

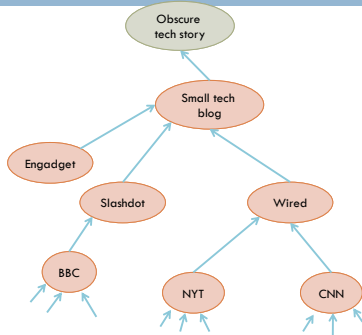
LECTURE 8: NETWORK EFFECTS AND CASCADING BEHAVIOR

CSWP4641: Social Information Network Analysis and Engineering
Wednesday March 11th 2015

Spreading Through Networks

- **Spreading through networks:**
 - Cascading behavior
 - Diffusion of innovations
 - Network effects
 - Epidemics
- **Behaviors that cascade from node to node like an epidemic**
- **Examples:**
 - **Biological:**
 - Diseases via contagion
 - **Technological:**
 - Cascading failures
 - Spread of information
 - **Social:**
 - Rumors, news, new technology
 - Viral marketing

Information Diffusion

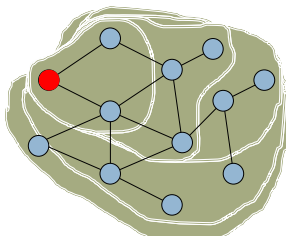


Diffusion in Viral Marketing

- **Product adoption:**
 - **Senders and followers of recommendations**

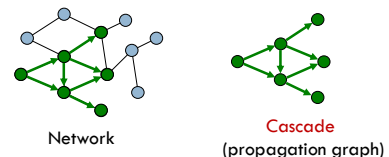


Spread of Diseases



Network Cascades

- **Contagion that spreads over the edges of the network**
- It creates a propagation tree, i.e., **cascade**

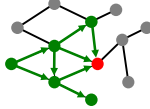


Terminology:

- Stuff that spreads: Contagion
- "Infection" event: Adoption, infection, activation
- We have: Infected/active nodes, adoptors

How to Model Diffusion?

- 7 **Probabilistic models:**
 - Models of influence or disease spreading
 - An infected node tries to “push” the contagion to an uninfected node
 - Example:**
 - You “catch” a disease with some prob. from each active neighbor in the network
- Decision based models (today!):**
 - Models of product adoption, decision making
 - A node observes decisions of its neighbors and makes its own decision
 - Example:**
 - You join demonstrations if k of your friends do so too



DECISION BASED MODEL OF DIFFUSION

Decision Based Models

- 9 **Two ingredients:**
 - Payoffs:**
 - Utility of making a particular choice
 - Signals:**
 - Public information:
 - What your network neighbors have done
 - (Sometimes also) Private information:
 - Something you know
 - Your belief
- Now you want to make the optimal decision**



Game Theoretic Model of Cascades

- 10 **Based on 2 player coordination game**
 - 2 players – each chooses technology A or B
 - Each person can only adopt **one** “behavior”, A or B
 - You gain more payoff if your friend has adopted the **same** behavior as you



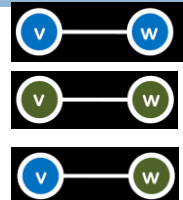
Local view of the network of node v

Example: BlueRay vs. HD DVD

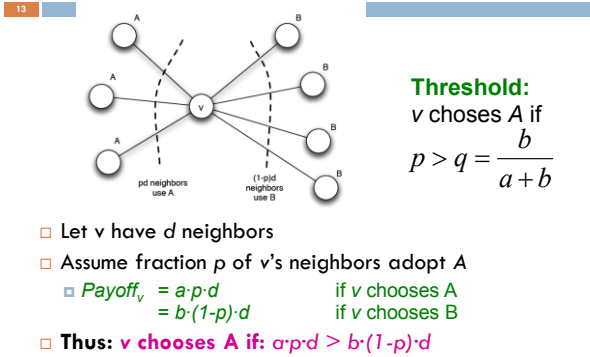


The Model for Two Nodes

- 12 **Payoff matrix:**
 - If both v and w adopt behavior A, they each get payoff $a > 0$
 - If v and w adopt behavior B, they each get payoff $b > 0$
 - If v and w adopt the opposite behaviors, they each get 0
- In some large network:**
 - Each node v is playing a copy of the game with each of its neighbors
 - Payoff:** sum of node payoffs per game



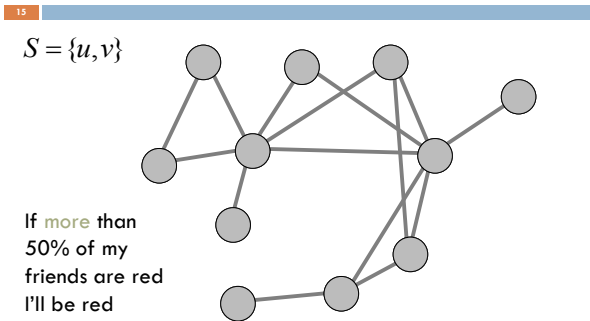
Calculation of Node v



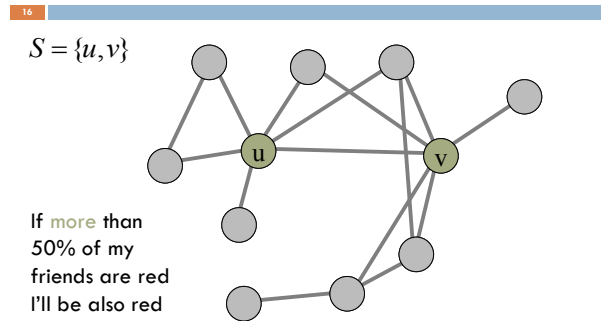
Example Scenario

- 14
- Scenario:** Graph where everyone starts with B. Small set S of early adopters of A
 - Hard-wire S – they keep using A no matter what payoffs tell them to do
 - Assume payoffs are set in such a way that nodes say:
If more than 50% of my friends take A I'll also take A
 (this means: $a = b \cdot \epsilon$ and $q > 1/2$)

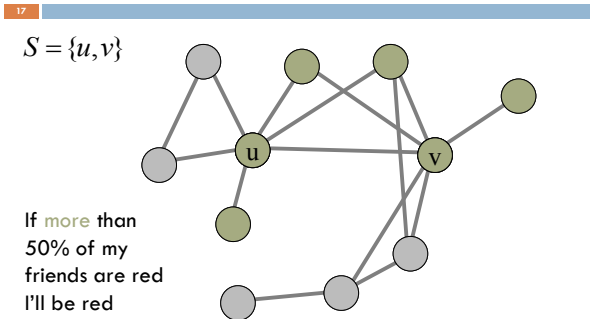
Example Scenario



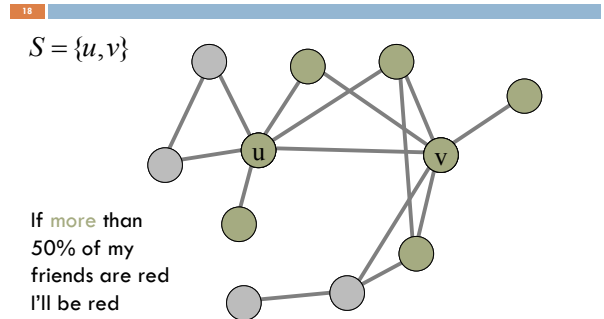
Example Scenario



Example Scenario



Example Scenario

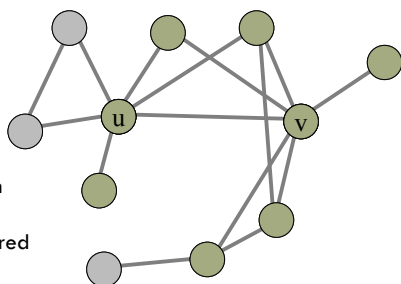


Example Scenario

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$$S = \{u, v\}$$

If **more** than
50% of my
friends are red
I'll be red

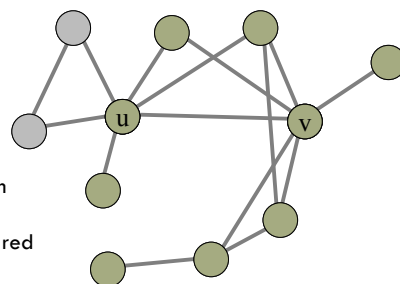


Example Scenario

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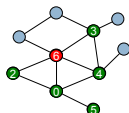
Monotonic Spreading

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- **Observation: Use of A spreads monotonically**
(Nodes only switch $B \rightarrow A$, but never back to B)

- **Why? Proof sketch:**

- **Nodes keep switching from B to A: $B \rightarrow A$**
- Now, suppose some node switched back from $A \rightarrow B$, consider the **first** node u to do so (say at time t)
- Earlier at some time t' ($t' < t$) the same node u switched $B \rightarrow A$
- So at time t' u was above threshold for A
- But up to time t no node switched back to B, so node u could only had more neighbors who used A at time t compared to t' .
There was no reason for u to switch.
!! Contradiction !!



Infinite Graphs

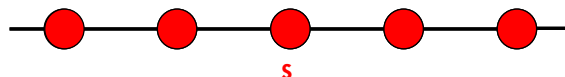
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$$v \text{ chooses A if } p > q$$

$$q = \frac{b}{a+b}$$

- **Consider infinite graph G**
□ (but each node has finite number of neighbors!)
- We say that a finite set S **causes a cascade** in G with threshold q if, when S adopts A, eventually **every** node adopts A
- **Example: Path**

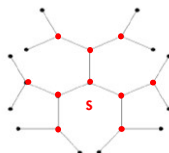
If $q < 1/2$ then cascade occurs



Infinite Graphs

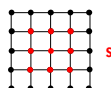
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- **Infinite Tree:**



If $q < 1/3$ then
cascade occurs

- **Infinite Grid:**

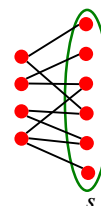


If $q < 1/4$ then
cascade occurs

Cascade Capacity

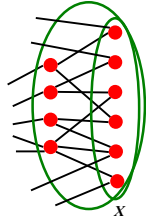
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- **Def:**
 - The **cascade capacity** of a graph G is the **largest q** for which some **finite set S** can cause a **cascade**
- **Fact:**
 - There is no G where cascade capacity $> 1/2$
- **Proof idea:**
 - Suppose such G exists: $q > 1/2$, finite S causes cascade
 - **Show contradiction:** Argue that nodes stop switching after a finite # of steps



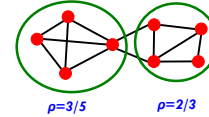
Cascade Capacity

- 25 25
 - **Fact:** There is no G where cascade capacity $> 1/2$
 - **Proof sketch:**
 - Suppose such G exists: $q > 1/2$, finite S causes cascade
 - **Contradiction:** Switching stops after a finite # of steps
 - Define "potential energy"
 - Argue that it starts finite (non-negative) and strictly decreases at every step
 - "Energy": $= |d^{out}(X)|$
 - $|d^{out}(X)| := \#$ of outgoing edges of active set X
 - The only nodes that switch have a strict majority of its neighbors in S
 - $|d^{out}(X)|$ strictly decreases
 - It can do so only a finite number of steps



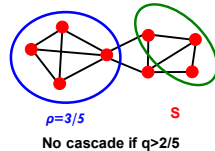
Stopping Cascades

- 26 26
 - What prevents cascades from spreading?
 - **Def:** Cluster of density ρ is a set of nodes C where each node in the set has at least ρ fraction of edges in C .



Stopping Cascades

- 27 27
 - Let S be an initial set of adopters of A
 - All nodes apply threshold q to decide whether to switch to A
 - **Two facts:**
 - 1) If $G \setminus S$ contains a cluster of density $> (1-q)$ then S can not cause a cascade
 - 2) If S fails to create a cascade, then there is a cluster of density $> (1-q)$ in $G \setminus S$



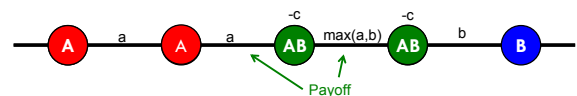
EXTENDING THE MODEL:
ALLOW PEOPLE TO ADOPT A
AND B

Cascades & Compatibility

- 29 29
 - **So far:**
 - Behaviors A and B compete
 - Can only get utility from neighbors of same behavior:
 $A-A$ get a , $B-B$ get b , $A-B$ get 0
 - **Let's add an extra strategy "A-B"**
 - $AB-A$: gets a
 - $AB-B$: gets b
 - $AB-AB$: gets $\max(a, b)$
 - Also: Some cost c for the effort of maintaining both strategies (summed over all interactions)

Cascades & Compatibility: Model

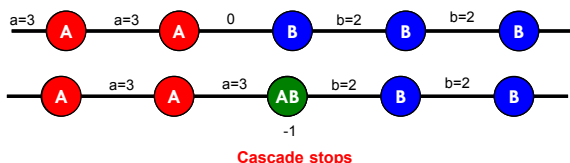
- 30 30
 - Every node in an infinite network starts with B
 - Then a finite set S initially adopts A
 - Run the model for $t=1, 2, 3, \dots$
 - Each node selects behavior that will optimize payoff (given what its neighbors did in at time $t-1$)



- How will nodes switch from B to A or AB ?

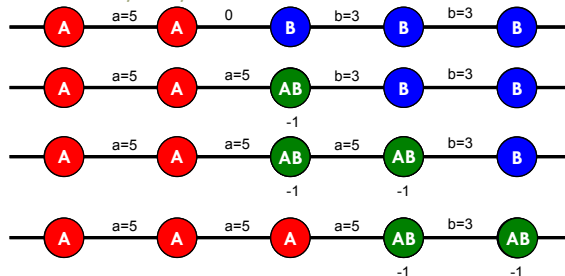
Example: Path Graph

- **Path graph:** Start with all Bs, $a > b$ (A is better)
- **One node switches to A – what happens?**
 - With just A, B: A spreads if $a > b$
 - With A, B, AB: **Does A spread?**
- Assume $a=3, b=2, c=1$:



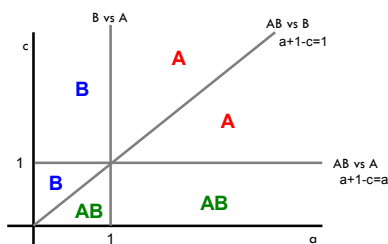
Example

- Let $a=5, b=3, c=1$



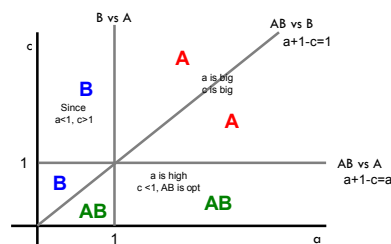
For what pairs (c,a) does A spread?

- **Infinite path, start with all Bs**
- **Payoffs for w:** A: a , B: 1 , AB: $a+1-c$
- What does node w in **A-w-B** do?



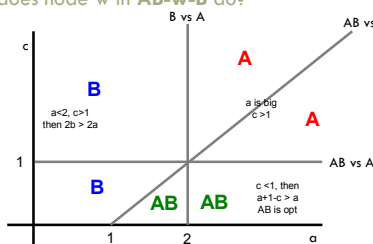
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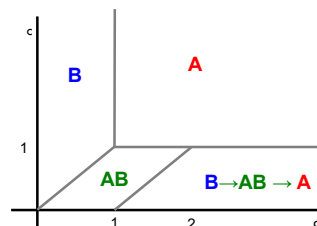
For what pairs (c,a) does A spread?

- **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 in **AB-w-B** do?



For what pairs (c,a) does A spread?

- **Joining the two pictures:**



Lesson

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□ You manufacture default B and new/better A comes along:

- **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

