



Networks Out of Control:
Evolution & Dynamics 1
Cascades



+ Announcement:
Project Selection

+ Project

Details:

- Pair project
- 4 page (2-column) report due Wednesday 23rd May.
- 15 min presentation + 5 min questions on Monday 28th May to be scheduled between 2pm-7pm.


Contact Farnood with the following by Friday 20 April:

- Who your team is (pairs)
 - In case you do not find a team, respond to the other two points and we will match you with someone
- List of 3 preferred papers, in order, on which you would like to do your project (ties get broken by when the email was sent).
- Availability to present in the 2pm-4pm time slot on Monday.

+ Cascades



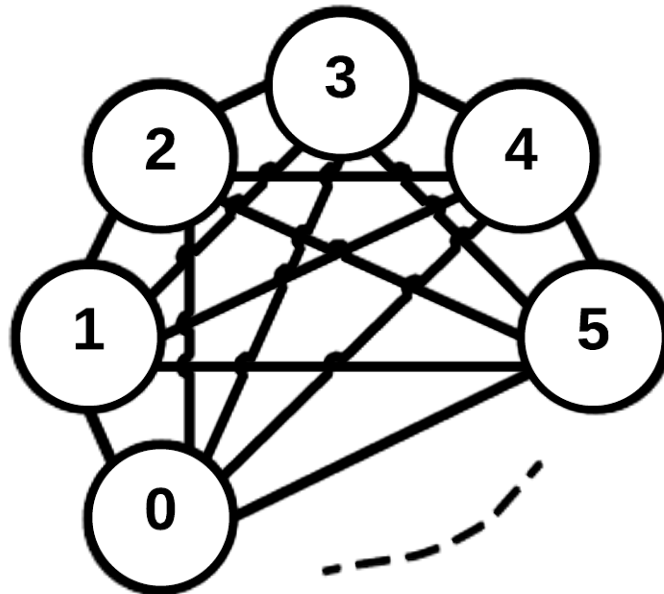
- ◆ There is a network of people, and a behavior that spreads through the network from person to person across edges.
 - ◆ E.g., an idea, product, illness or habit spreads across the network.
 - ◆ (Already discussed this informally in the context of homophily)
- ◆ How does it spread?
 - ◆ Direct-benefit effects
 - ◆ Rational effects
- ◆ What do we study?
 - ◆ Will it spread to the entire network?
 - ◆ Are there threshold properties that determine if/when it does?



+ Cascades 0:
Shelling Threshold Model

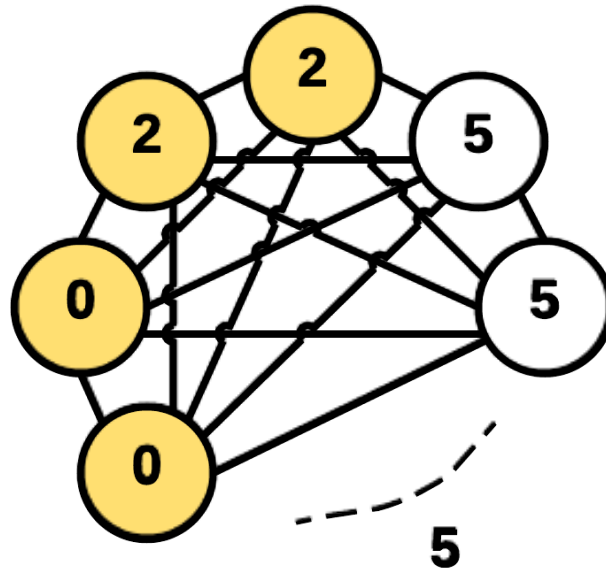
+ Schelling Threshold Model

- ◆ There are n people (labeled $i = 0, 1, 2, \dots, n-1$). Each has a “willingness to riot” coefficient r_i which is *how many others* decide to riot before they join in.
- ◆ Can think of this process as occurring on the complete graph:
 - ◆ Is there a complete riot in this situation?



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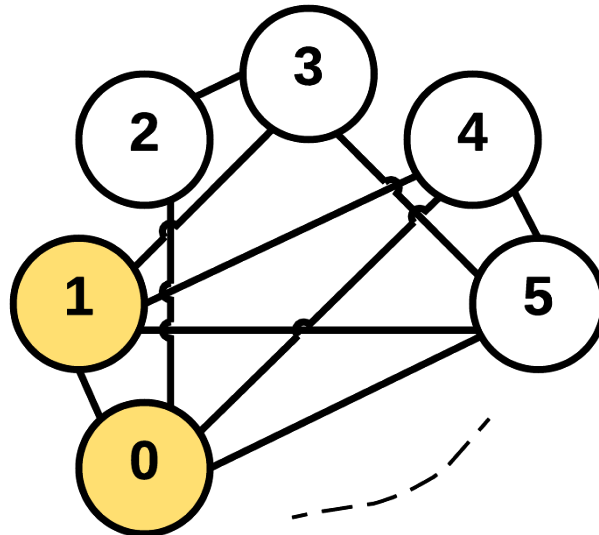
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- ◆ I.e., the existence of a cascade depends on the parameters.

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- ◆ Can think of this process as occurring on arbitrary graphs:
 - ◆ Is there a complete riot in this situation?



- ◆ I.e., the existence of a cascade depends on the network.



+ Cascades 1:
Direct-Benefit Effects



Crash Course in Game Theory

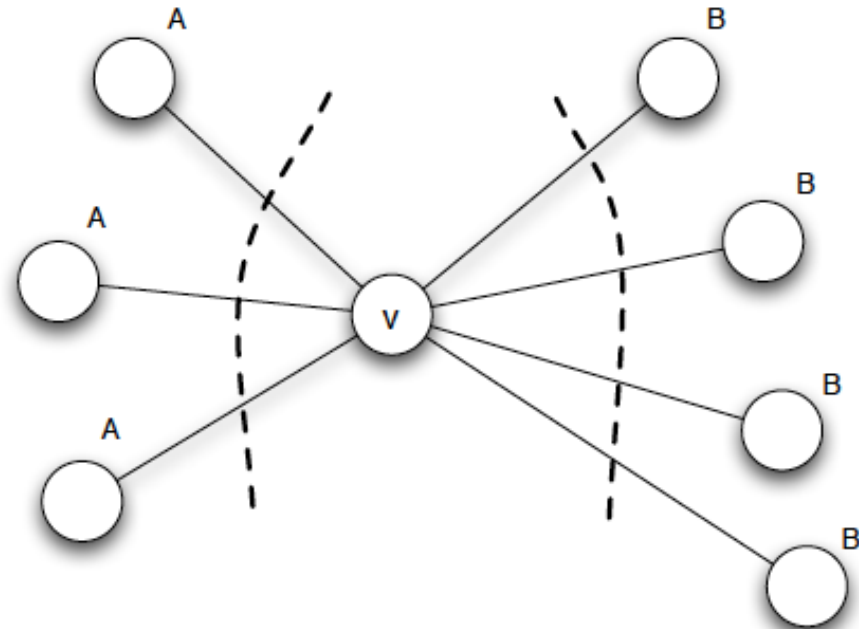


- ◆ **Agents:** two or more participants.
- ◆ **Actions:** options available to the agent.
- ◆ **Outcome:** global end result (function of all agent's actions).
- ◆ **Utility:** real-valued function of outcome for a given agent.

- ◆ In game theory:
 - ◆ Agents want to maximize their own utility (selfish behavior)
 - ◆ We study Nash equilibria, i.e., outcomes such that no agent can unilaterally increase their utility by changing their action.

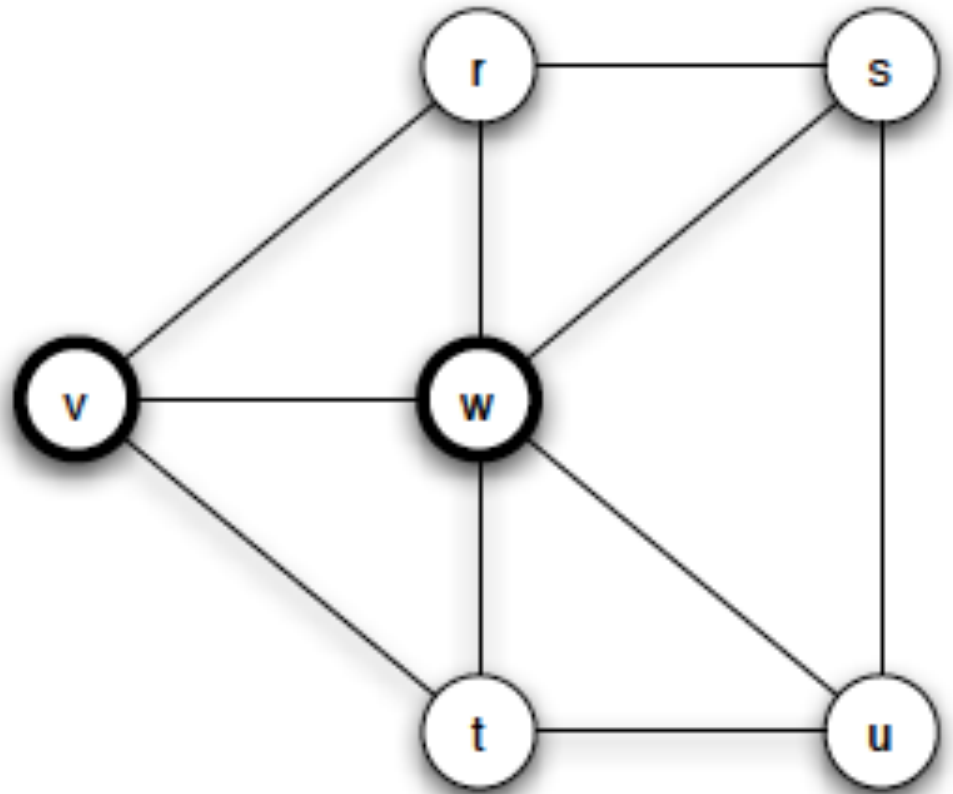
+ Diffusion on Networks

- ◆ There are two products, A and B
- ◆ There is a social network G , and each node (*agents*) can select to use a single product A or B (*actions*).
- ◆ Let $n_A(v)$ be the number of v 's neighbors using A. Similarly $n_B(v)$
- ◆ Let $a, b > 0$. The *utility* for agent v is
 - ◆ If v uses A: $a n_A(v)$
 - ◆ If v uses B: $b n_B(v)$
- ◆ When does v select A instead of B?
 - ◆ $p > b / (a+b)$
- ◆ When does the whole network converge (cascade) to one option?



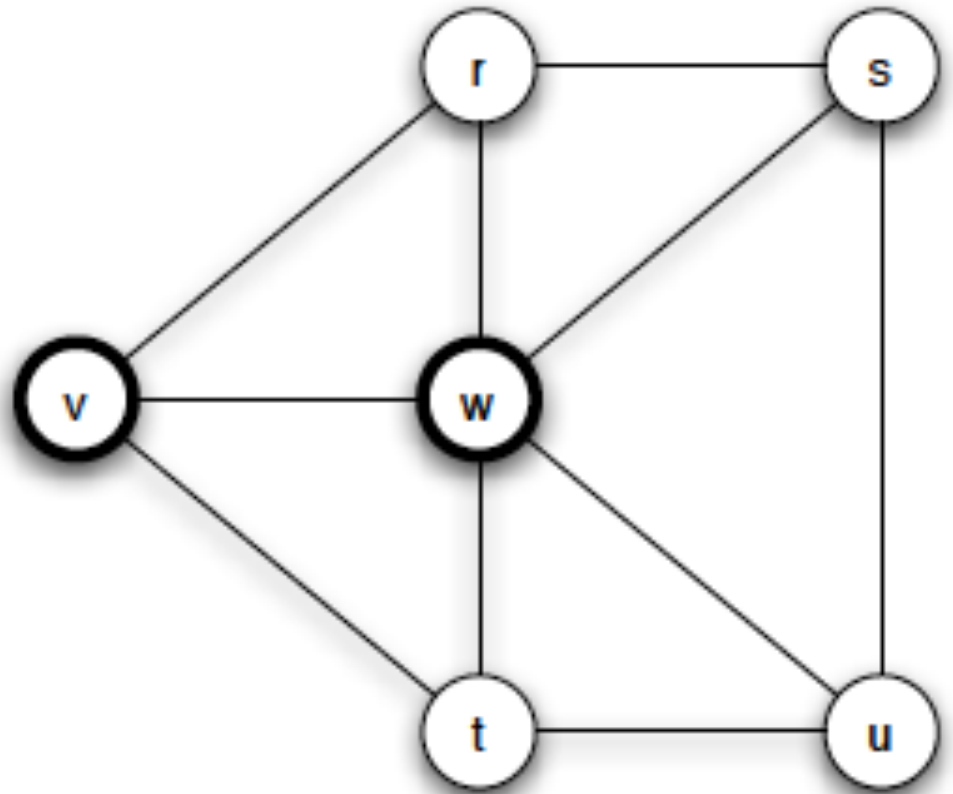
+ Example

- ◆ Let $a = 3$ and $b = 2$
 - ◆ So $p = 2/5$ is the threshold for selecting A



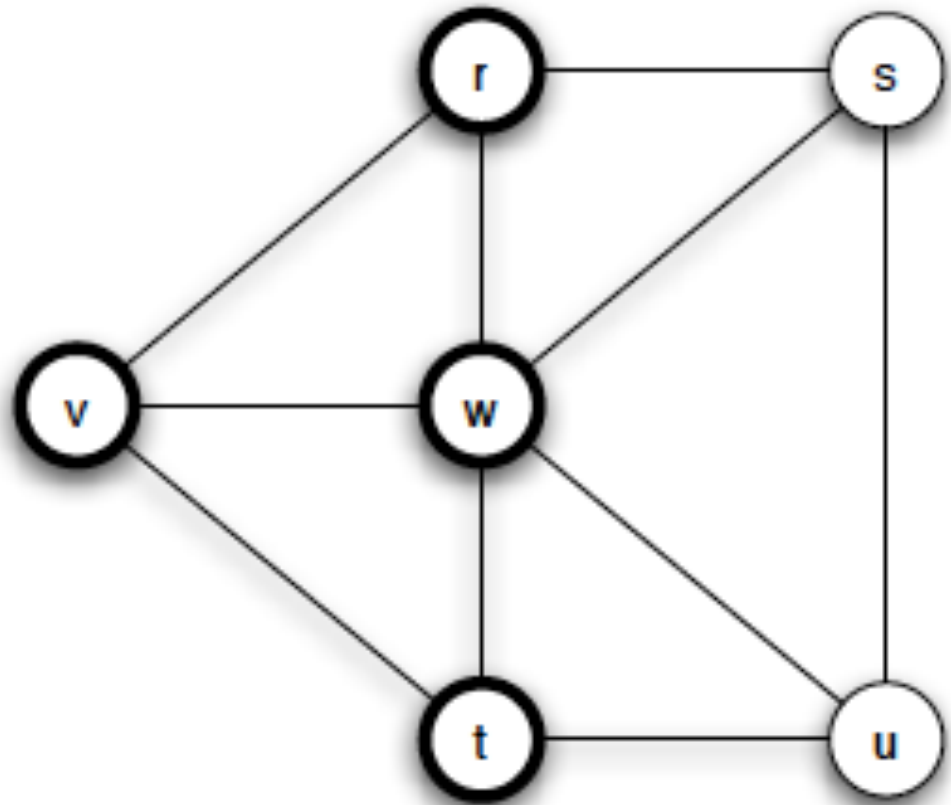
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- ◆ Let $a = 3$ and $b = 2$ (so $p = 2/5$)
- ◆ Is there a complete cascade?



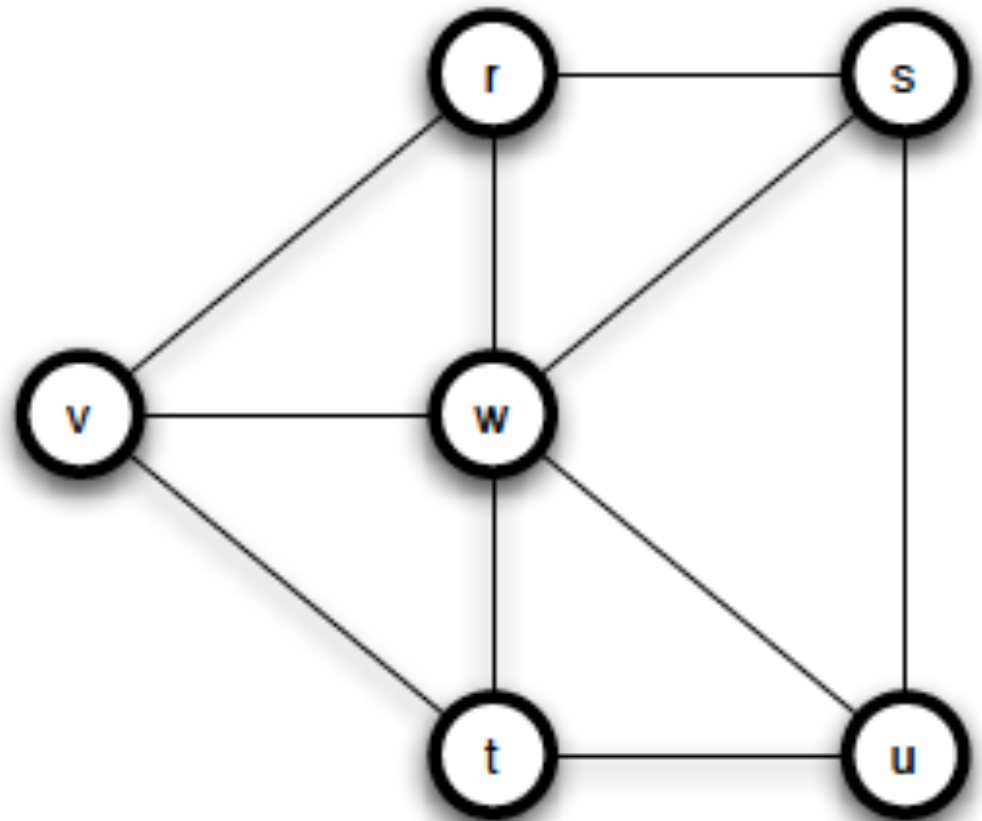
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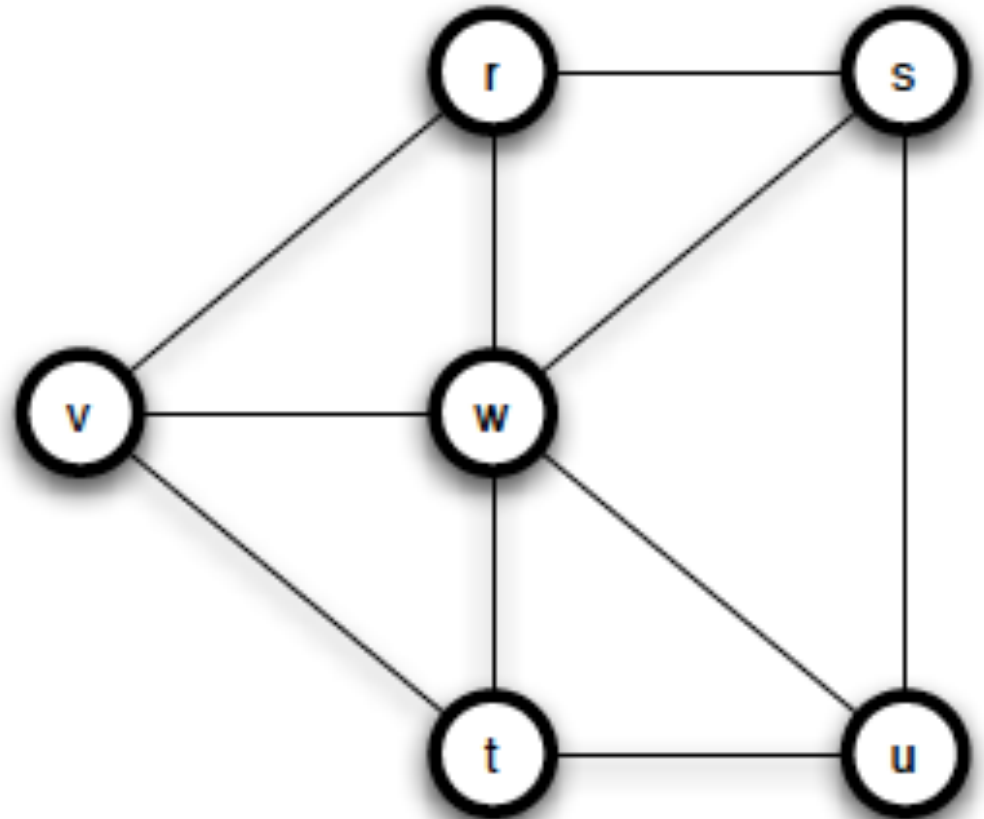
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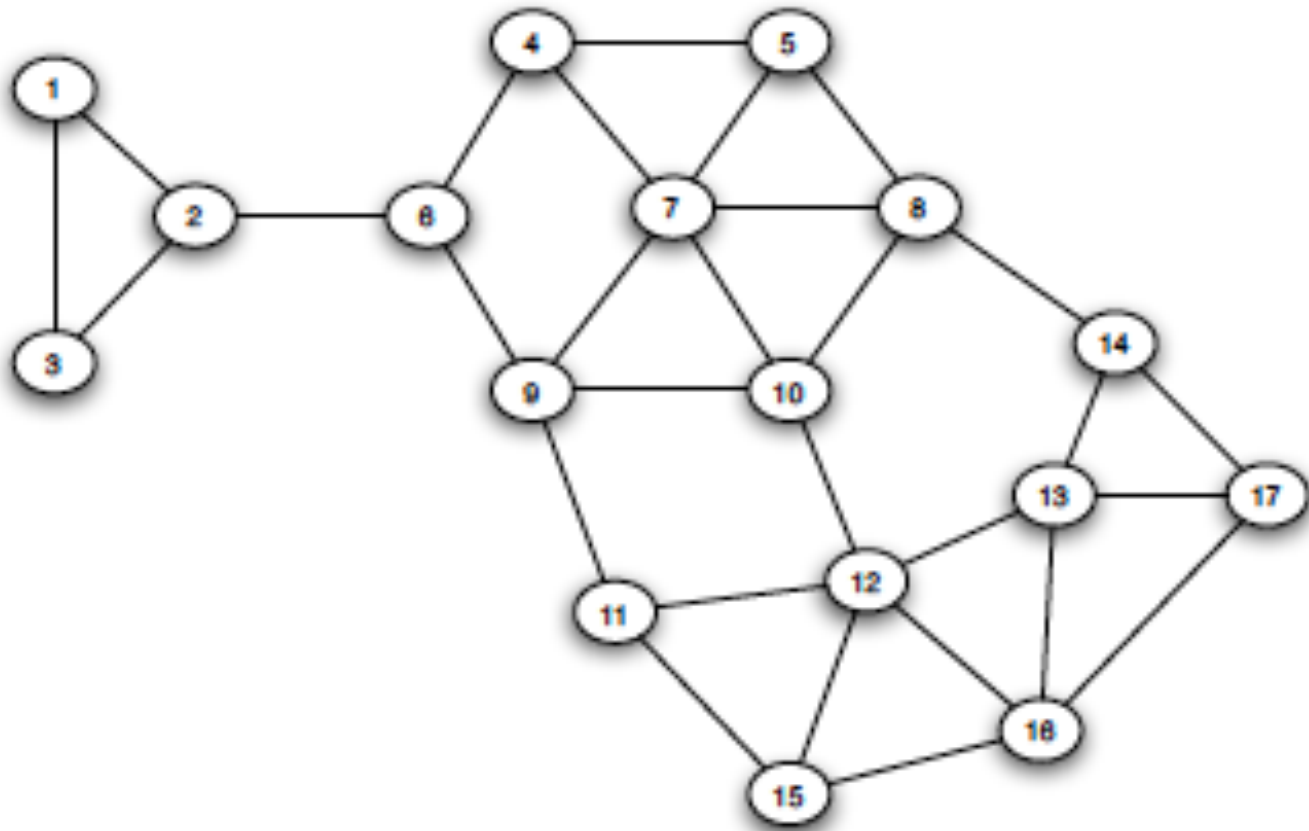
+ Example

- ◆ Let $a = 3$ and $b = 2$ (so $p = 2/5$).
- ◆ Has applications to viral marketing – which nodes should one target (and then hope they influence their neighbors)?
- ◆ Will a cascade always occur?



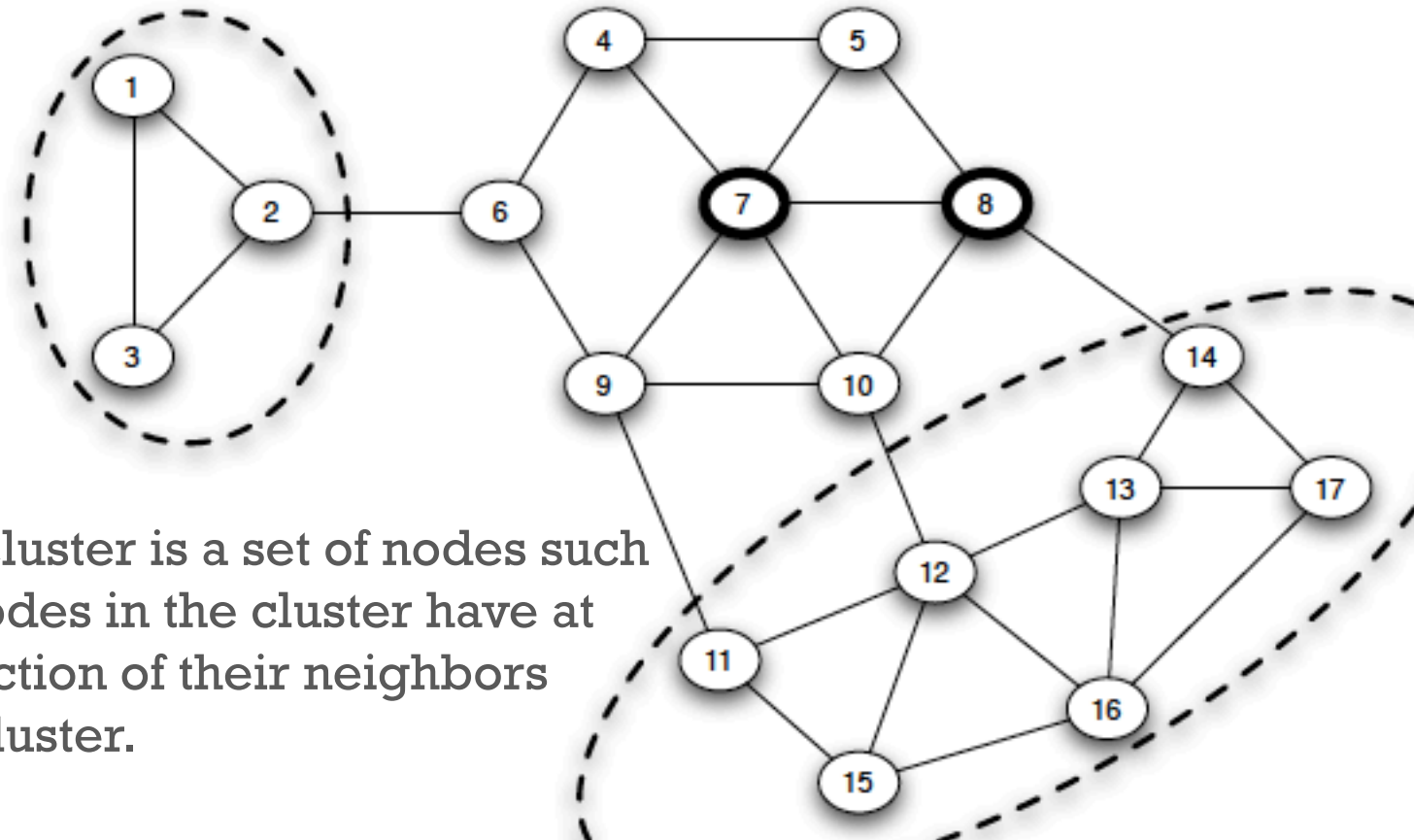
+ Cascades

- ◆ Will a cascade always occur?



+ Cascades

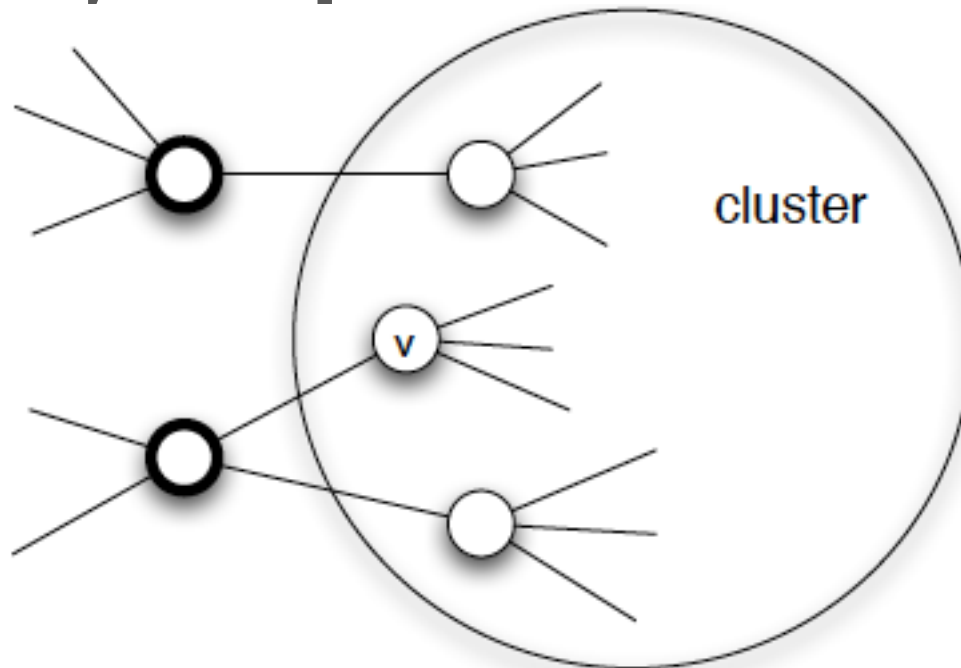
- ◆ Will a cascade always occur?
 - ◆ No – “clusters” can get in the way.



- ◆ A p -dense cluster is a set of nodes such that all of nodes in the cluster have at least a p fraction of their neighbors inside the cluster.

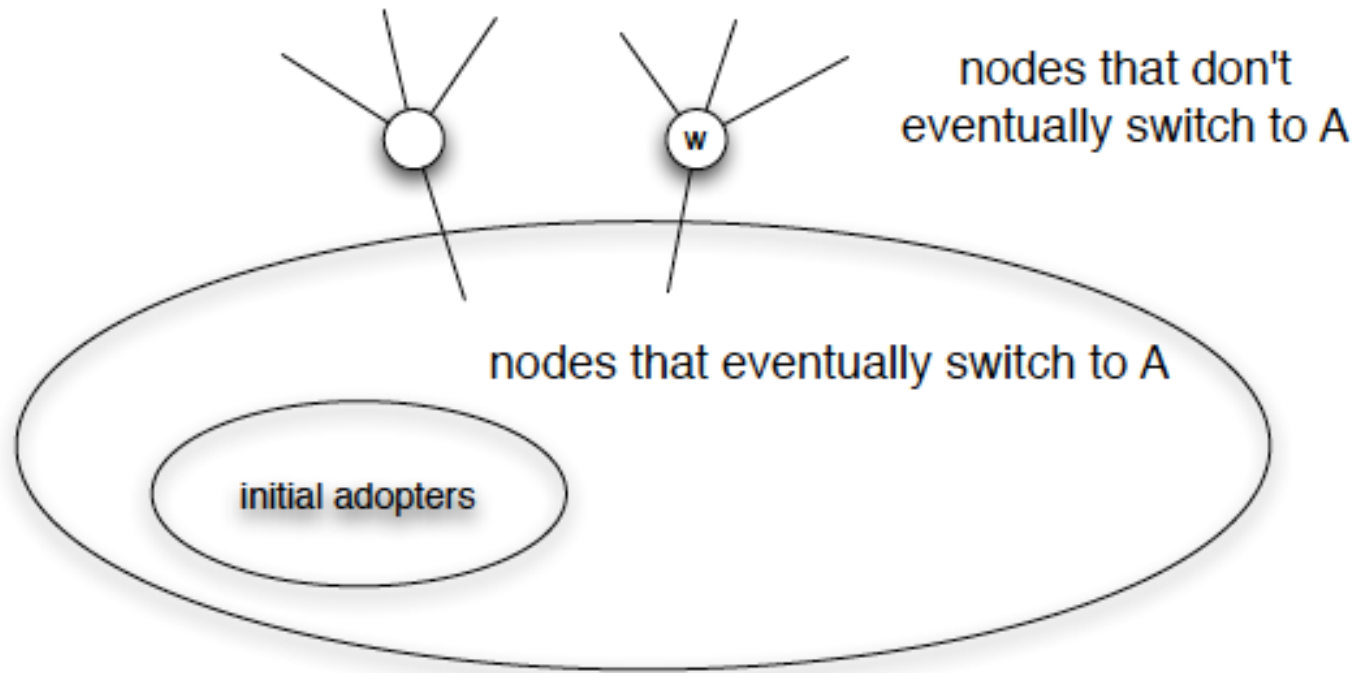
+ Cascades

- ◆ A p -dense cluster is a set of nodes such that all of nodes in the cluster have at least a p fraction of their neighbors inside the cluster.
- ◆ Theorem: Consider an initial set of adopters V' and a threshold p for adoption. The nodes in $V \setminus V'$ contain a cluster of density greater than $1-p$ if and only if a complete cascade does not occur.



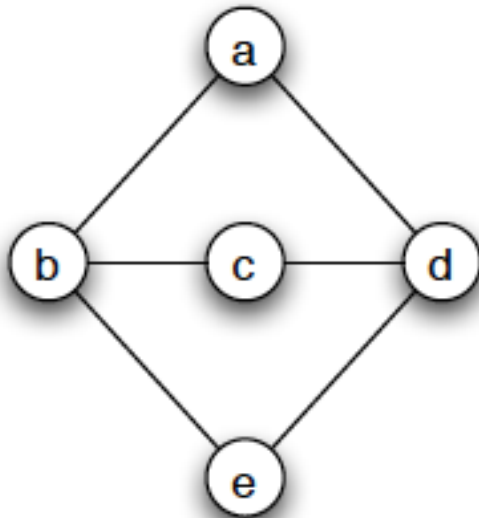
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+ Small Exercise

- If you wanted to initiate a cascade, which vertex would you want to as an initial adopter?





+ Cascades 2:
Rational Effects

+ Information Cascades

- ◆ Two options A and B
 - ◆ You have some (private) information as to which option is better.
 - ◆ You have some public information as to which option other people think is better.

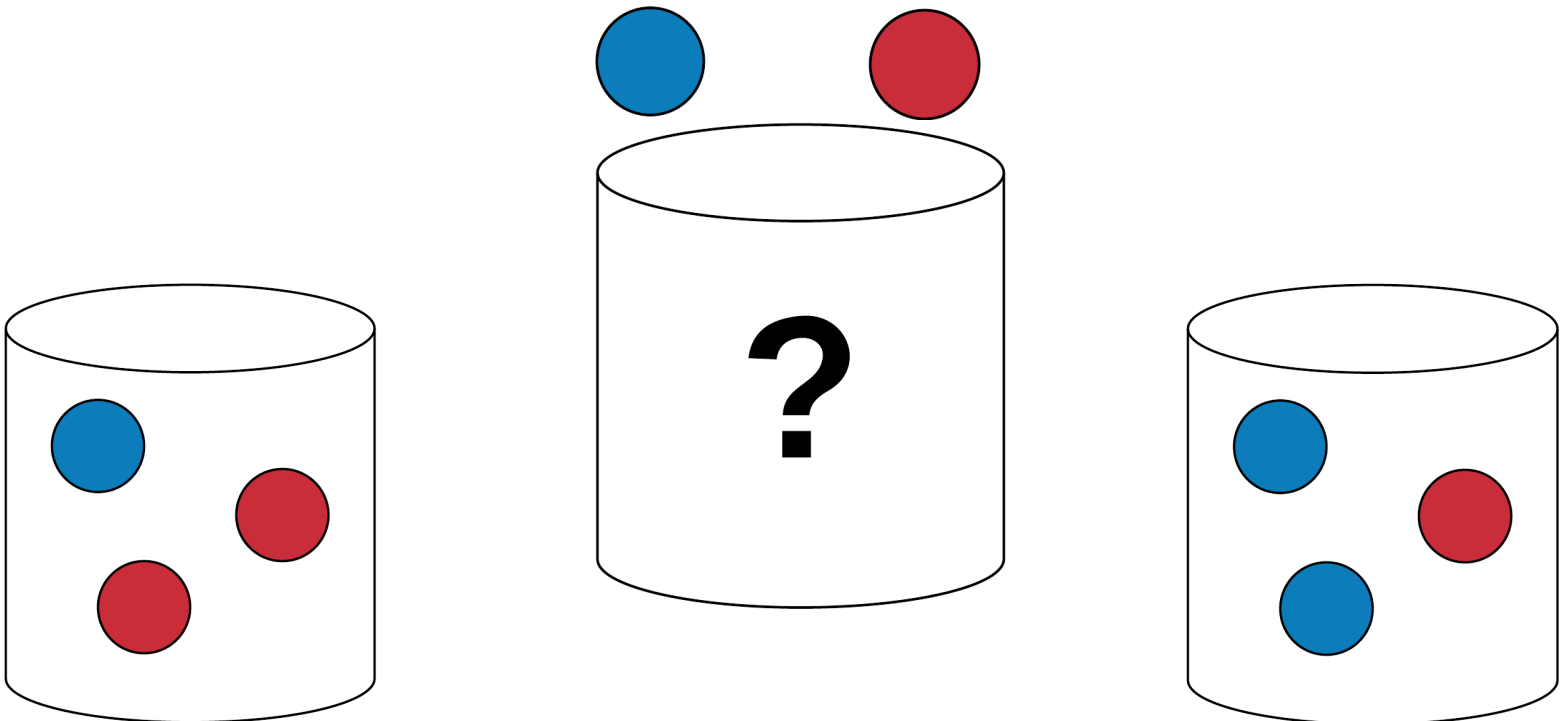


+ Cascades



◆ Example:

- ◆ A = guess red urn, B = guess blue urn
- ◆ Noisy signal = sample from urn
- ◆ Public decision = guess of previous players



+ Cascades



- ◆ Two options A and B
 - ◆ Each player receives a private **noisy signal** that indicates which option is better.
 - ◆ Each player observes sequential public **decisions**.

B

R

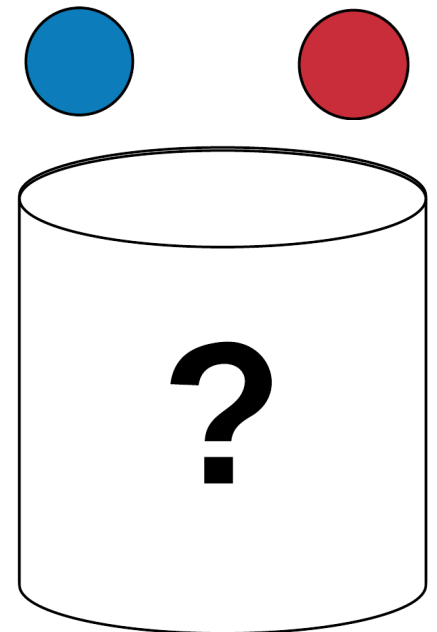
B

B

B

...

B



+ Cascade Example

- ◆ How should players make their decisions?

$$\Pr [\textit{majority-blue} \mid \textit{what she has seen and heard}] > \frac{1}{2}$$

- ◆ In order to analyze this, we'll make heavy use of Bayes' rule:

$$\Pr [A \mid B] = \frac{\Pr [A \cap B]}{\Pr [B]}.$$

- ◆ Note:

$$\Pr [\textit{majority-blue}] = \Pr [\textit{majority-red}] = \frac{1}{2}.$$

$$\Pr [\textit{blue} \mid \textit{majority-blue}] = \Pr [\textit{red} \mid \textit{majority-red}] = \frac{2}{3}.$$

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Cascade Example

$$\Pr[A | B] = \frac{\Pr[A \cap B]}{\Pr[B]}.$$

$$\Pr[\textit{majority-blue}] = \Pr[\textit{majority-red}] = \frac{1}{2}.$$

$$\Pr[\textit{blue} | \textit{majority-blue}] = \Pr[\textit{red} | \textit{majority-red}] = \frac{2}{3}.$$



◆ Consider the first player:

◆ Assume they saw blue (the argument for red is symmetric).

$$\Pr[\textit{majority-blue} | \textit{blue}] =$$

$$\frac{\Pr[\textit{blue} | \textit{majority-blue}] \Pr[\textit{majority-blue}]}{\Pr[\textit{blue}]}$$

$$\Pr[\textit{majority-blue} | \textit{blue}] = \frac{1/3}{1/2} = \frac{2}{3}.$$



Cascade Example

$$\Pr[A | B] = \frac{\Pr[A \cap B]}{\Pr[B]}$$

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$$\Pr[\textit{blue} | \textit{majority-blue}] = \Pr[\textit{red} | \textit{majority-red}] = \frac{2}{3}$$



- ◆ Consider the second player:
 - ◆ If the first player saw blue and they also see blue:
 - ◆ Clearly $P[\textit{majority-blue}] > \frac{1}{2}$ so they say blue
 - ◆ If the first player saw blue and they see red:
 - ◆ Won't do this formally, but with Bayes' rule can prove that $P[\textit{majority-blue}] = P[\textit{majority-red}] = 1/2$
 - ◆ Tiebreak – (let's assume they stick with color they see).

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Cascade Example

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- ◆ Consider the third player:
 - ◆ If the first two see one red and one blue
 - ◆ Reduces to the first-player setting because first two samples were not informative.
 - ◆ If all three see blue
 - ◆ Clearly $P[\textit{majority-blue}] > \frac{1}{2}$, and they say blue.
 - ◆ If the first two see blue and they see red...

+ Cascade Example

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$$\Pr[\textit{blue} | \textit{majority-blue}] = \Pr[\textit{red} | \textit{majority-red}] = \frac{2}{3}.$$



- ◆ Consider the third player:
 - ◆ Assume the first two see blue, but they see red.

$\Pr[\textit{majority-blue} | \textit{blue}, \textit{blue}, \textit{red}] :$

$\Pr[\textit{blue}, \textit{blue}, \textit{red} | \textit{majority-blue}]$

$\Pr[\textit{blue}, \textit{blue}, \textit{red}]$

$$\Pr[\textit{majority-blue} | \textit{blue}, \textit{blue}, \textit{red}] = \frac{\frac{4}{27} \cdot \frac{1}{2}}{\frac{1}{9}} = \frac{2}{3}.$$

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Cascade Example

$$\Pr[A | B] = \frac{\Pr[A \cap B]}{\Pr[B]}.$$

$$\Pr[\textit{majority-blue}] = \Pr[\textit{majority-red}] = \frac{1}{2}.$$

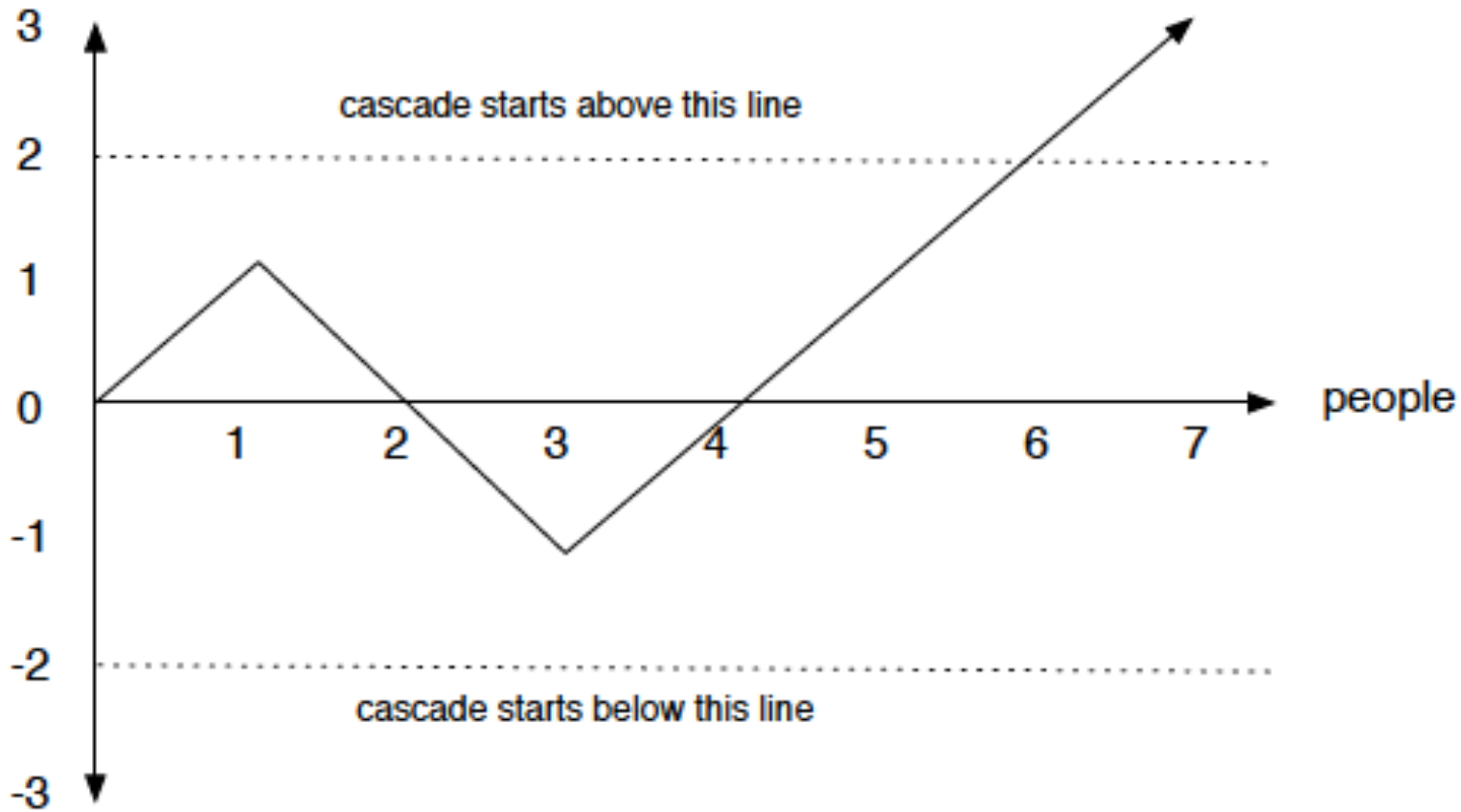
$$\Pr[\textit{blue} | \textit{majority-blue}] = \Pr[\textit{red} | \textit{majority-red}] = \frac{2}{3}.$$



- ◆ Consider the fourth player:
 - ◆ If the first three players say:
 - ◆ blue blue x
 - ◆ blue red x

 - ◆ Reduces to the three-player setting!

+ Cascade Example



+ Information Cascades

- ◆ Generally, cascades
 - ◆ Can lead to sub-optimal outcomes (the crowd may not be wise)
 - ◆ Can be based on very little information
 - ◆ Are fragile
- ◆ Main bottleneck to “wisdom” -- decisions NOT independent.
 - ◆ Also known as “herding”
- ◆ Can think of this process as occurring on a graph.

