

Computational issue



Nash bargaining solution: captures the weak power advantages, and subtle differences because of the network structure

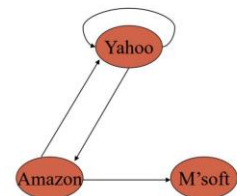
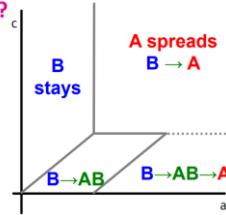
Cooperative game theory: how a collection of players will divide up the value arising from activity



In: new pages  
Out: cooperate pages that links to private pages.  
Tendrils: (1) nodes that reachable from IN but cannot reach the SCC  
(2) nodes that reach OUT but cannot be reached from SCC  
Tubes: Both (1) (2)

B is the default throughout the network until new/better A comes along. What happens?

- Infiltration:** If B is too compatible then people will take on both and then drop the worse one (B)
- Direct conquest:** If A makes itself not compatible – people on the border must choose. They pick the better one (A)
- Buffer zone:** If you choose an optimal level then you keep a static “buffer” between A and B

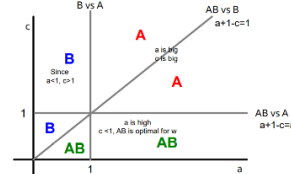
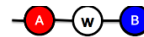


	0.8	$\begin{bmatrix} 1/2 & 1/2 & 0 \\ 1/2 & 0 & 0 \\ 0 & 1/2 & 0 \end{bmatrix}$	+ 0.2	$\begin{bmatrix} 1/3 & 1/3 & 1/3 \\ 1/3 & 1/3 & 1/3 \\ 1/3 & 1/3 & 1/3 \end{bmatrix}$
y		$\begin{bmatrix} 7/15 & 7/15 & 1/15 \\ 7/15 & 1/15 & 1/15 \\ 1/15 & 7/15 & 1/15 \end{bmatrix}$		
a				
m				

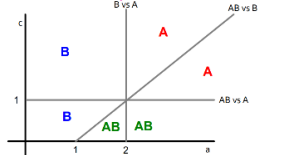
Infinite path, start with Bs

Payoffs for w: A:a, B:1, AB:a+1-c

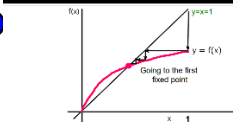
What does node w adopt?



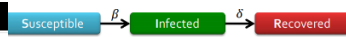
Same reward structure as before but now payoffs for w change: A:a, B:1+1, AB:a+1-c  
Notice: Now also AB spreads  
What does node w adopt?



Fixed Point:  $f(x) = 1 - (1 - qx)^d$



What do we know about the shape of  $f(x)$ ?  
•  $f(0) = 0$   
•  $f(1) = 1 - (1 - q)^d < 1$   
•  $f'(x) = q \cdot d(1 - qx)^{d-1}$   
•  $f'(0) = q \cdot d$   
 $f'(x)$  is monotone non-increasing on  $[0, 1]$



- Each time unit, any infected individual contact  $k$  other individuals
- The probability of infection on contact:  $\pi$
- Per unit of time, contact  $kS/N$  susceptible individuals
- Infect  $\pi kS/N$  individuals
- Transmission rate:  $\beta = \pi k$

$$R_0 = \frac{\beta}{\delta} = \frac{k\pi}{\delta}$$

$\delta$ : Expected number of secondary cases caused by a single primary infected individual

Reduce the infectious period  $\frac{1}{\delta} \rightarrow$  therapeutics

Reduce the  $\beta$

Reduce  $\pi$ : sewage systems, hand-washing, air filters, and so on  
mitigating the number of contagious particles that are exchanged among individuals.

Reduce  $k$ : quarantines, social distancing and travel, restrictions

Reduce the contact rate

$$\mathcal{L} = \sum_{u \in V} \sum_{v \in N_H(u)} -\log(P(v|z_u))$$

Intuition: Optimize embeddings to maximize likelihood of random walk co-occurrences

Parameterize  $P(v|z_u)$  using softmax:

$$P(v|z_u) = \frac{\exp(z_u^T z_v)}{\sum_{n \in V} \exp(z_u^T z_n)}$$

Why softmax?  
We want node  $v$  to most similar to node (out of all nodes  $n$ ).  
Intuition:  $\sum_{n \in V} \exp(z_u^T z_n)$

Estimate of  $R_0$ :

Estimating  $q$ : Given an infected node count the proportion of its neighbors subsequently infected and average

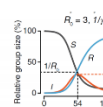
$$R_0 = q \cdot d \cdot \frac{\text{avg}(d_i^2)}{(\text{avg } d_i)^2}$$

Empirical  $R_0$

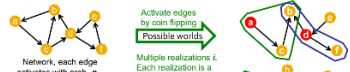
Correction factor due to skew degree distribution of the net

Model dynamics are:

$$\frac{dS}{dt} = -\frac{\beta SI}{N} \quad \frac{dR}{dt} = \delta I$$



Find most influential set  $S$  of size  $k$ : largest expected cascade size  $f(S)$  if set  $S$  is activated



Want to solve:

$$\arg \max_{|S|=k} f(S) = \frac{1}{|I|} \sum_{i \in I} f_i(S)$$

Consider  $S=(a,d)$  then:  
 $f_1(S)=5, f_2(S)=4, f_3(S)=3$   
and  $f(S) = 1/3 \cdot (5+4+3)=4$

influence set of node  $a$   
influence set of node  $d$