



# **Social and Information Networks**

**CSCC46H, Fall 2025**

**Lecture 11**

Prof. Ashton Anderson  
[ashton@cs.toronto.edu](mailto:ashton@cs.toronto.edu)



# Today

**A3 due next Tuesday, Dec 2 @ 11:59pm, plus no-penalty extension to next Friday, Dec 5 @ 11:59pm (late days not needed)**



# Today

**Epidemics and Contagion  
Voting**



# Epidemics

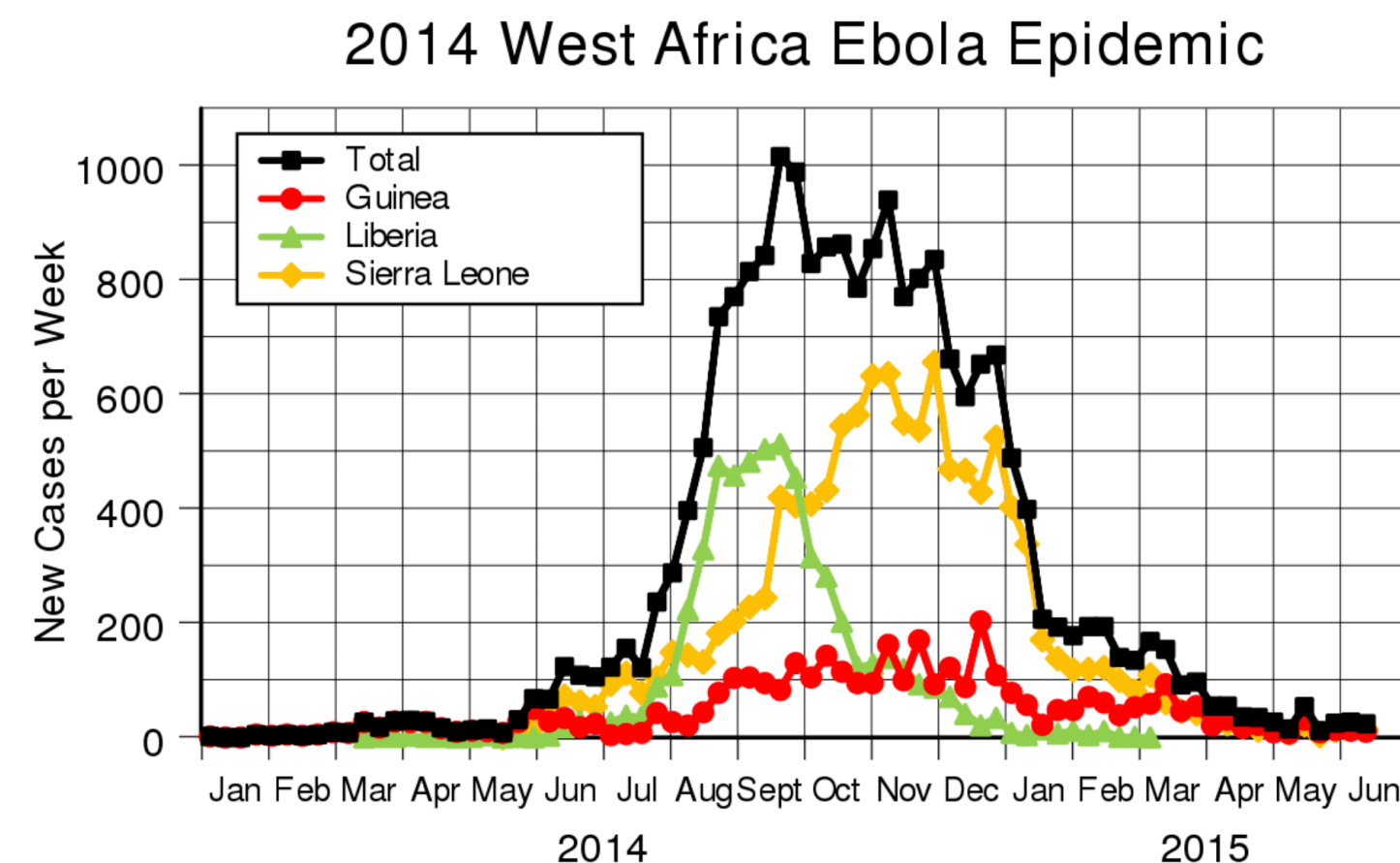


# Epidemics

Why study epidemics in a computer science class?

Epidemics are diseases that **travel socially**

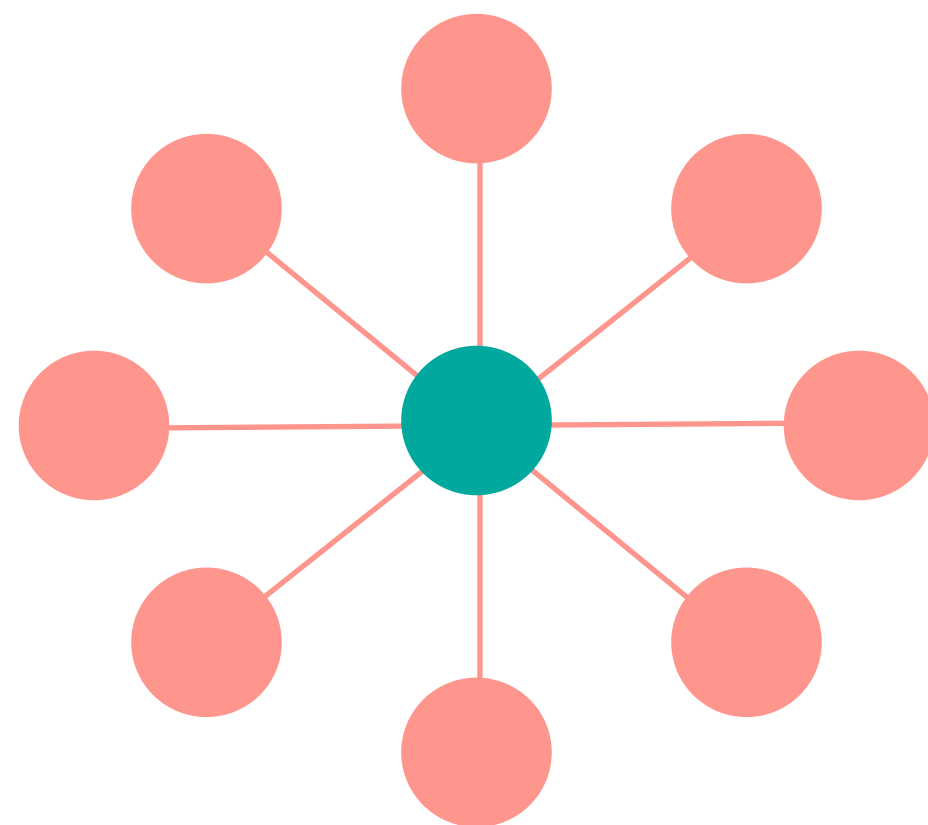
The **structure of social interaction networks** determine the spread of disease



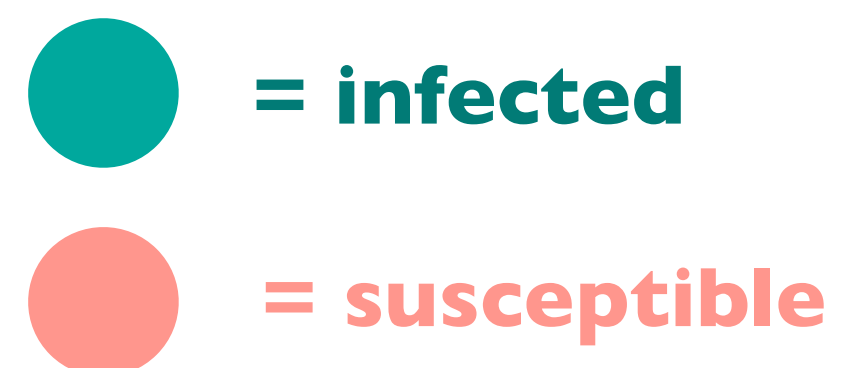
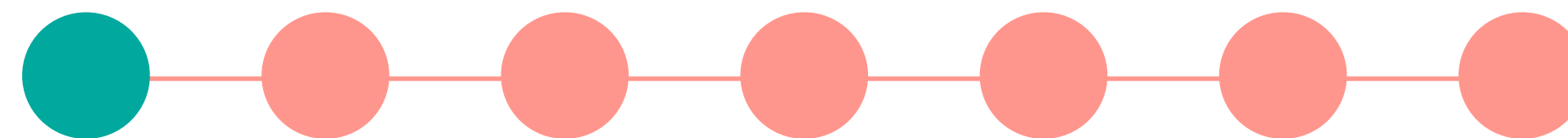


# Epidemics

Which outbreak is more dangerous to the population?



vs.

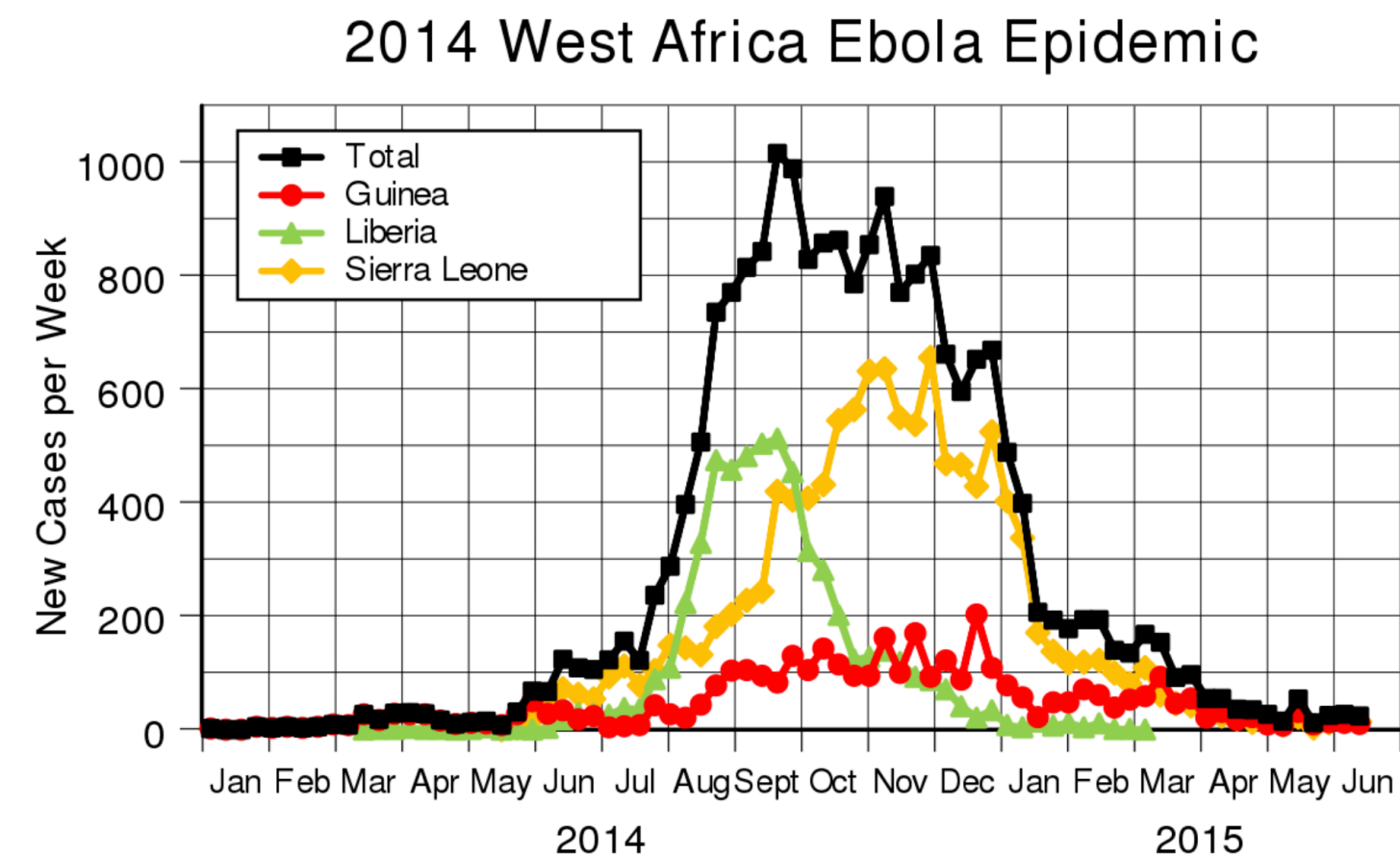




# Epidemics

Types of epidemic diffusion:

- **Explosive spread** through a population
- **"Slow burn"** persistence over long periods of time
- Wave-like **cyclical patterns**





# Epidemics

**Explosive spread:** Bubonic Plague (the “Black Death”): wiped out ~50% of the population in Europe (~150 million people) in 7 years



1346 1347 1348 1349 1350 1351 1352 1353

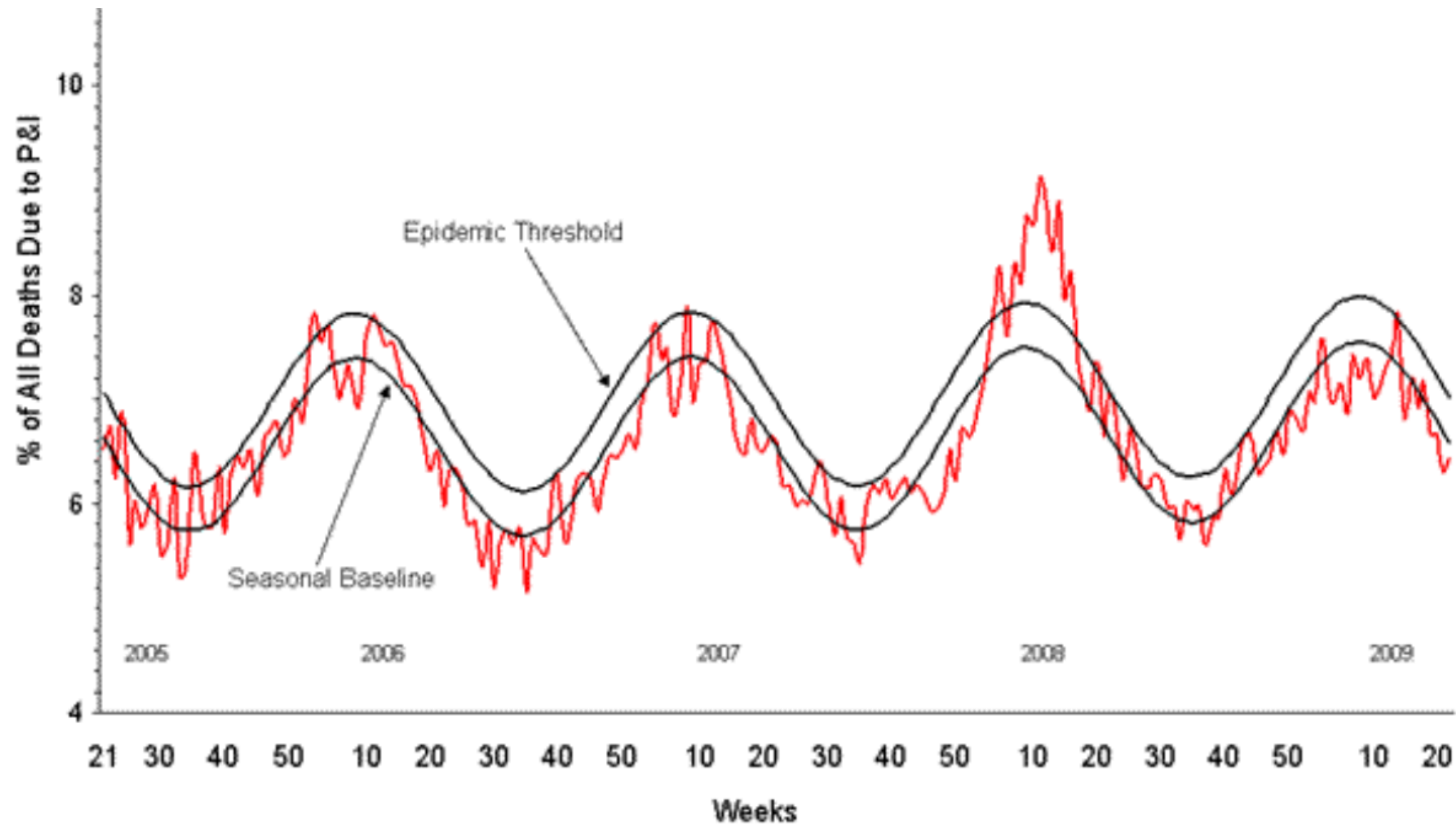
--- Approximate border between the Principality of Kiev and the Golden Horde - passage prohibited for Christians.

Land trade routes  
Maritime trade routes



# Epidemics

Other epidemics are **cyclical**



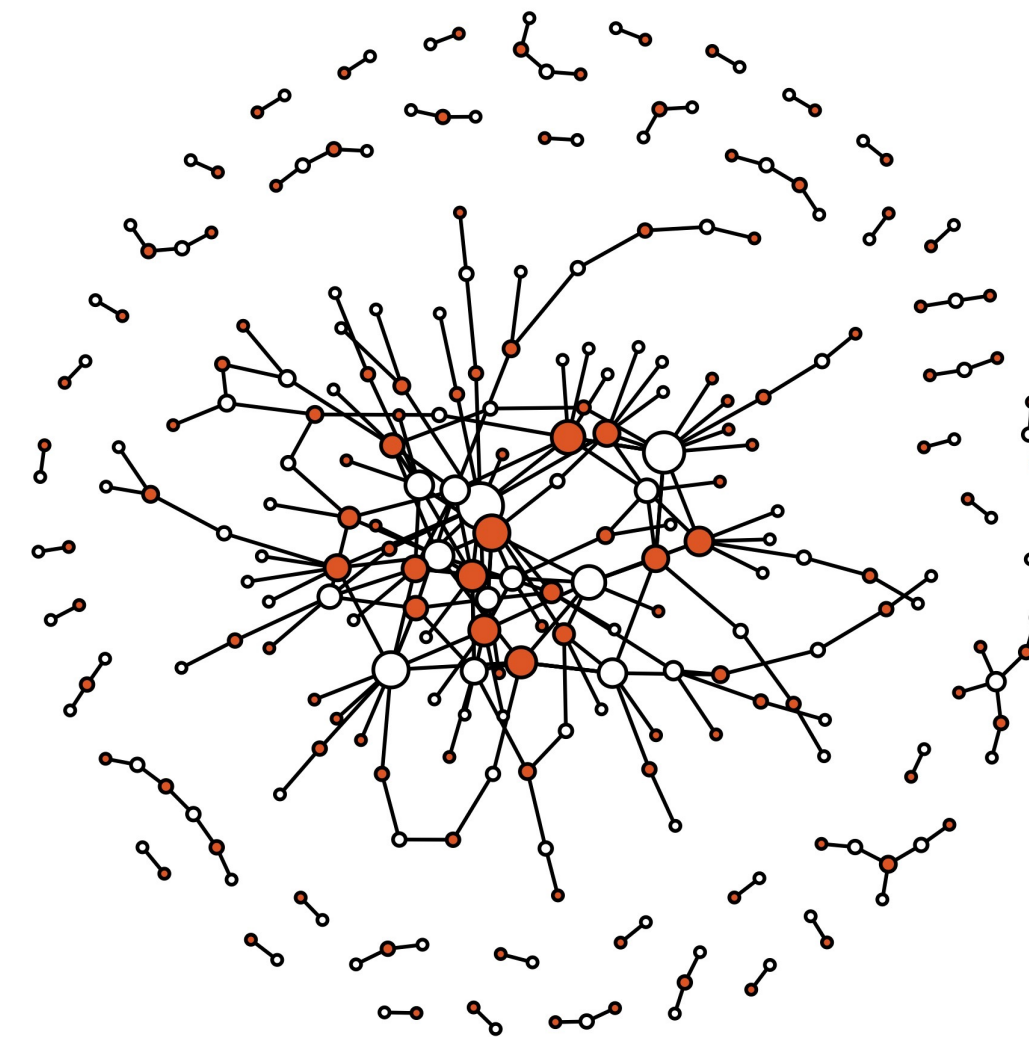


# Epidemics

What determines how an epidemic might spread?

- Properties of the disease
- Structure of the network

What network?

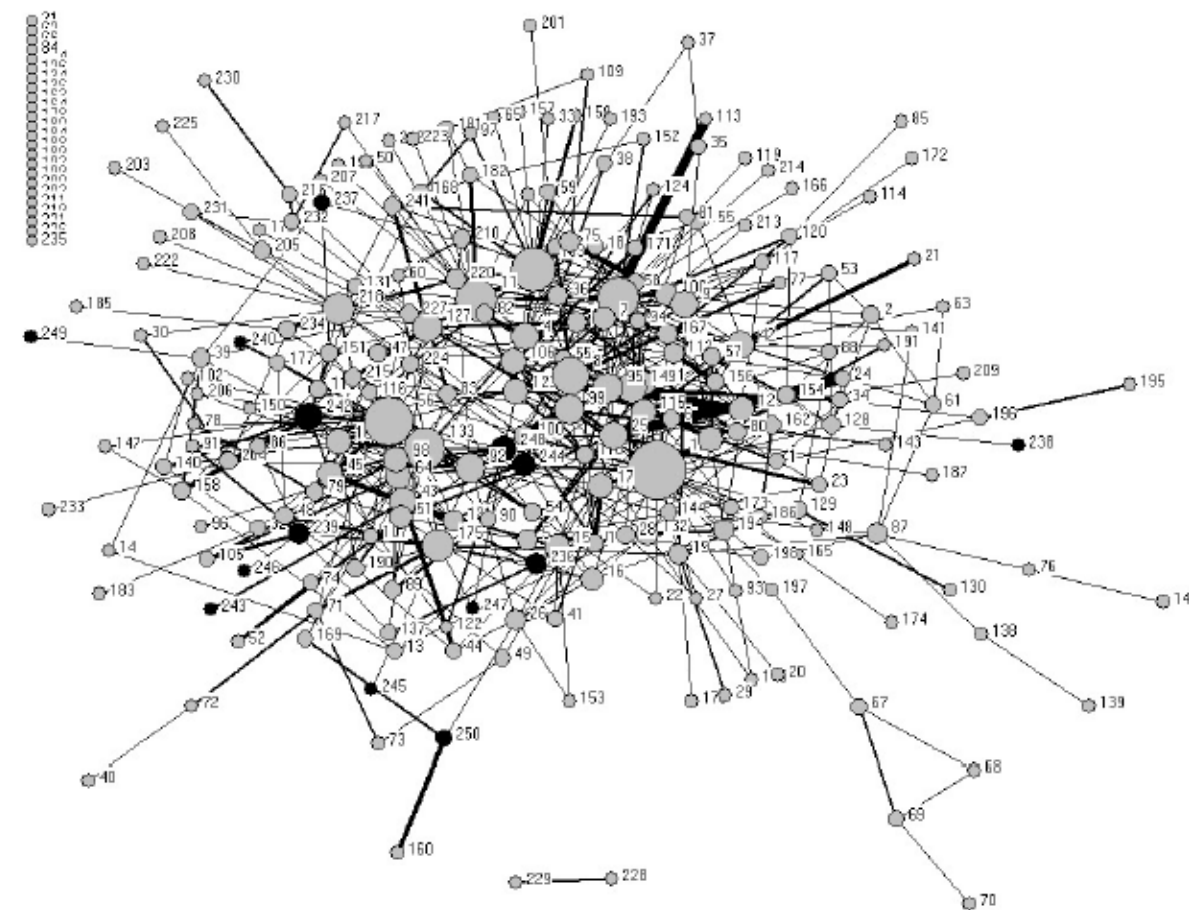




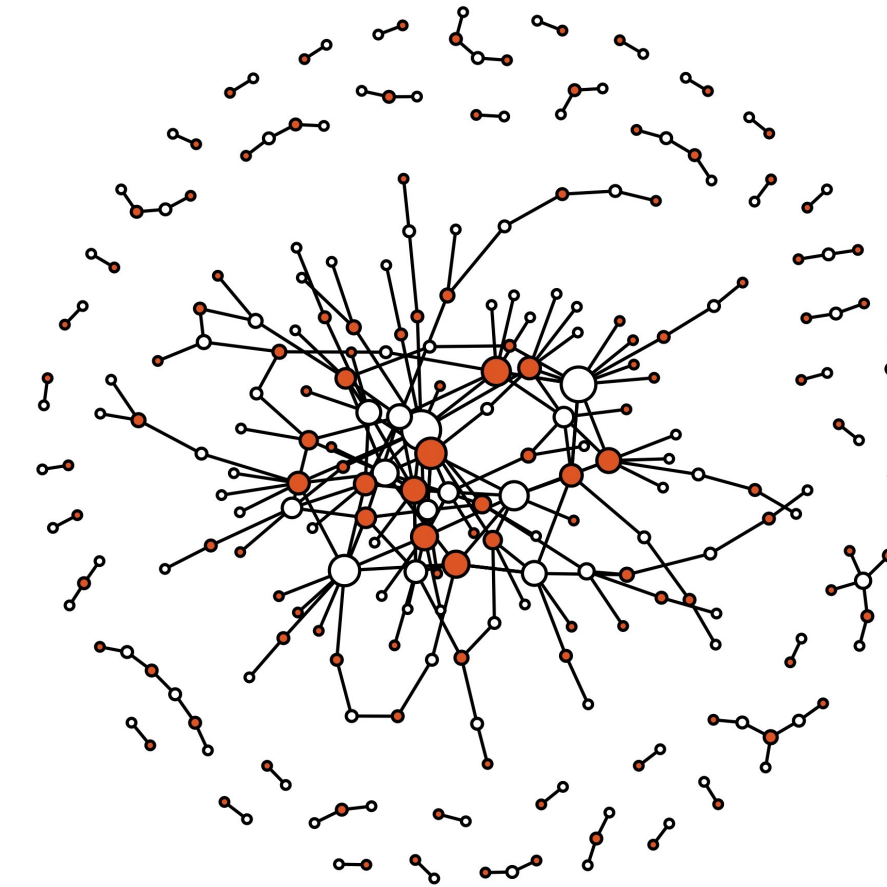
# Contact Networks

**Node** for each person

**Edge** if two people come into contact with each other in a way that makes it possible for a disease to spread



UK fish farm exchanges



CH sexual contact network

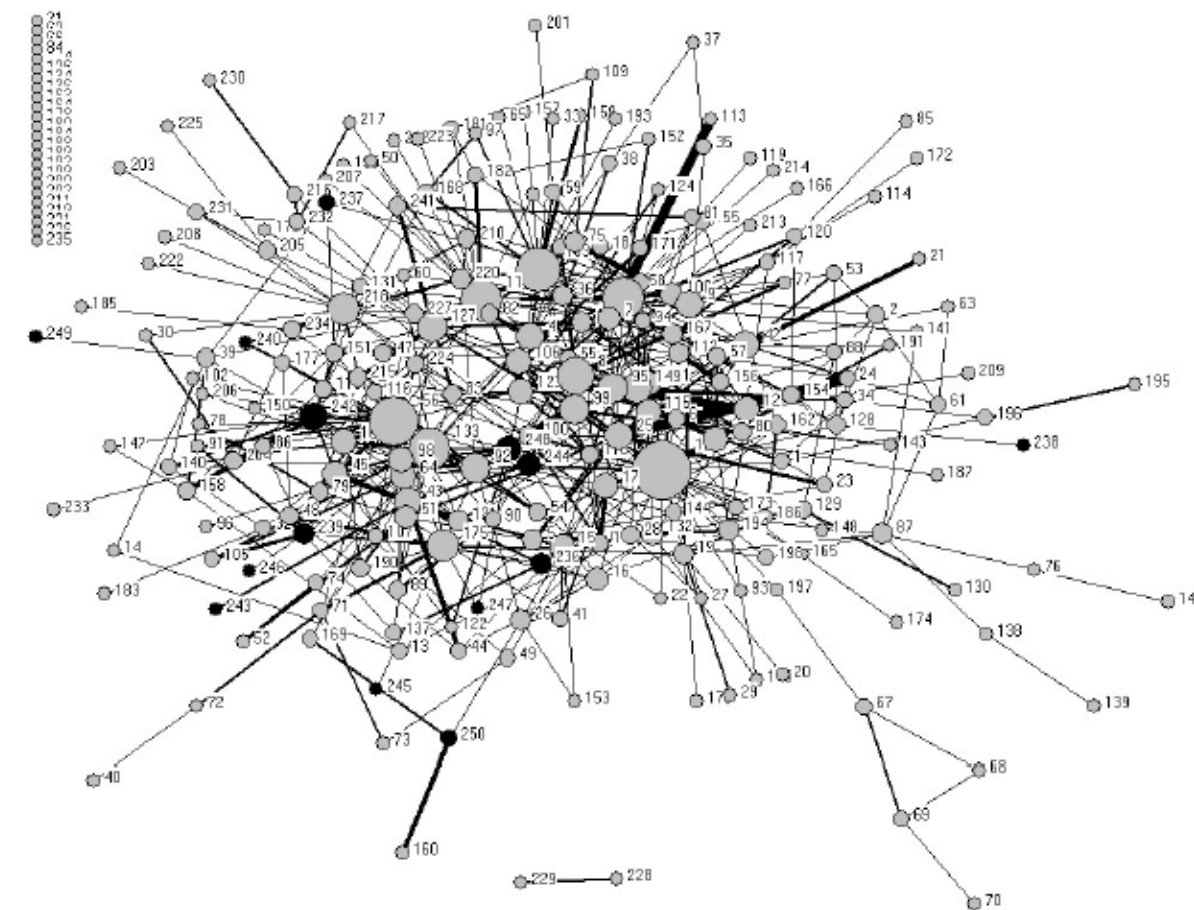


# Contact Networks

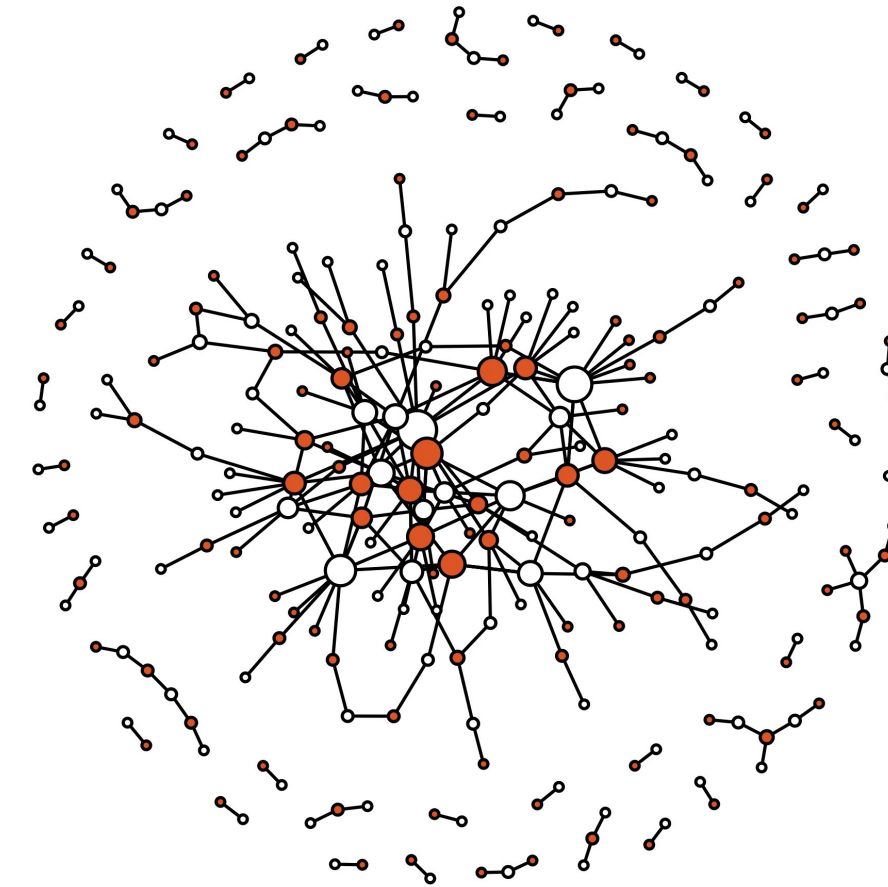
Once you've got through the laborious process of mapping out the contact network, **can you use it to study any disease?**

**No!** Definition of “contact” depends on the disease

- Airborne transmission: edge between everyone who was in the same car, etc.)  
— many edges
- Close contact / sexual transmission: sparser graph



UK fish farm exchanges



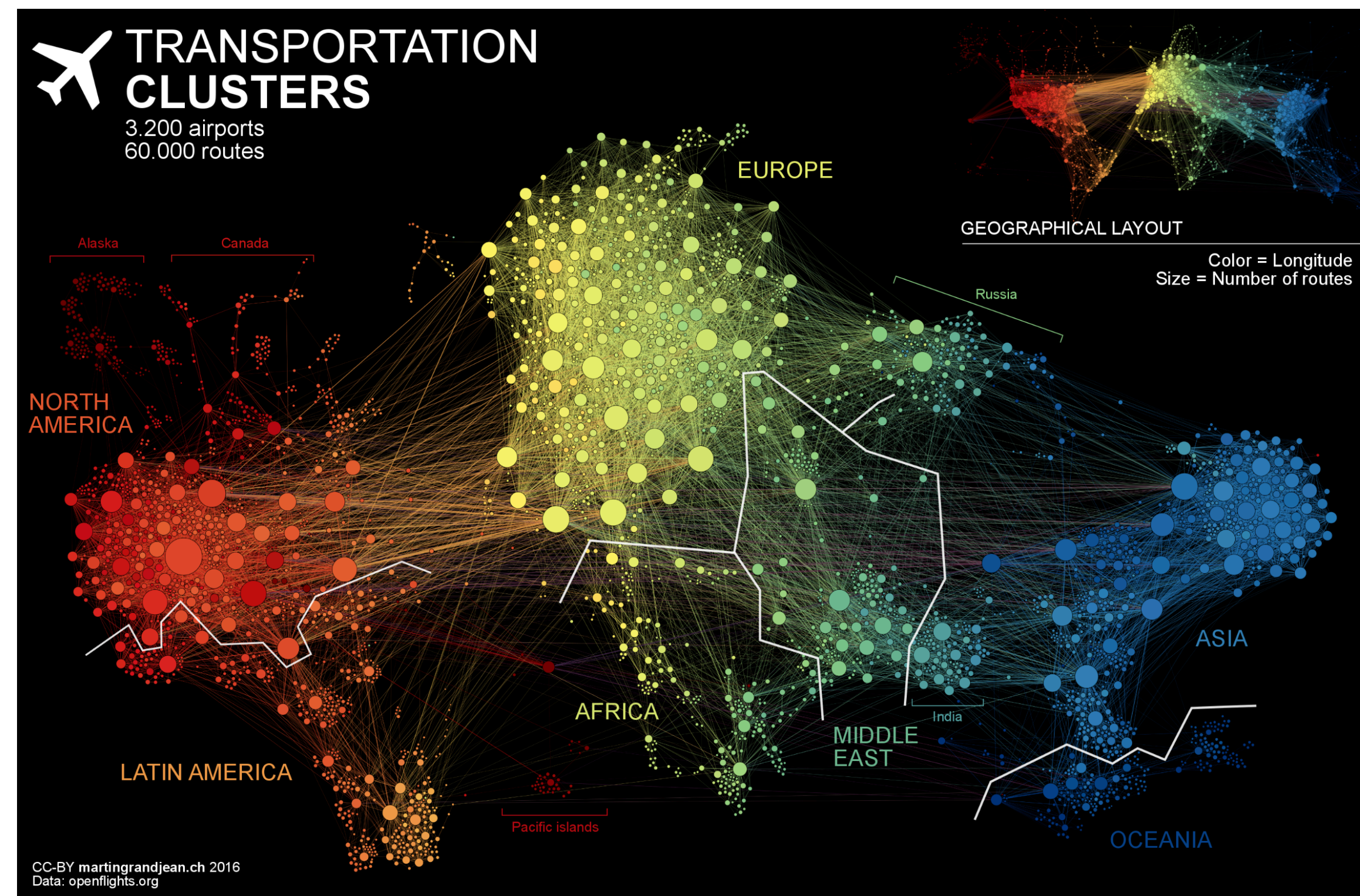
CH sexual contact network



# Contact Networks

Big part of real-world epidemic research is constructing contact networks

Lots of work on travel patterns in cities, the worldwide airline network, etc. to understand how diseases can spread in today's world

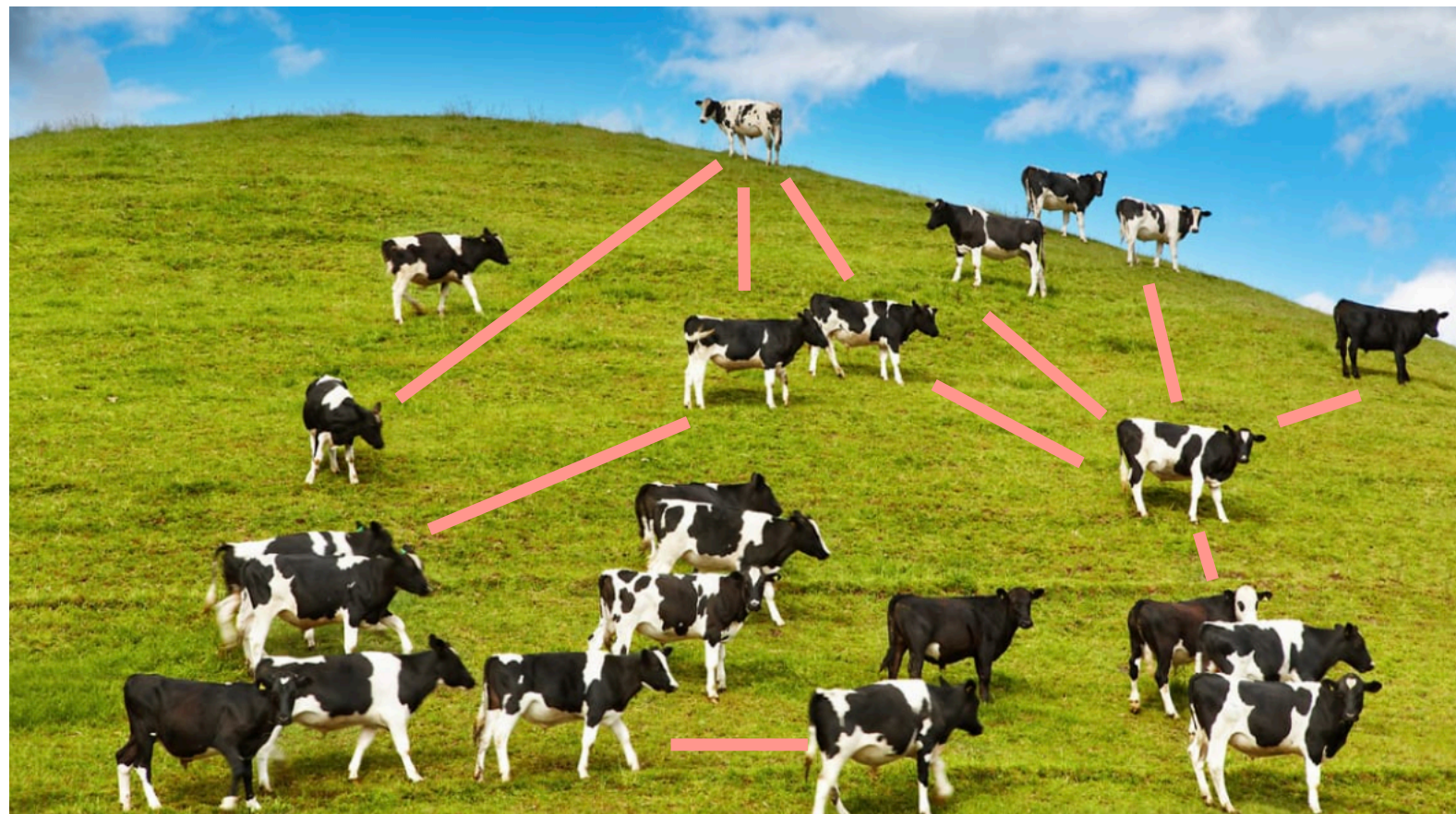




# Contact Networks

Not just human contact networks

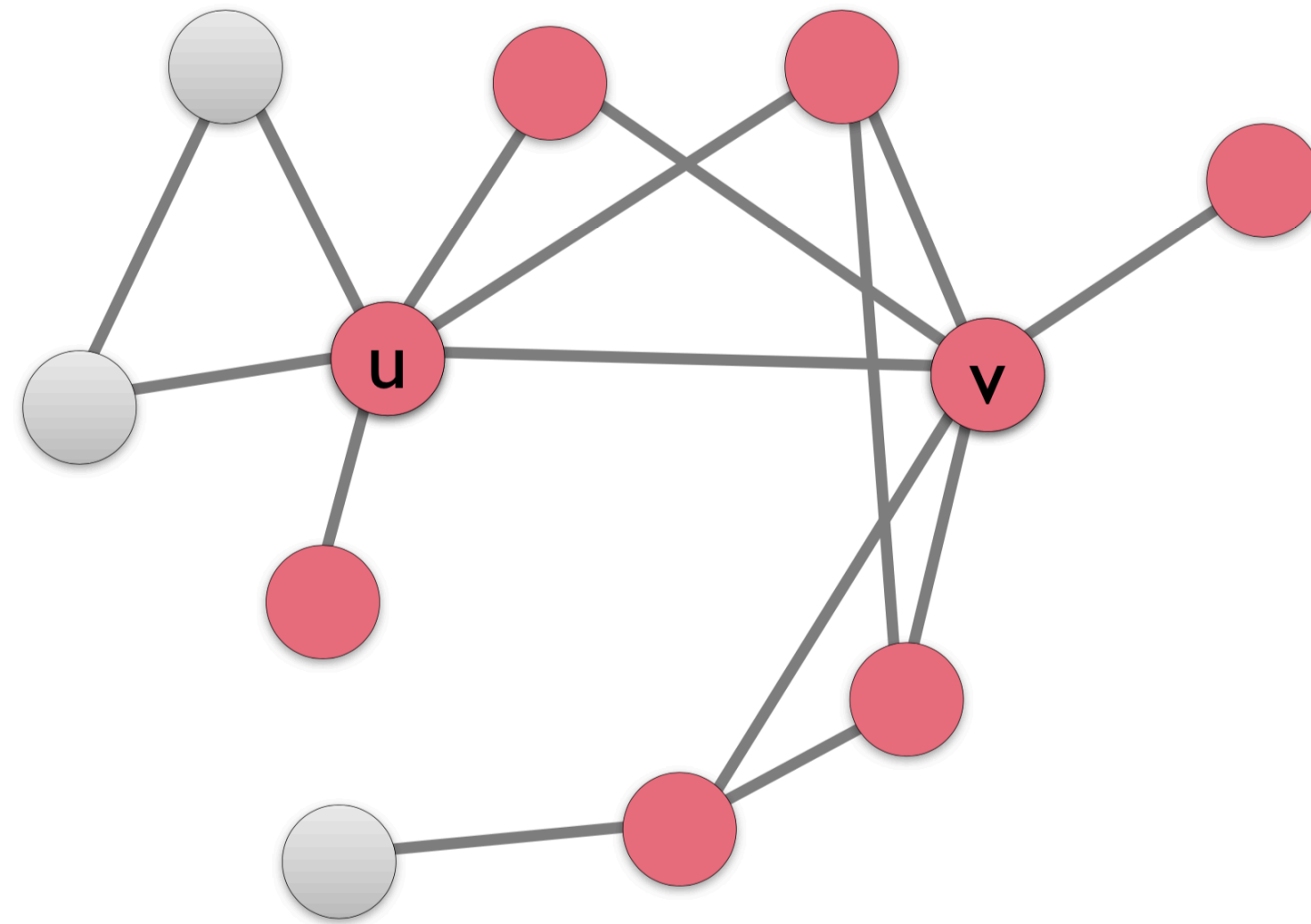
Animal/livestock networks and plant networks



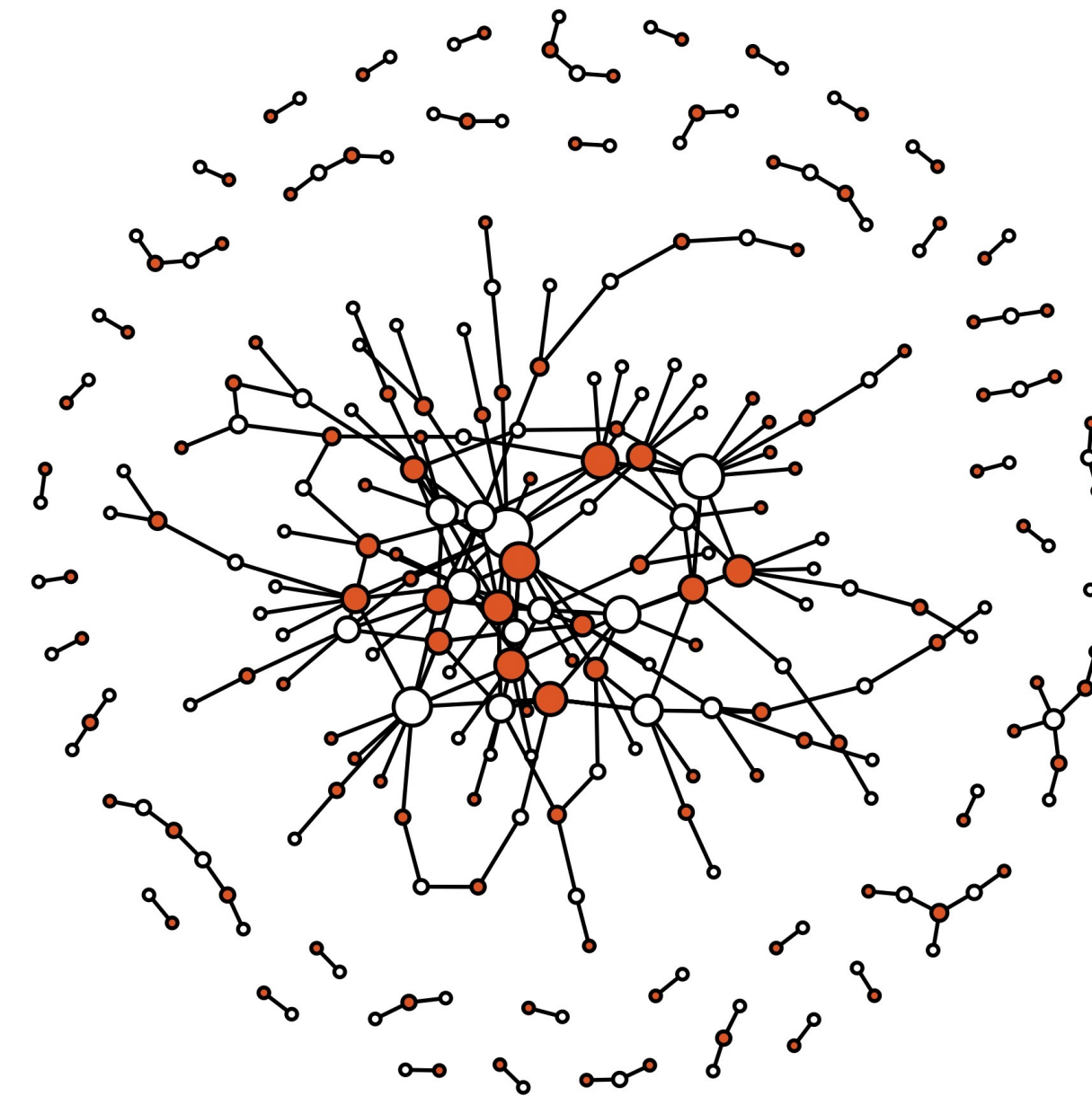


# Behavioural vs. Biological Contagion

Biological/epidemic diffusion: **no decision-making!**



Decision cascade



High school contact network



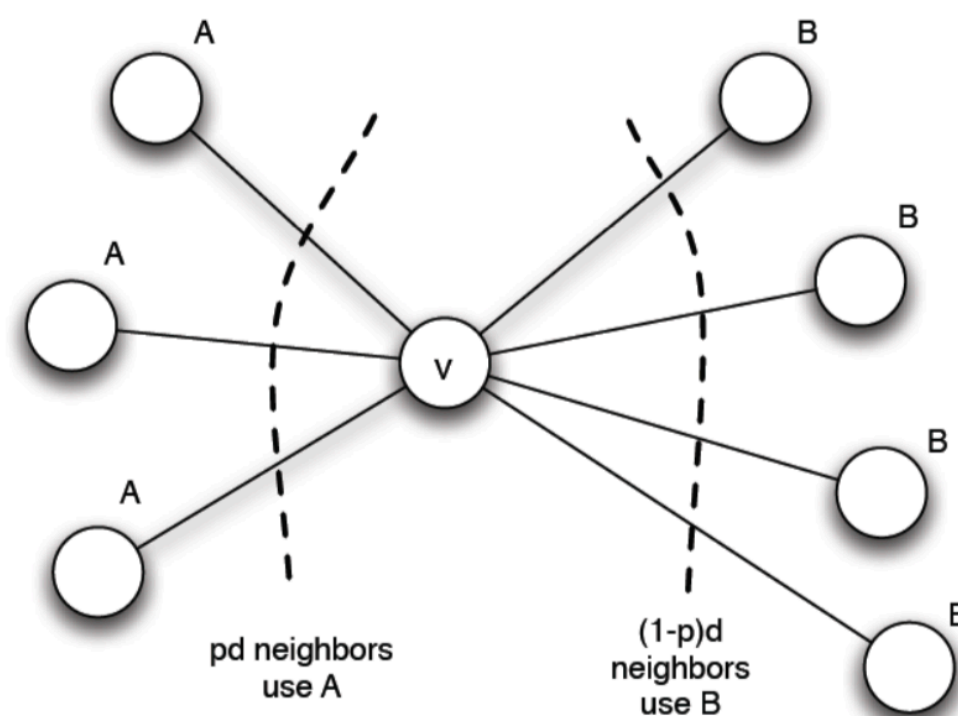
# Modeling Epidemic Diffusion

Biggest difference: model transmission as **random**

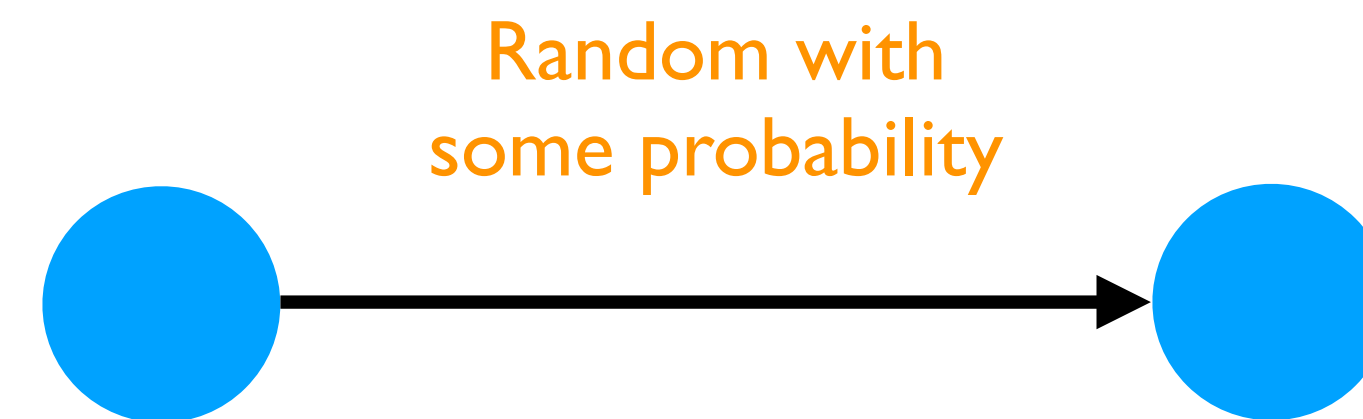
No **decision-making**, but also the processes by which diseases spread from one person to another are **so complex and unobservable at the individual level** that it's **most useful to think of them as random**

Use randomness to **abstract away** difficult biological questions about the mechanics of spread

**Behaviour (last class):**



**Epidemics (today):**





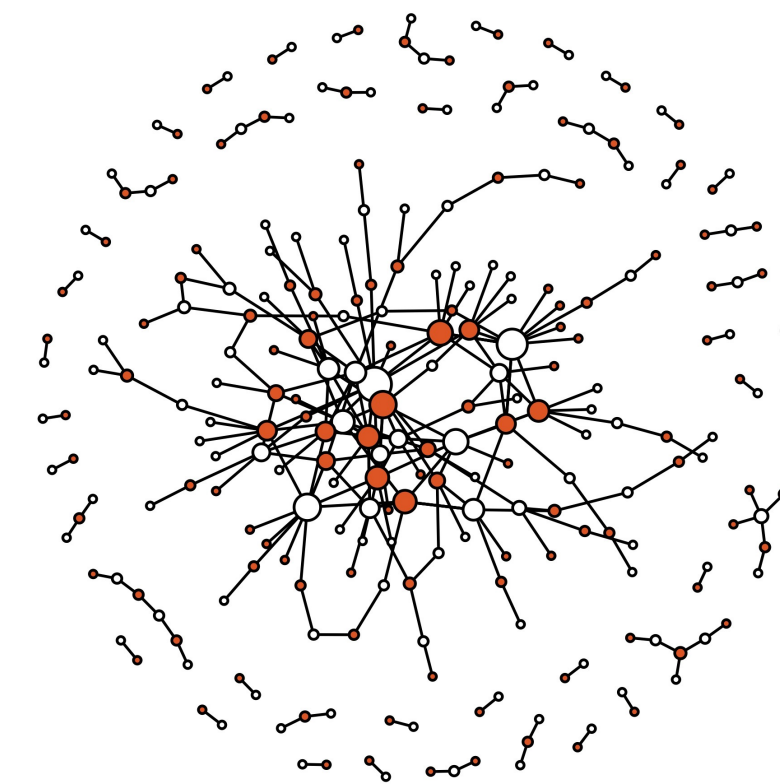
# Modeling Epidemic Diffusion



# Branching Process

## Basic structure of epidemic diffusion:

- Someone gets infected
- Then they infect some number of people
- Those people infect others





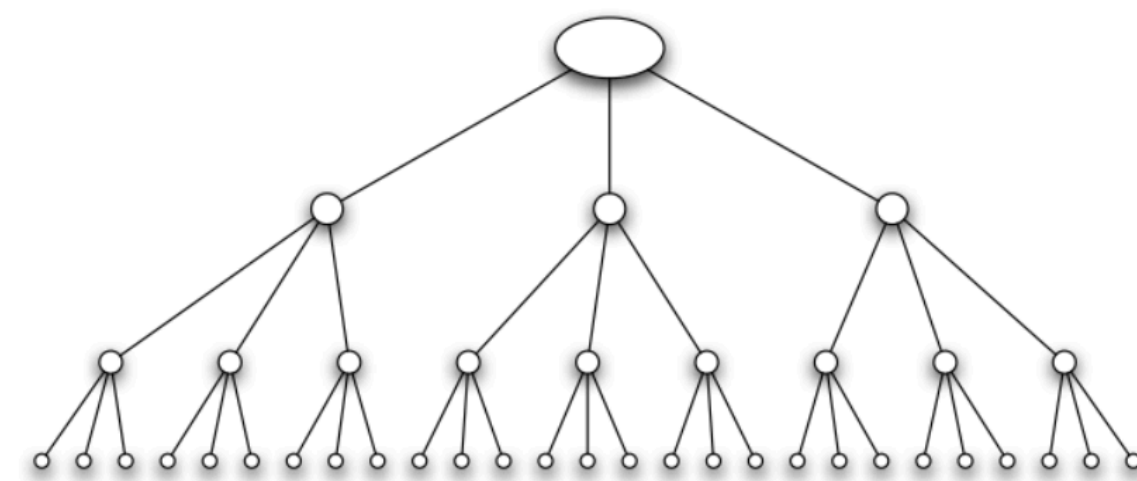
# Branching Process

Model as a **random process on a tree**:

**Wave 1:** First person infected, infects each of  $k$  neighbors with independent probability  $p$

**Wave 2:** For each infected person, they infect each of  $k$  neighbors with independent probability  $p$

**Wave 3+:** repeat for each infected person



**Here  $k=3$**

Extends infinitely below





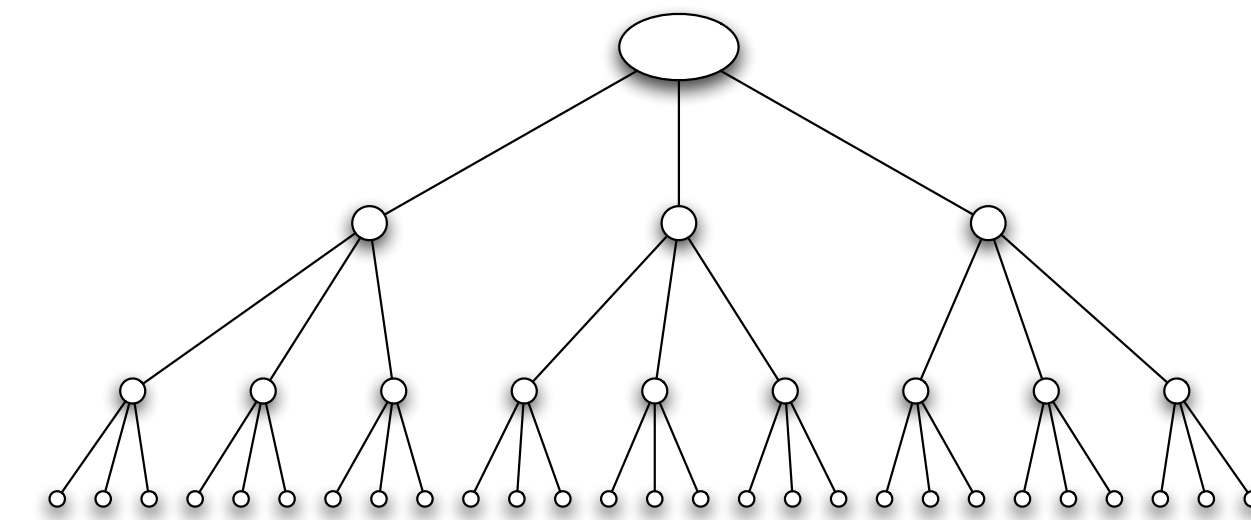
# Branching Process

Model parameters:

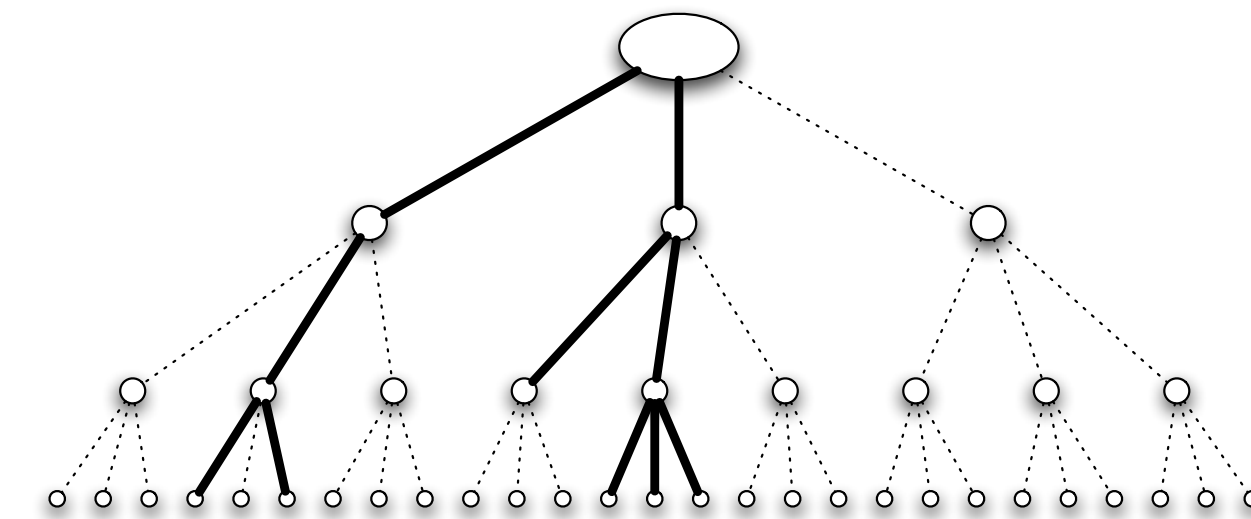
$k$ : number of individuals each person can possibly infect:

Higher transmission probability  $p$ :

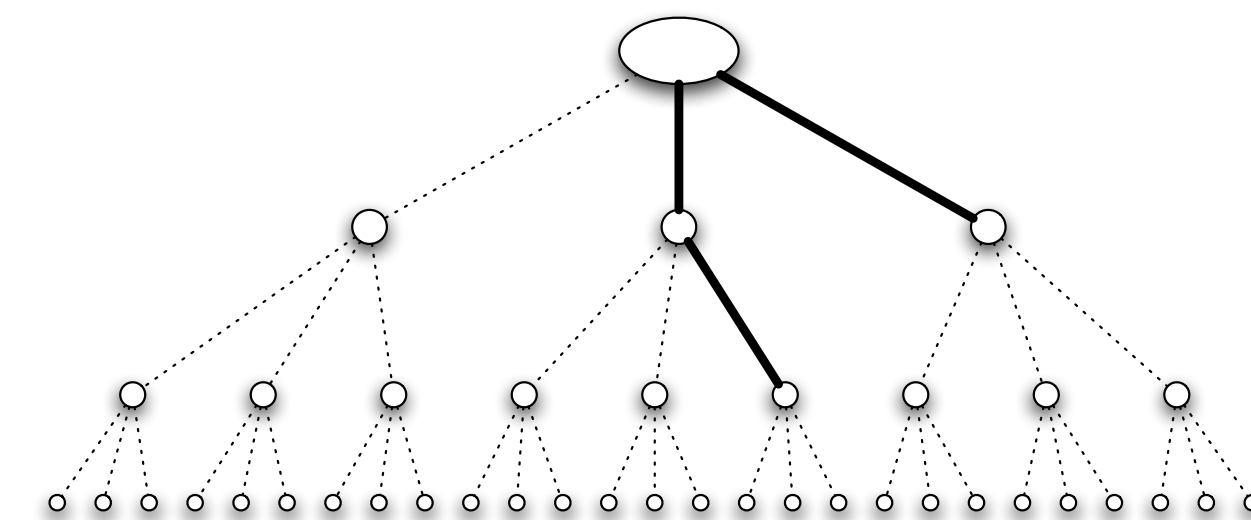
Lower transmission probability  $p$ :



(a) The contact network for a branching process



(b) With high contagion probability, the infection spreads widely



(c) With low contagion probability, the infection is likely to die out quickly



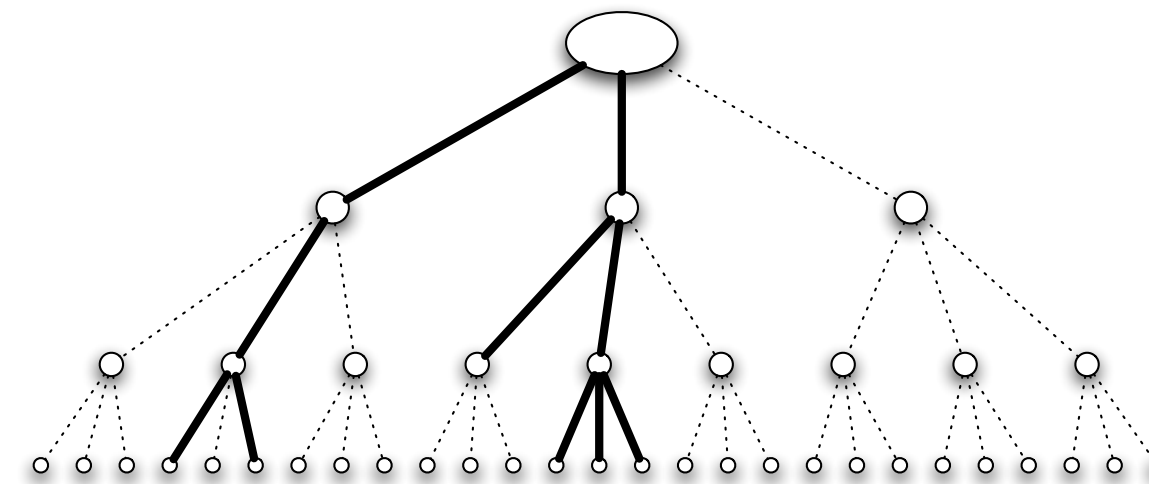
# Branching Process: Outcomes

Only two possibilities in the long run: **blow up** or **die out**

How does it die out?

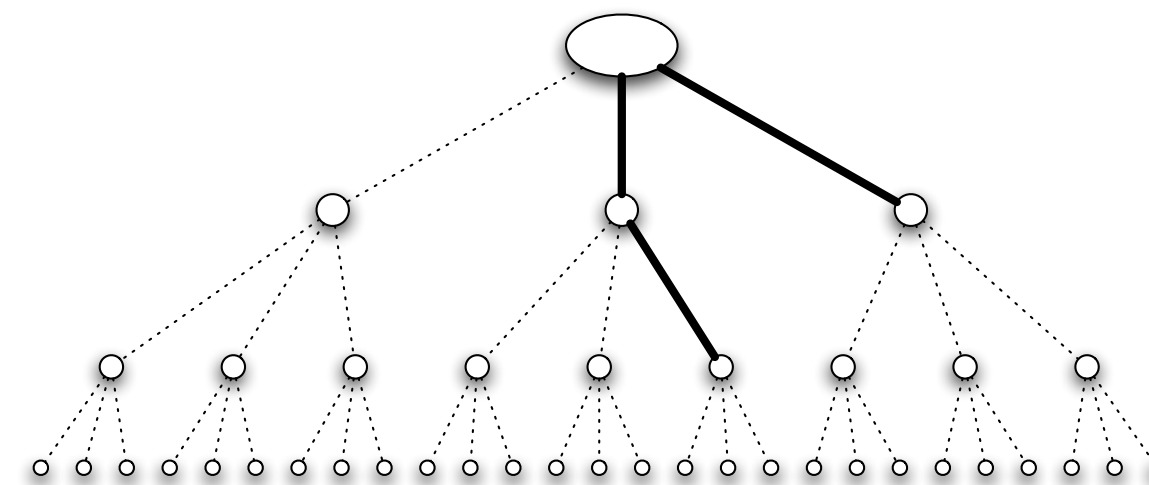
- Dies out if and only if none of the nodes on a given level are infected

**Disease might blow up:**



(b) *With high contagion probability, the infection spreads widely*

**Disease has already died out:**



(c) *With low contagion probability, the infection is likely to die out quickly*

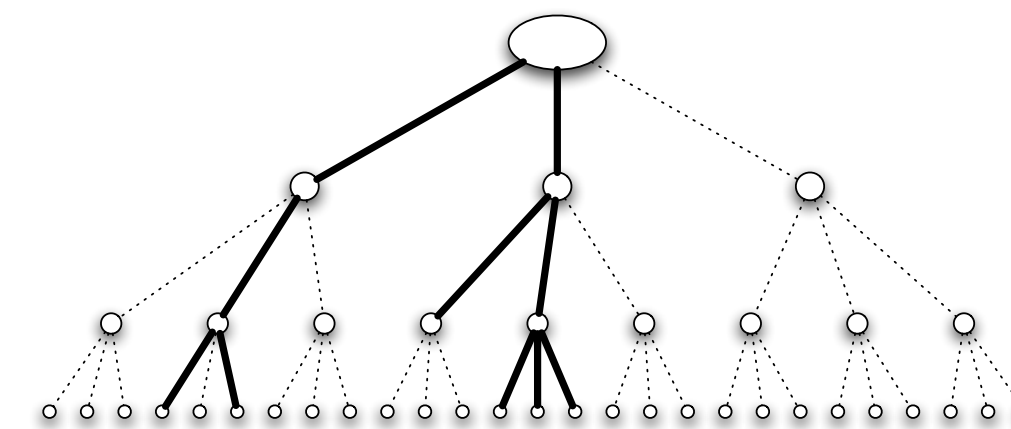
# Branching Process

Only two possibilities in the long run: **blow up** or **die out**

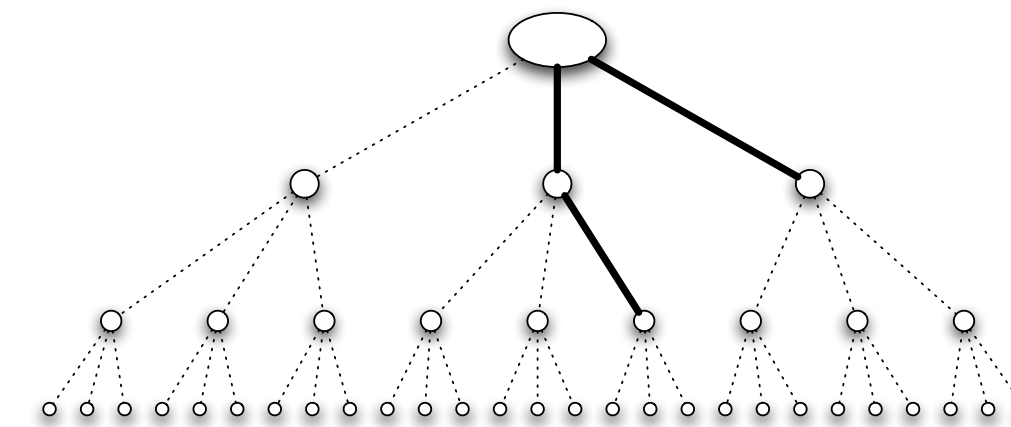
How does it die out?

- Dies out if and only if none of the nodes on a given level are infected

Define **Basic reproductive number  $R_0$** : the number of expected new cases caused by an individual



(b) With high contagion probability, the infection spreads widely



(c) With low contagion probability, the infection is likely to die out quickly



# Branching Process

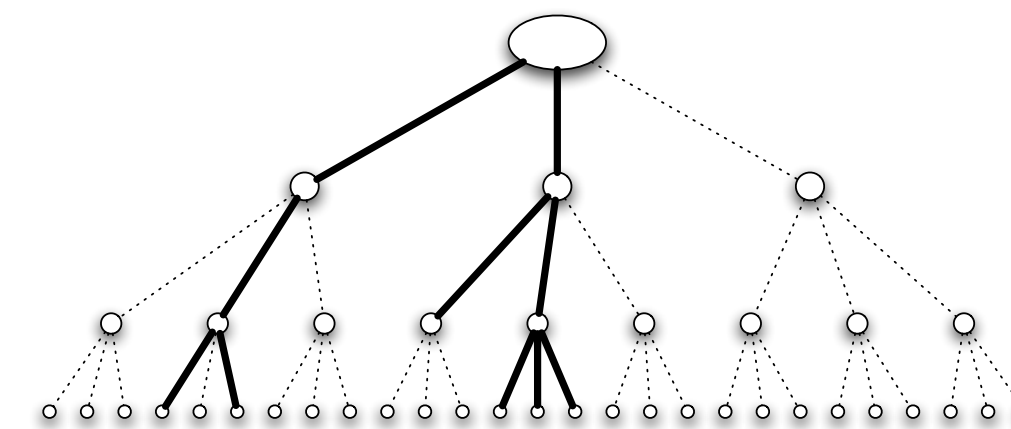
Only two possibilities in the long run: **blow up** or **die out**

How does it die out?

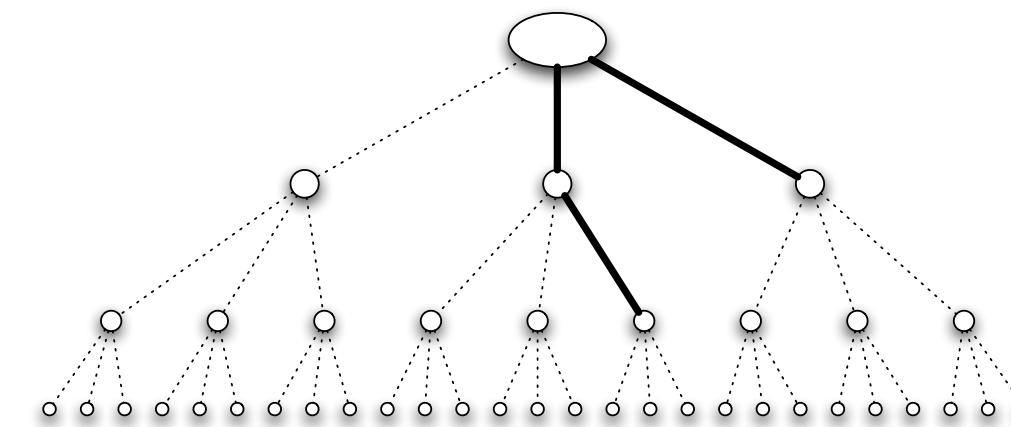
- Dies out if and only if none of the nodes on a given level are infected

Define **Basic reproductive number  $R_0$** : the number of expected new cases caused by an individual

$$R_0 = pk$$



(b) With high contagion probability, the infection spreads widely



(c) With low contagion probability, the infection is likely to die out quickly

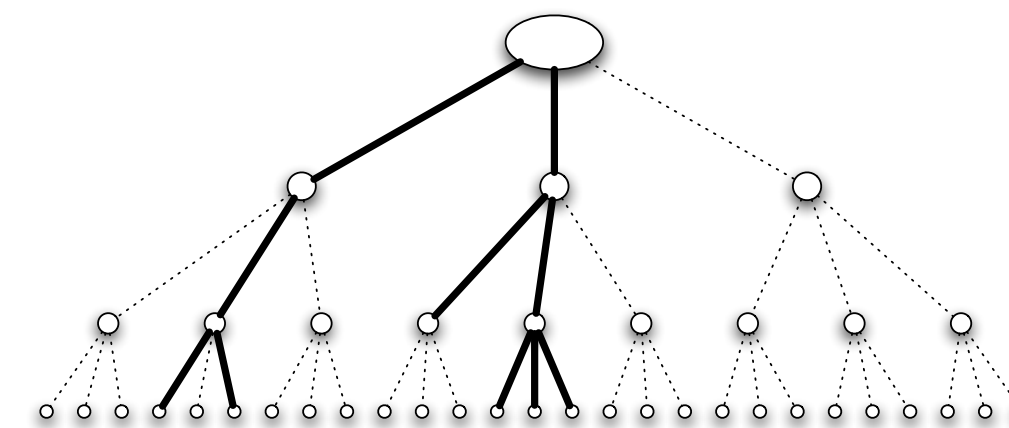
# Branching Process: $R_0$

Claim: Epidemic spread in the branching process model is **entirely controlled by the reproductive number  $R_0$**  :

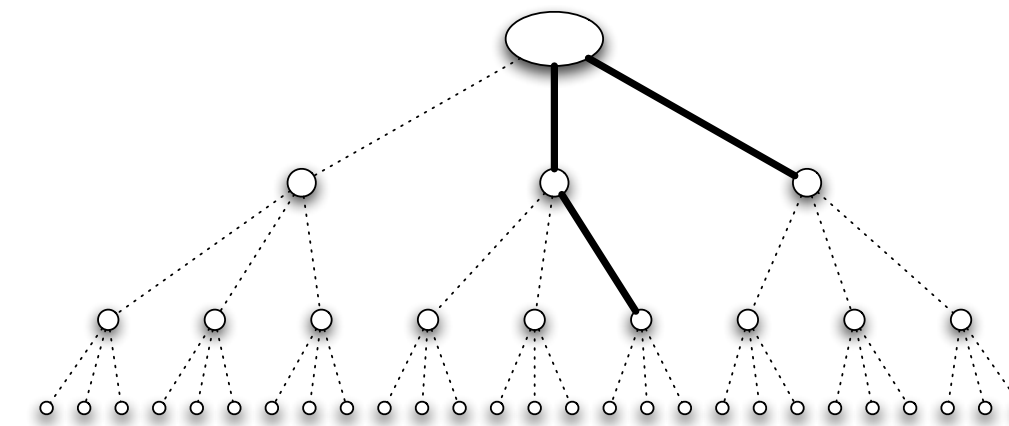
- If  $R_0 < 1$  then with probability 1 the disease dies out after a finite number of steps.
- If  $R_0 > 1$  then with probability  $> 0$  the disease persists by infecting at least one person in each wave.

“Go big or go home.”

$$R_0 = pk$$



(b) With high contagion probability, the infection spreads widely



(c) With low contagion probability, the infection is likely to die out quickly



# Branching Process: $R_0$

**$R_0 = pk < 1$ :**

With probability 1 the disease dies out after a finite number of steps

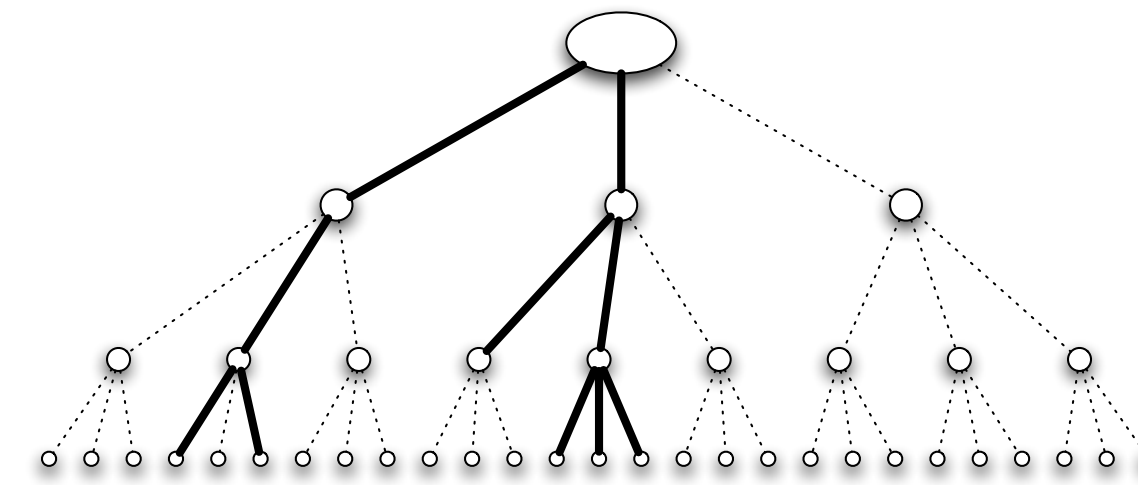
Below replacement; disease isn't able to replenish itself.

Even if it grows momentarily, it trends downward.

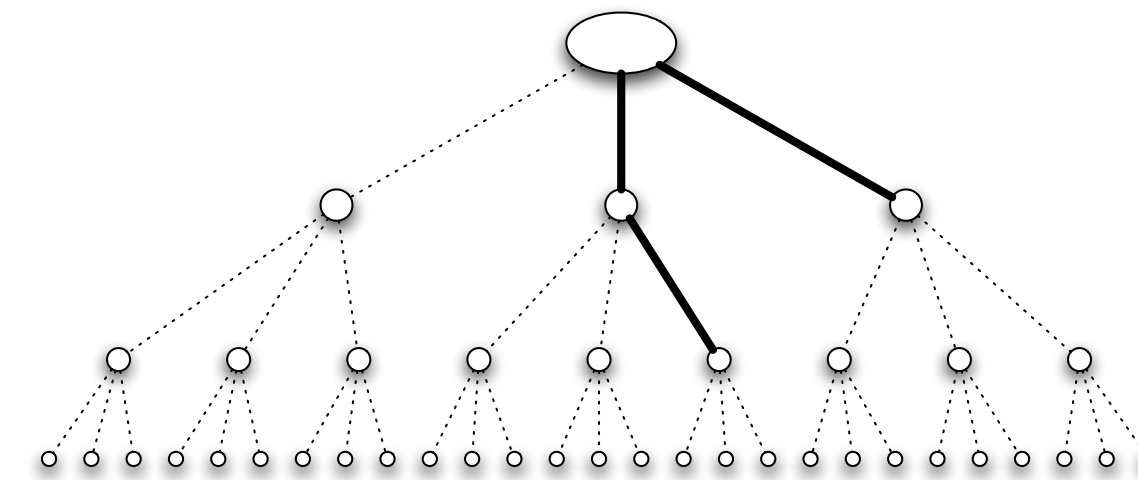
**$R_0 = pk > 1$ :**

with probability  $> 0$  the disease persists by infecting at least one person in each wave

Always trending upward. Could still get “unlucky” and die out, but there's a non-zero chance it runs forever.



(b) With high contagion probability, the infection spreads widely



(c) With low contagion probability, the infection is likely to die out quickly

# Branching Process: $R_0$

**$R_0 = pk < 1$ :**

With probability 1 the disease dies out after a finite number of steps

Below replacement; disease isn't able to replenish itself.

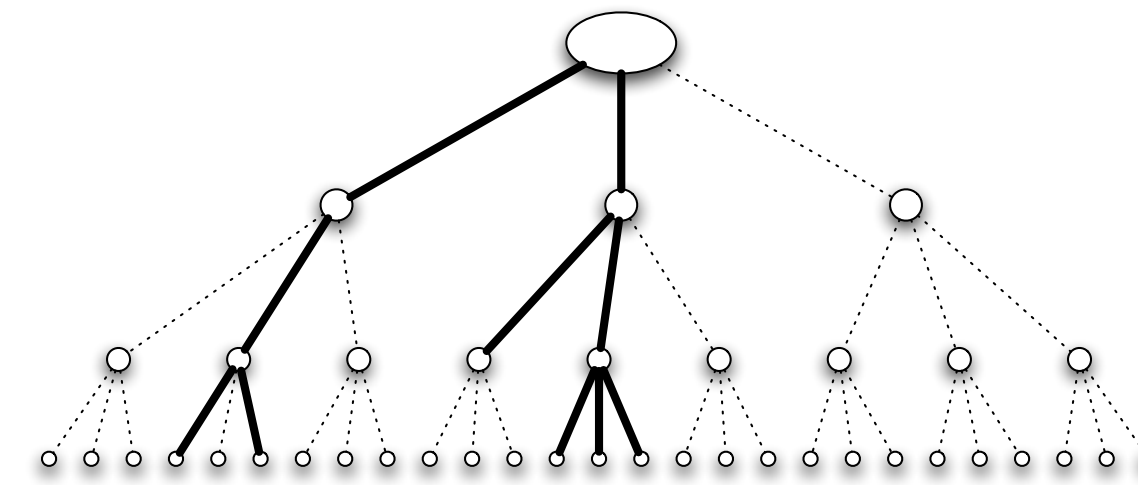
Even if it grows momentarily, it trends downward.

**$R_0 = pk > 1$ :**

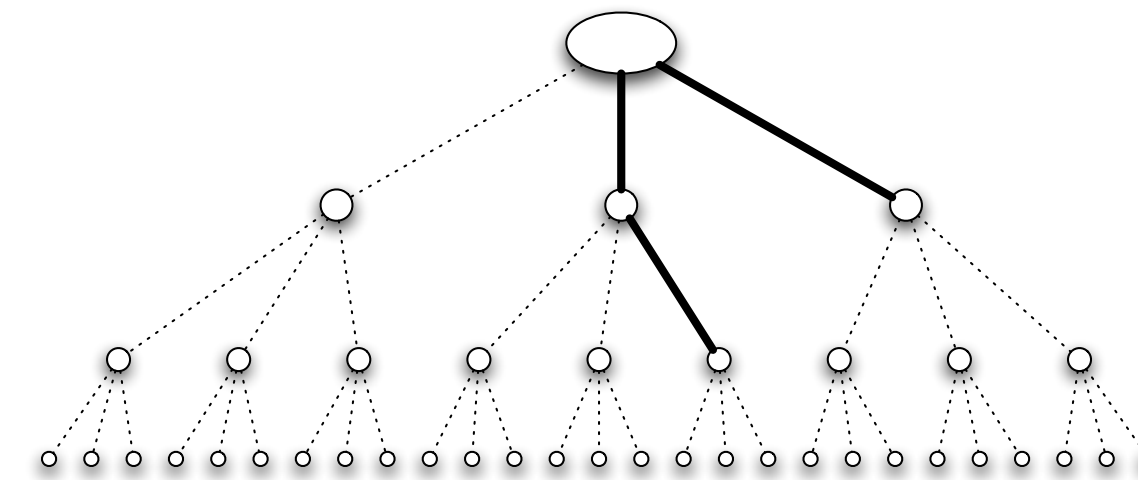
with probability  $> 0$  the disease persists by infecting at least one person in each wave

Always trending upward. Could still get “unlucky” and die out, but there's a non-zero chance it runs forever.

What happens when  $p$  or  $k$  change near  $pk=1$ ?



(b) With high contagion probability, the infection spreads widely



(c) With low contagion probability, the infection is likely to die out quickly

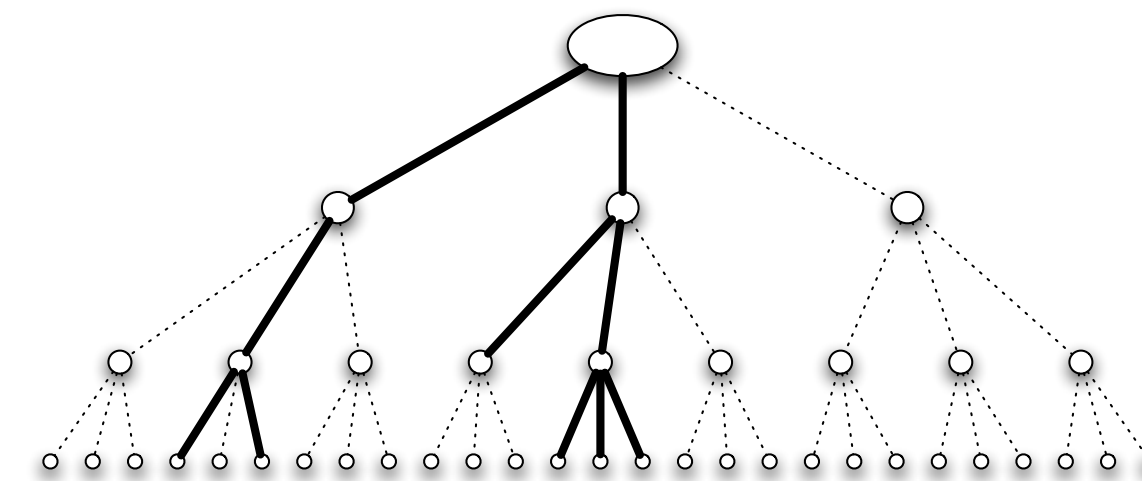


# Sensitivity of $p$ and $k$

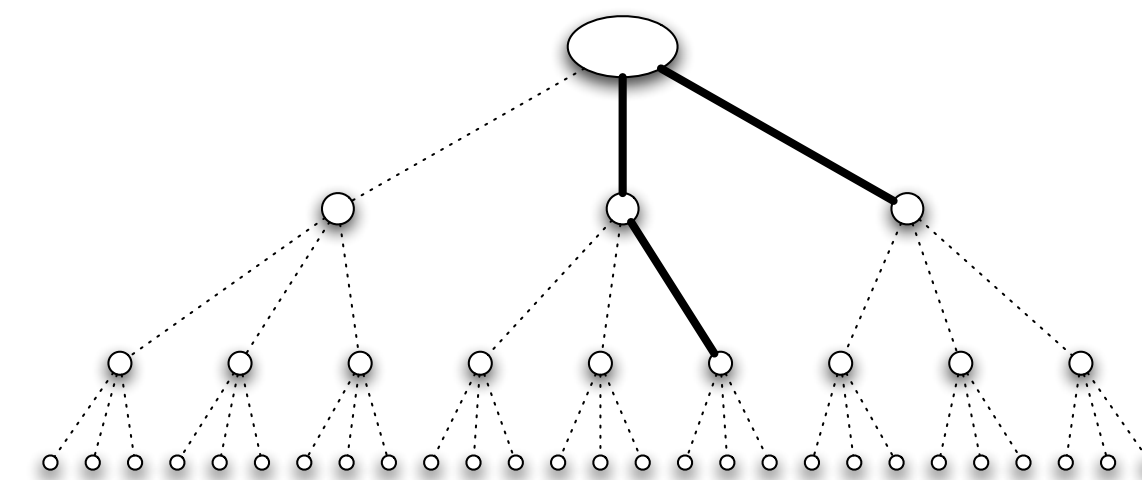
Because epidemics have a “critical threshold”, it can be worth it to **do a lot of work or expend resources** to push  **$p$**  or  **$k$  down** a little bit.

Quarantine = reduce  **$k$**

Improved sanitation = reduce  **$p$**



(b) With high contagion probability, the infection spreads widely



(c) With low contagion probability, the infection is likely to die out quickly

Disease	Transmission	R <sub>0</sub>
Measles	Airborne	12–18
Diphtheria	Saliva	6–7
Smallpox	Airborne droplet	5–7
Polio	Fecal-oral route	5–7
Rubella	Airborne droplet	5–7
Mumps	Airborne droplet	4–7
HIV/AIDS	Sexual contact	2–5
Pertussis	Airborne droplet	5.5 <sup>[2]</sup>
SARS	Airborne droplet	2–5 <sup>[3]</sup>
Influenza (1918 pandemic strain)	Airborne droplet	2–3 <sup>[4]</sup>
Ebola (2014 Ebola outbreak)	Bodily fluids	1.5–2.5 <sup>[5]</sup>

**COVID-19: ~2**  
**COVID Delta: ~5**  
**COVID Omicron: ~8**



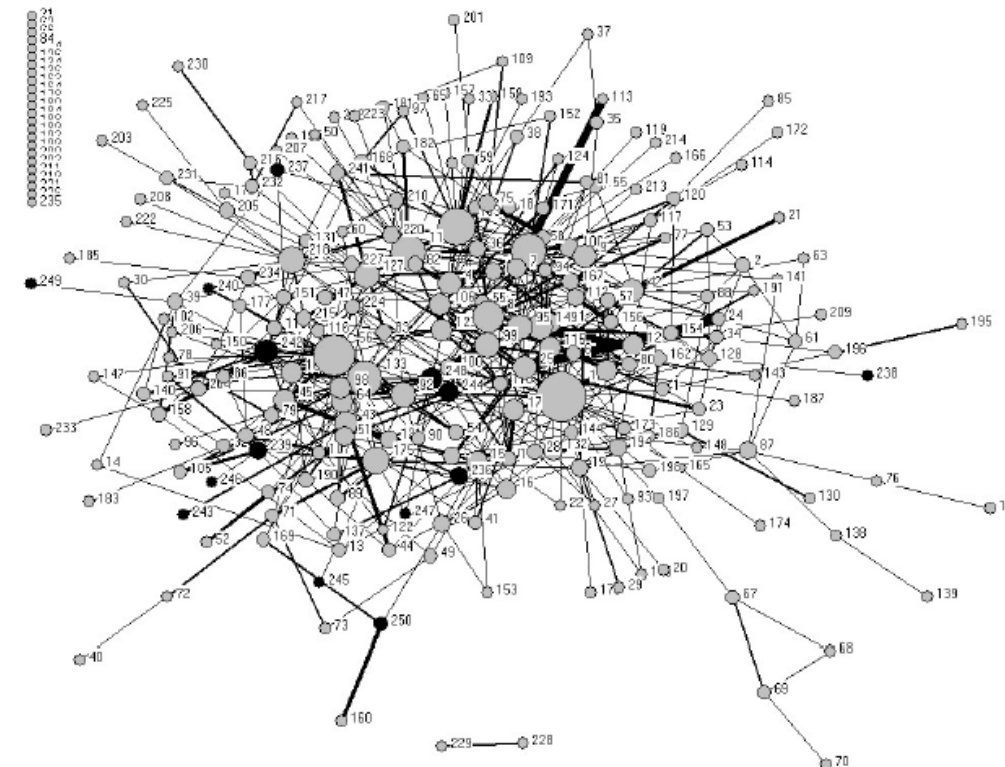
# General Models of Contagion

# Epidemics on General Graphs

We just studied epidemics as **ideal trees**

But of course real-life networks are **more complicated** than that

What does epidemic diffusion look like in general graphs?

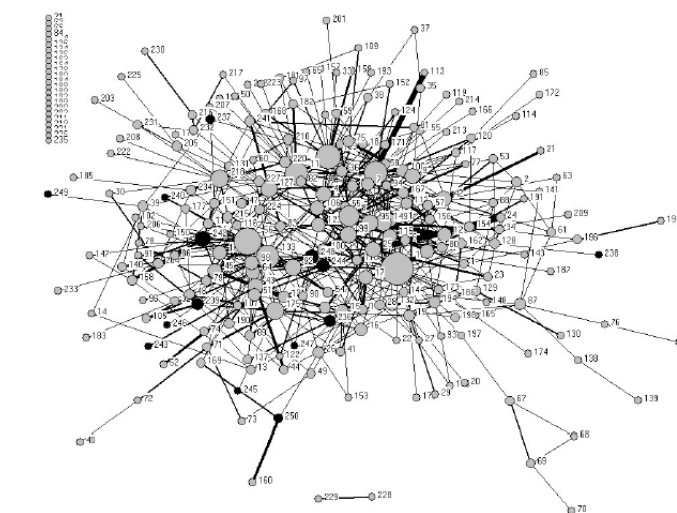
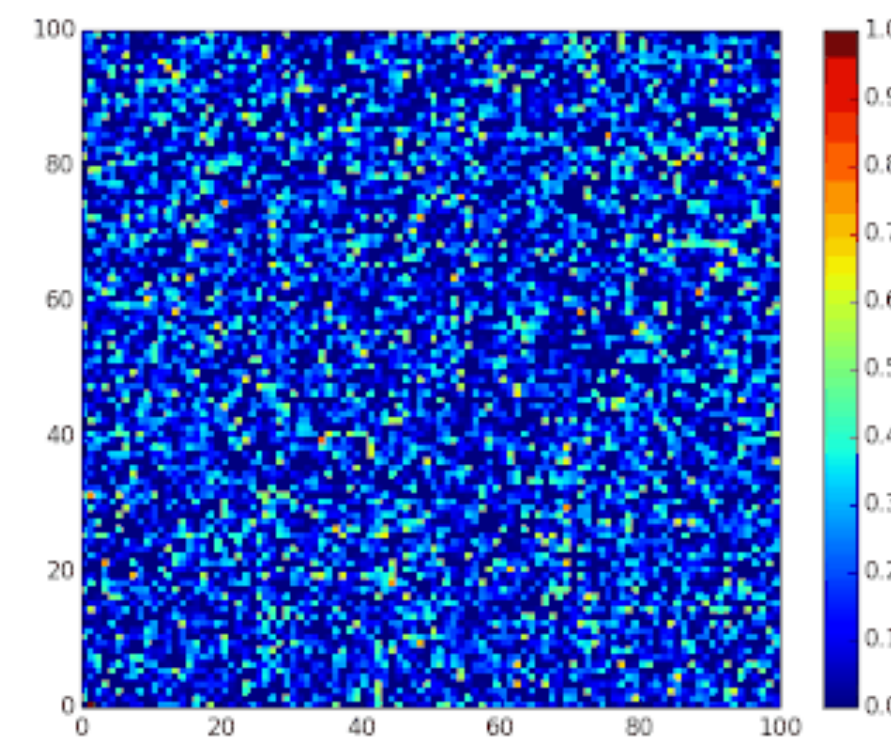
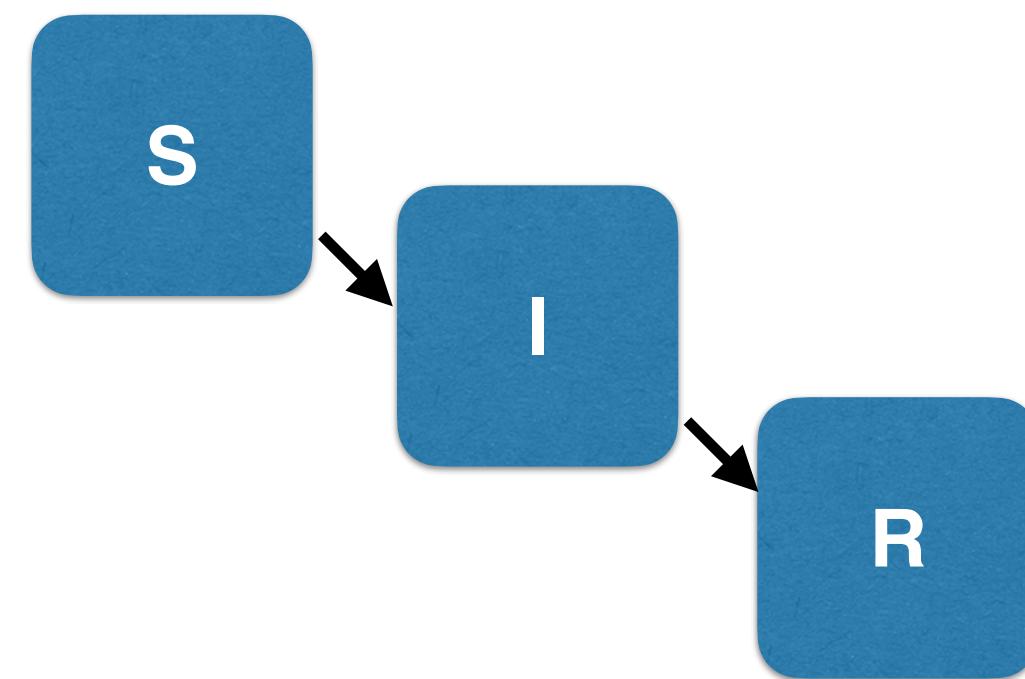




# SIR Epidemic Models

Simple lifecycle model with three stages:

- **S** = Susceptible
- **I** = Infectious: node is infected and infects with prob **p**
- **R** = Removed: after  **$t_i$**  time, no longer infected or infectious



# SIR Epidemic Models

**S** = Susceptible

**I** = Infectious: node is infected and infects with prob **p**

**R** = Removed: after **t<sub>i</sub>** time, no longer infected or infectious

Initially some nodes in **I** state, rest in **S** state.

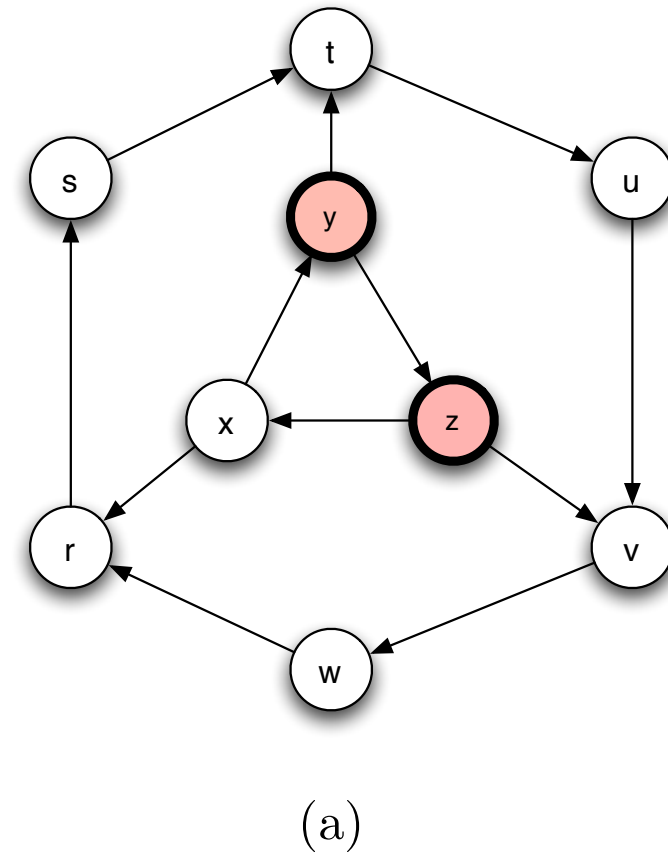
Each node in **I** state remains infected for **t<sub>i</sub>** time steps

During each step, each node has probability **p** of infecting each susceptible neighbour

After **t<sub>i</sub>** time steps, no longer **S** nor **I**; removed to **R**

# SIR Epidemics on Networks

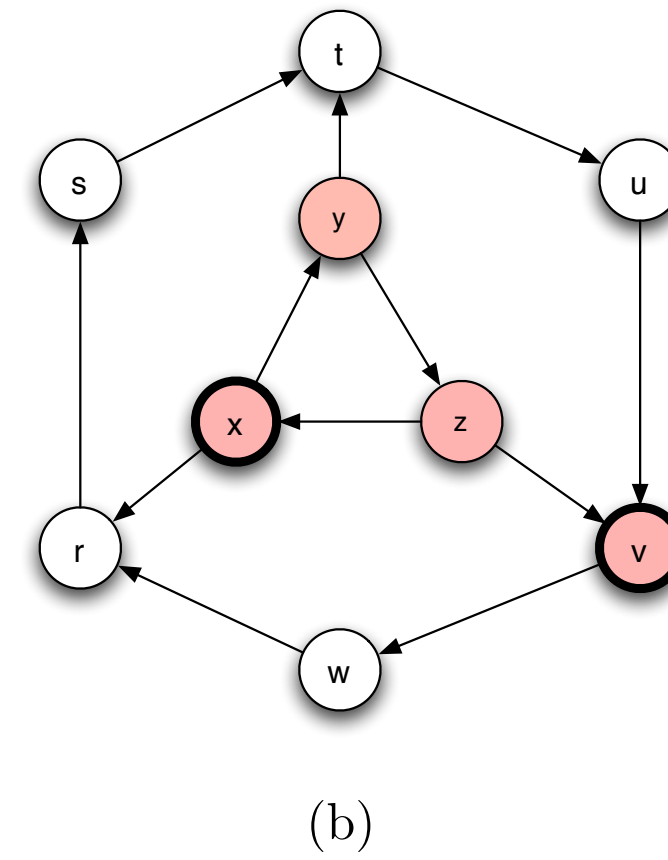
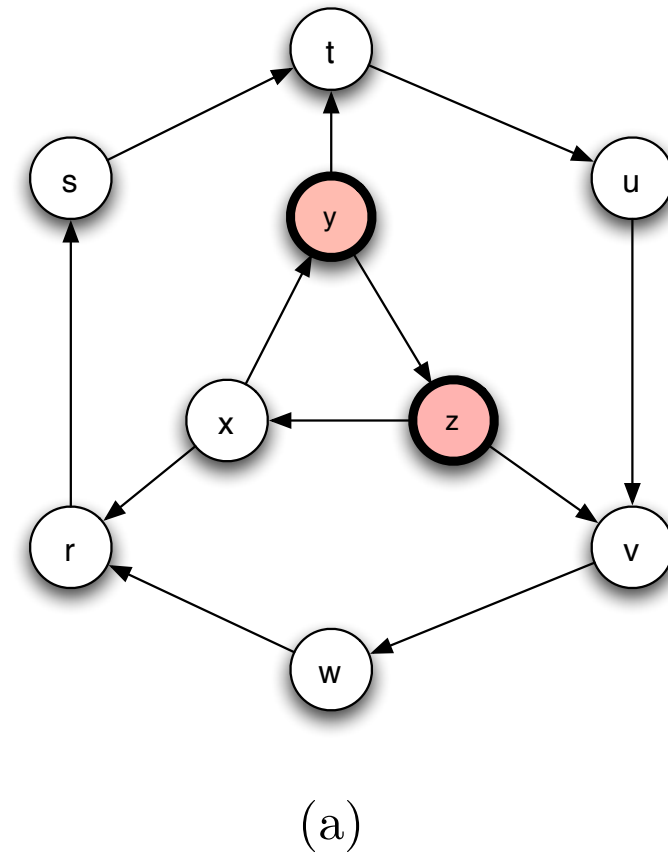
$p = 1/2$   
 $t_i = 1$





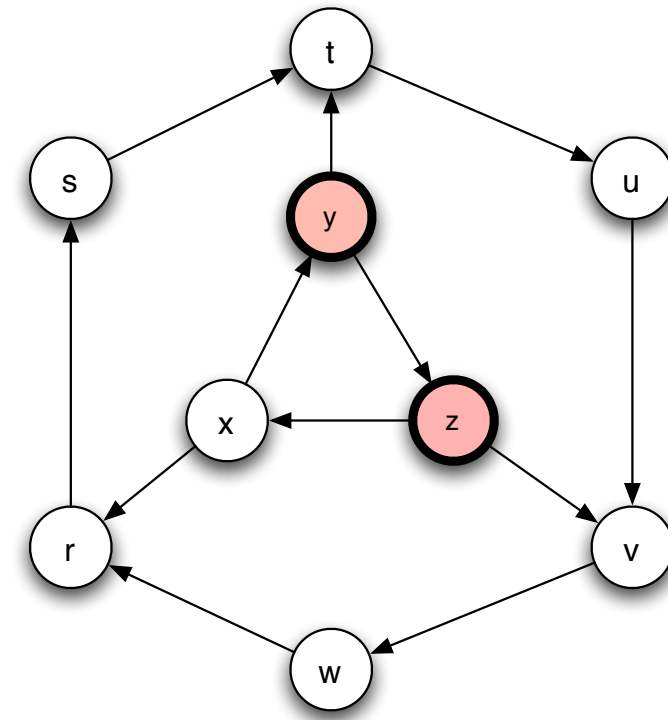
# SIR Epidemics on Networks

$p = 1/2$   
 $t_i = 1$

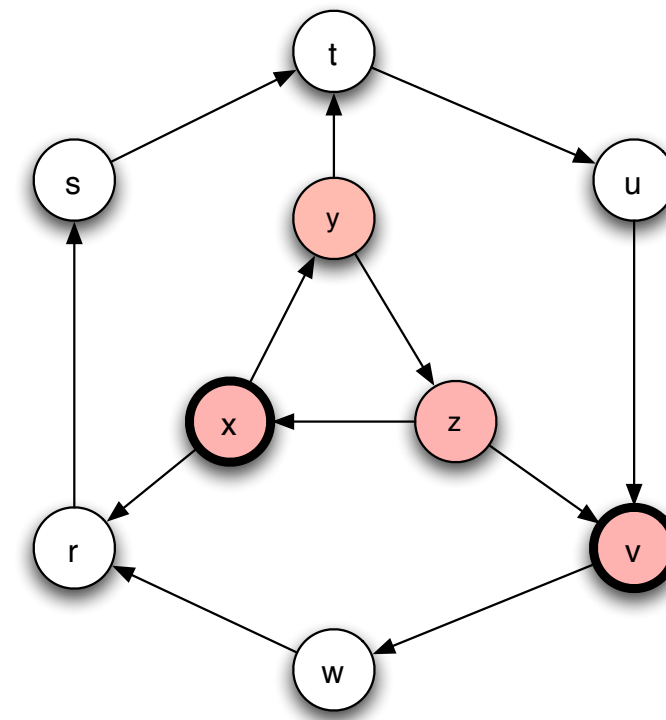


# SIR Epidemics on Networks

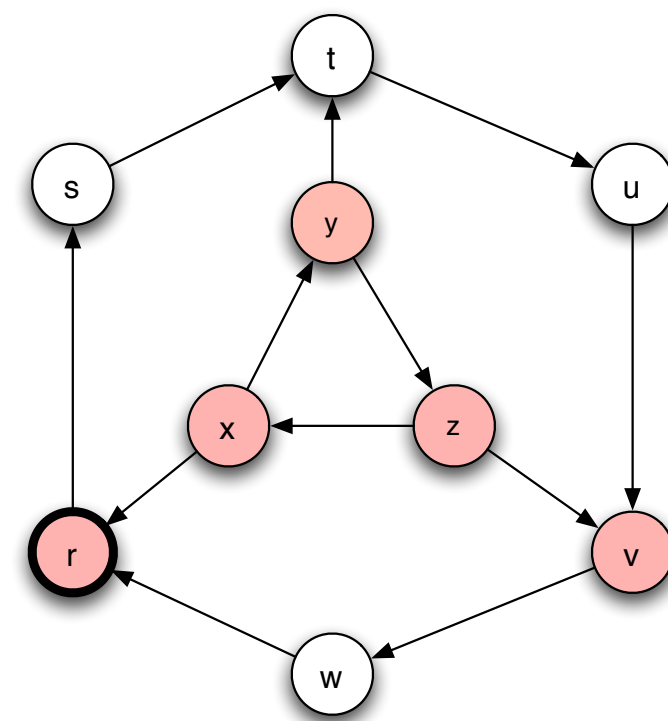
$p = 1/2$   
 $t_i = 1$



(a)



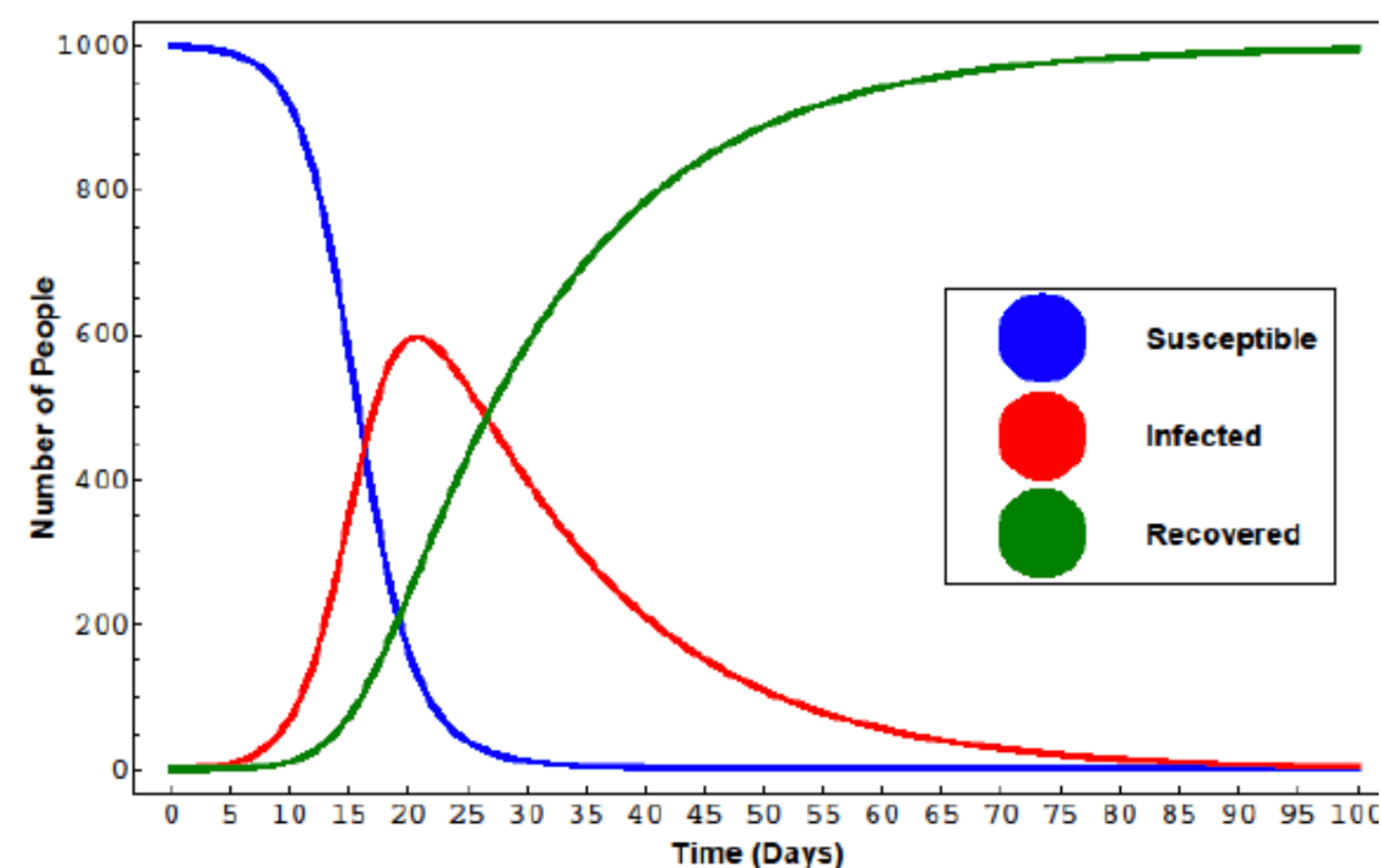
(b)



(c)

# SIR Epidemics on Networks

Typical run of SIR on a graph representing a contact network



**Big questions in epidemiology:** how many will an epidemic infect?  
How will the spread change with changes in parameters?  
Based on that, what are best defences?



# SIR Epidemic Extensions

Many extensions to accommodate different parameters

Some contacts more likely than others:

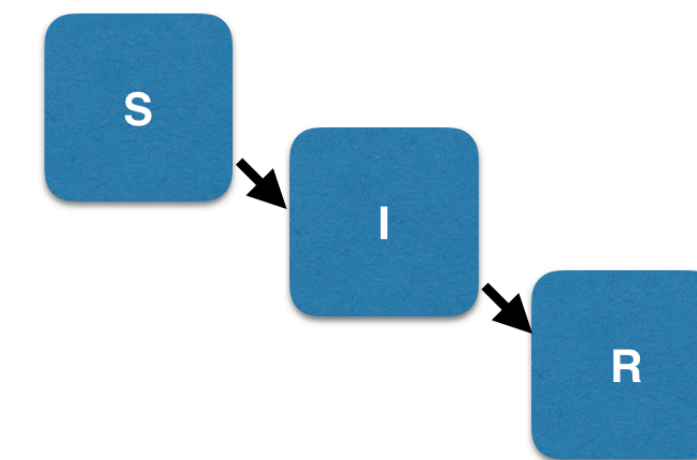
→ probability  $p_{uv}$  that is pair-dependent

Disease goes through different stages (infectious incubation, then less infectious symptomatic transmission):

→ **SEIR or S“III”R**: either **E**xposed state or several different infectious states (with different **p**'s or **t**'s)

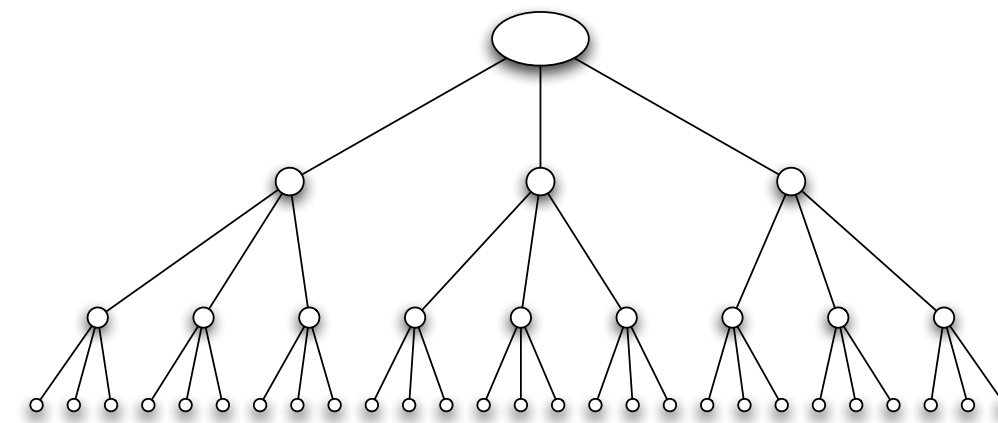
**SIS**: later in the lecture

Mutations (infectiousness, breaking immunity, etc)



# From trees to networks

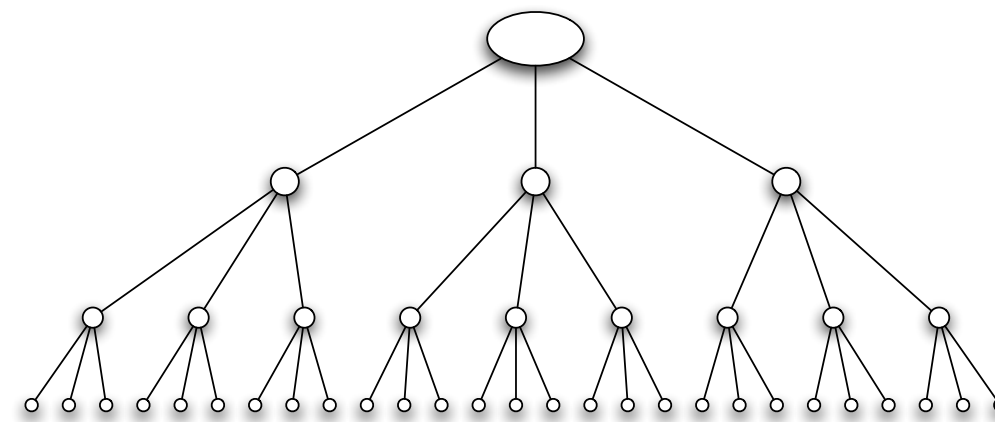
Recall that analysis of  $R_0$  was for trees:



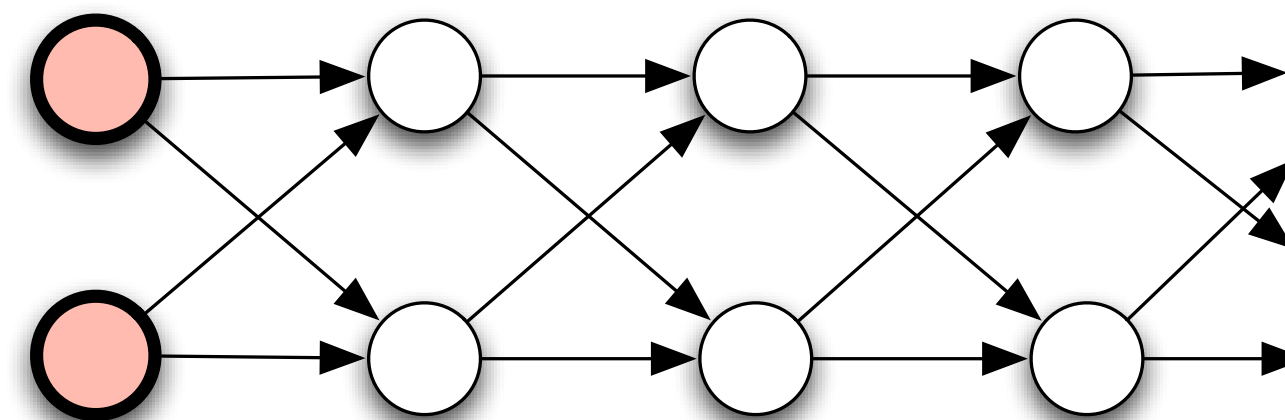
**Do we have the same knife-edge  $R_0 \sim 1$  result in general graphs?**

# From trees to networks

Recall that analysis of  $R_0$  was for trees:



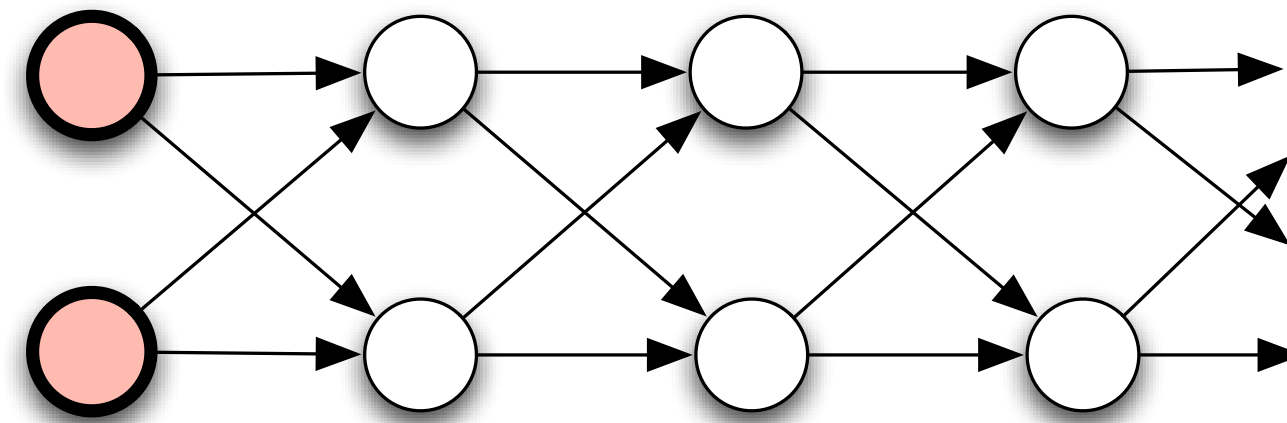
**What happens on other networks?** Consider  $p=2/3$ ,  $k=2$ .





# From trees to networks

What happens on other networks? Consider  **$p=2/3$ ,  $k=2$** .



Calculate  $R_0$  as number of expected new cases per node

$$R_0 = (2/3)*2 = 4/3 > 1$$

But this will almost certainly die out:  $(1/3)^4 = 1/81$  chance that all four edges fail even if both nodes are infected

Prob that this happens after finite number of steps converges to 1

# Now: SIS Epidemic Model

**S** = Susceptible

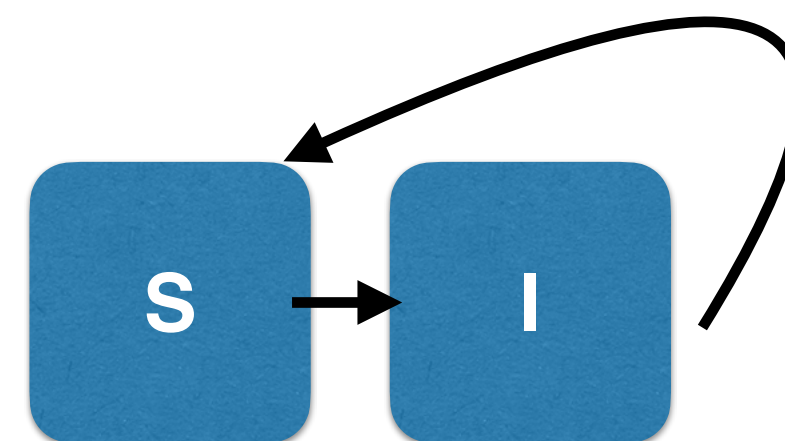
**I** = Infectious: node is infected and infects with prob **p**

Initially some nodes in **I** state, rest in **S** state.

Each node in **I** state remains infected for  **$t_i$**  time steps

During each step, each node has probability **p** of infecting all neighbors

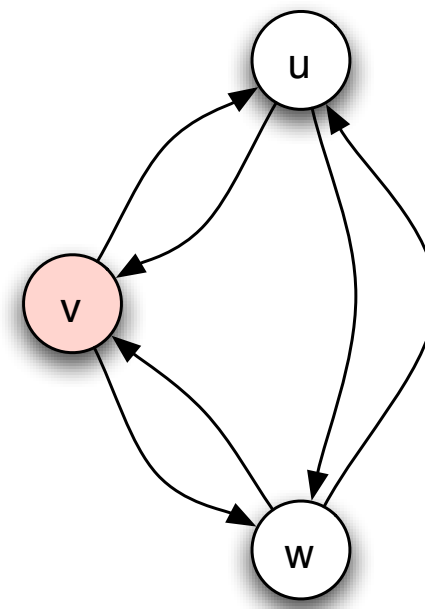
After  **$t_i$**  time steps, node **returns to S**



# SIS Epidemic Example

$$p = 1/2$$

$$t_i = 1$$

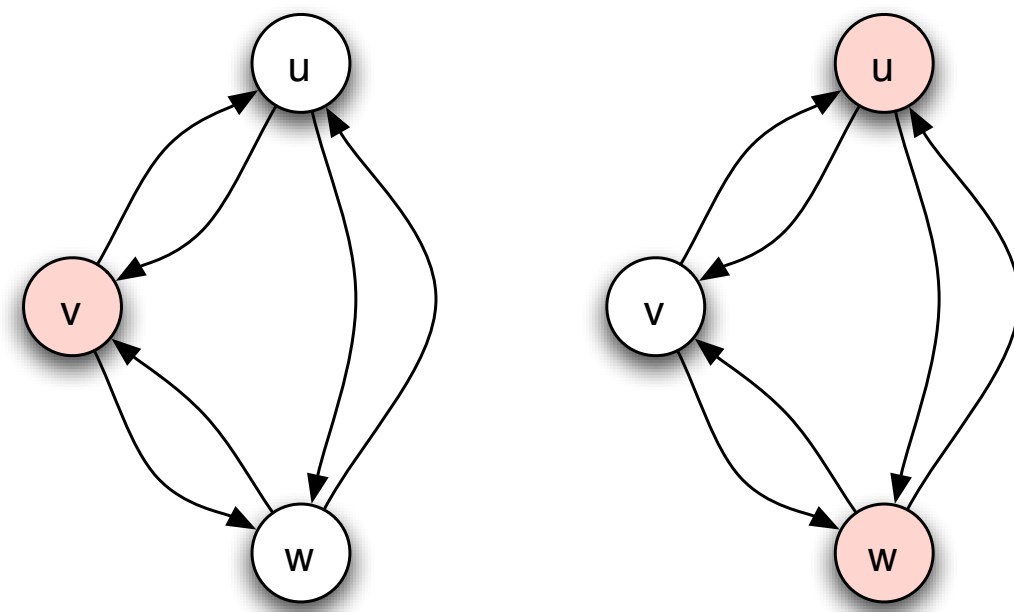




# SIS Epidemic Example

$$p = 1/2$$

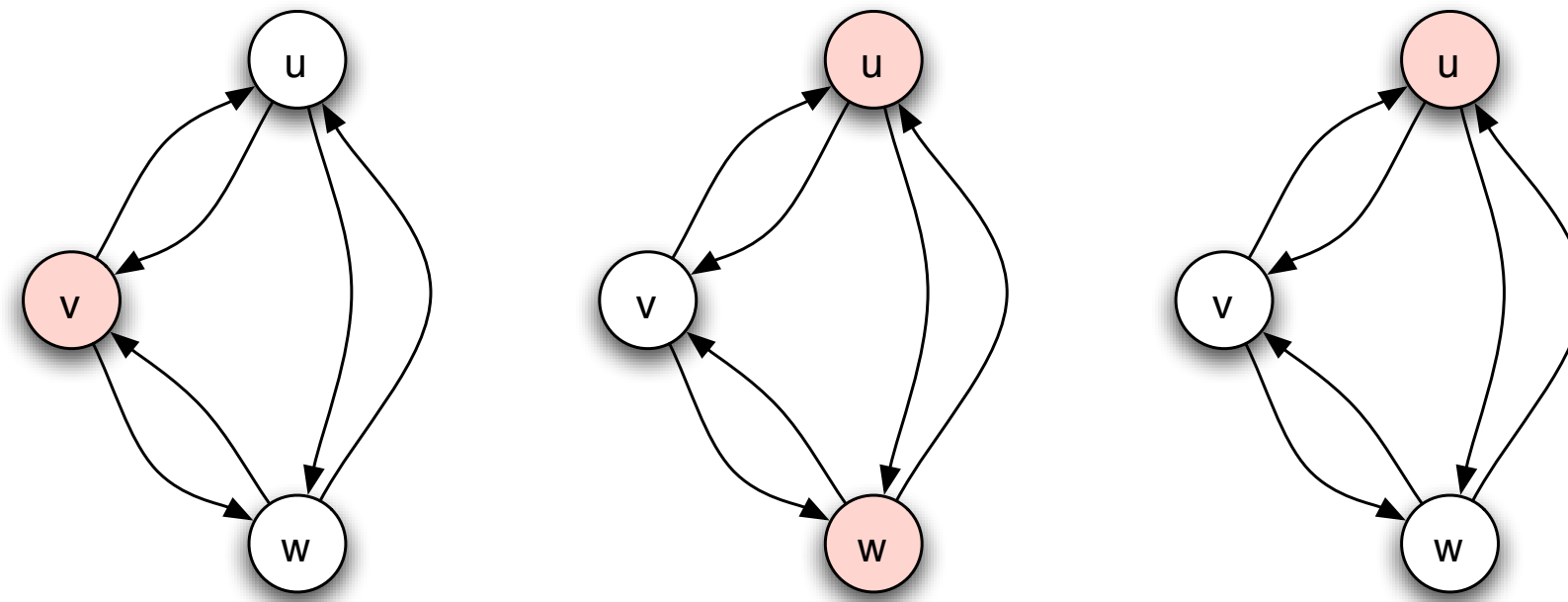
$$t_i = 1$$



# SIS Epidemic Example

$$p = 1/2$$

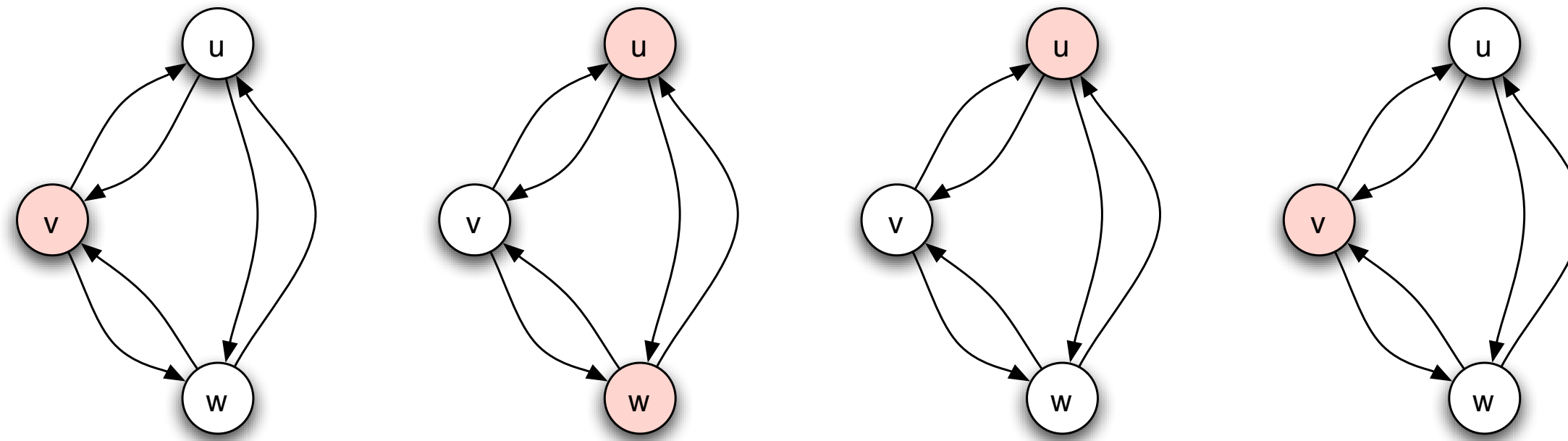
$$t_i = 1$$



# SIS Epidemic Example

$$p=1/2$$

$$t_i=1$$

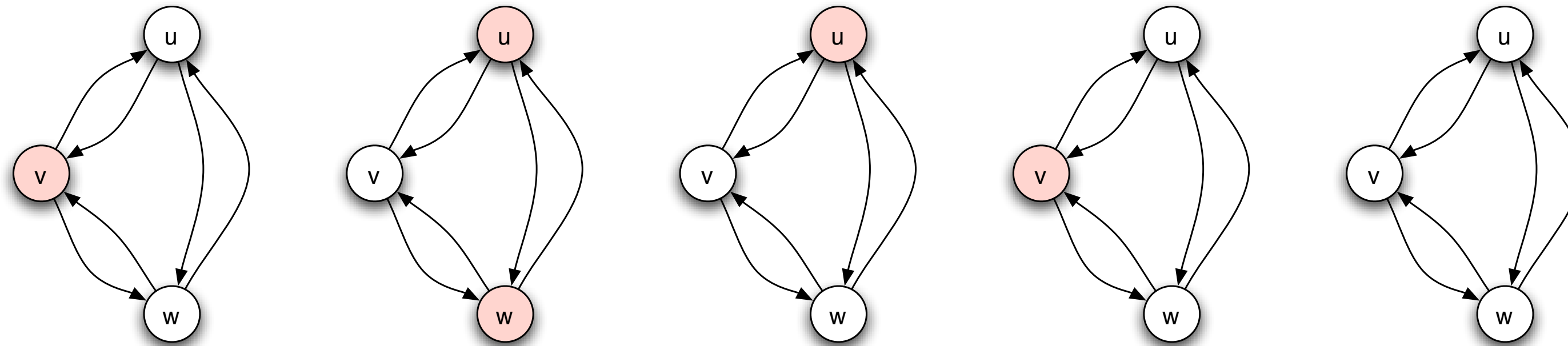




# SIS Epidemic Example

$$p=1/2$$

$$t_i=1$$

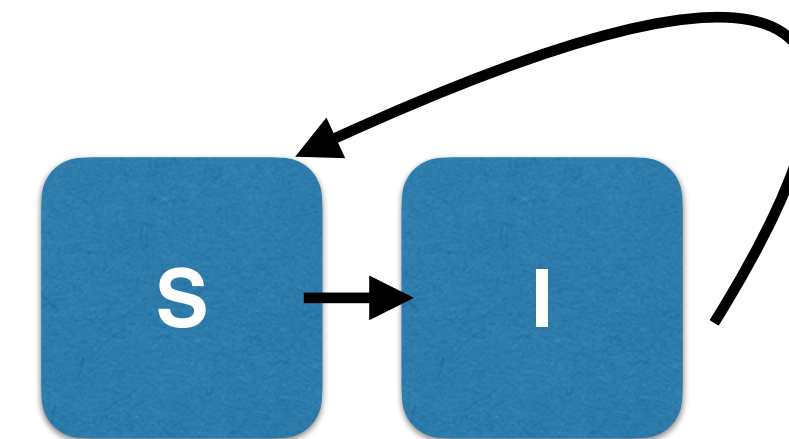
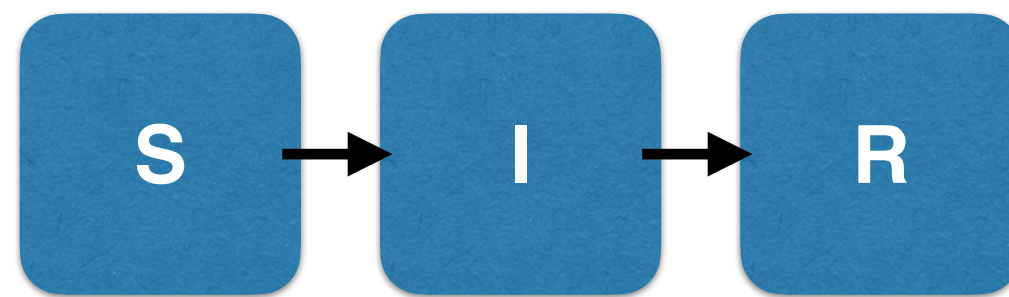


# SIR vs. SIS

**SIR:** “burning through” a finite supply of susceptible

**SIS:** can run for a very long time, cycling through targets

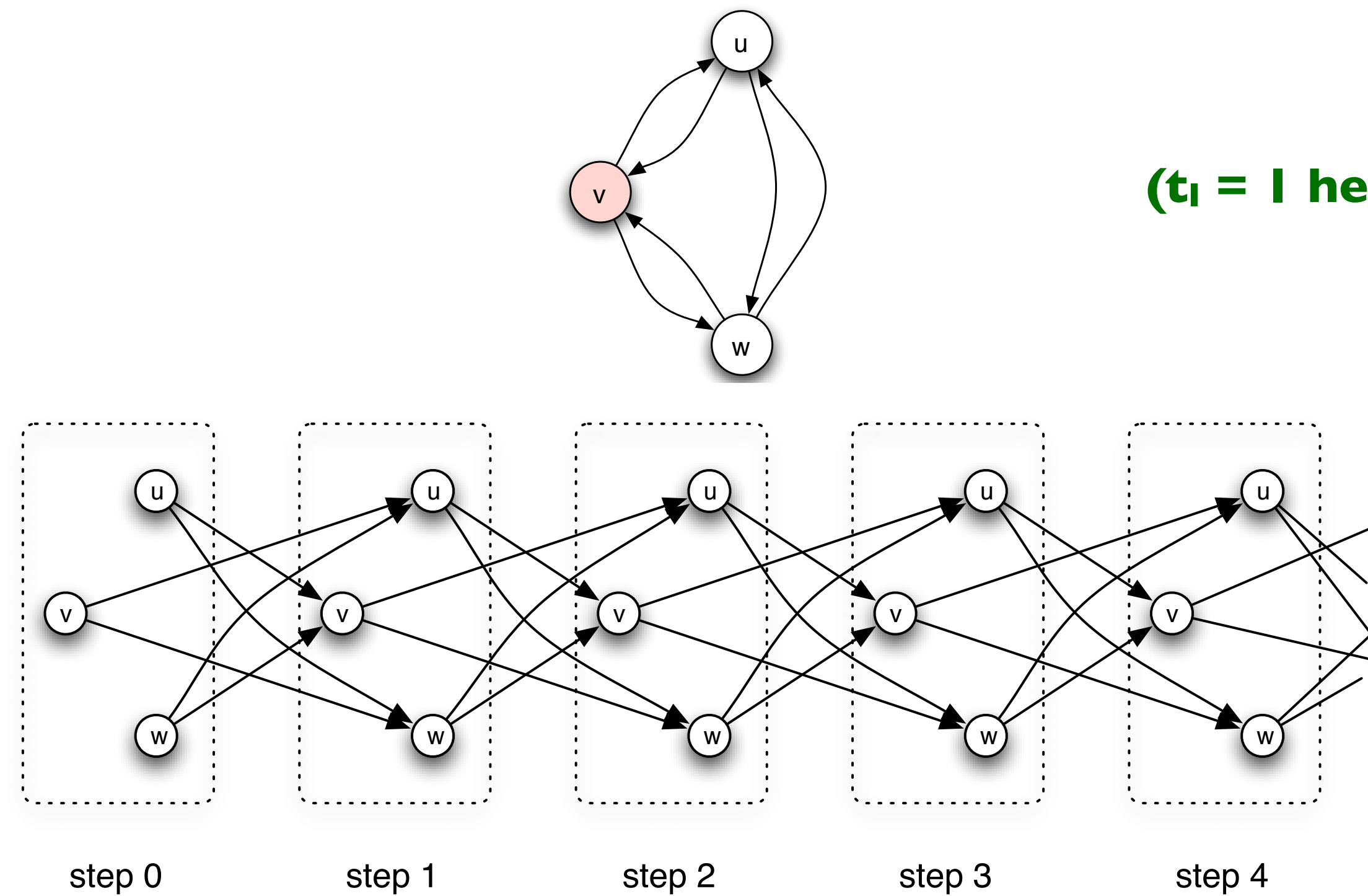
SIS, like SIR, has a **critical threshold** (“knife-edge”/“tipping point”); trickier mathematical analysis. On non-trees both depend on more than just  $R_0$ .



# SIS as SIR on a bigger network

Consider **time-expanded** network: if  $u$  connects to  $v$  in network, have  $u_t$  connect to  $v_{t+1}$

SIS is SIR on a time-expanded network.



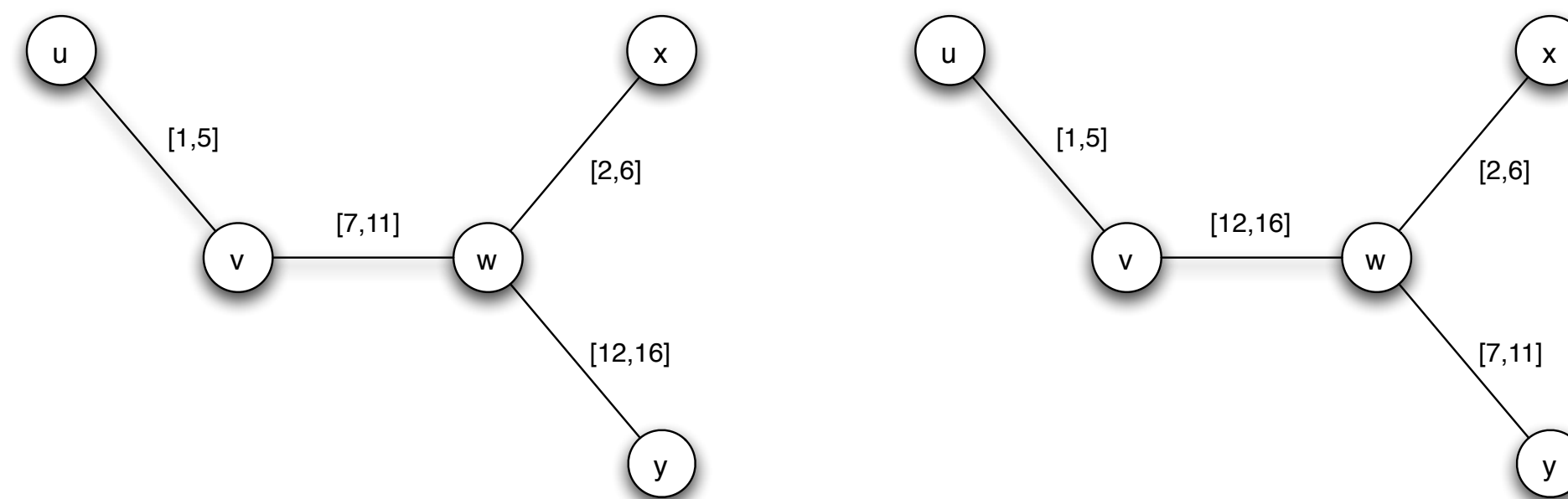


# Transient Contacts & Concurrency

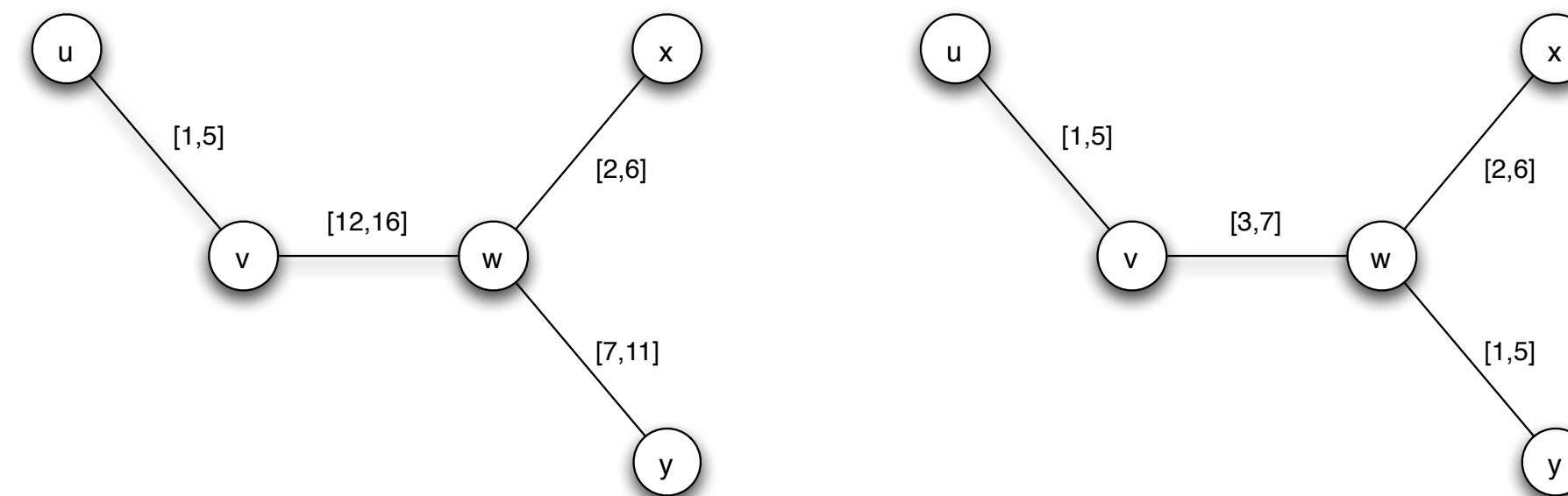
- So far, we've been analyzing **static** networks
- This is reasonable when the rate of transmission is **typically much faster** than edge creation/deletion
- But some epidemic diseases last for years (HIV)
- **When edges are active** becomes very important

# Transient Contacts & Concurrency

A less random model: it matters in what order contact is made in the contact network.



**Concurrency:** having two or more contacts at once.

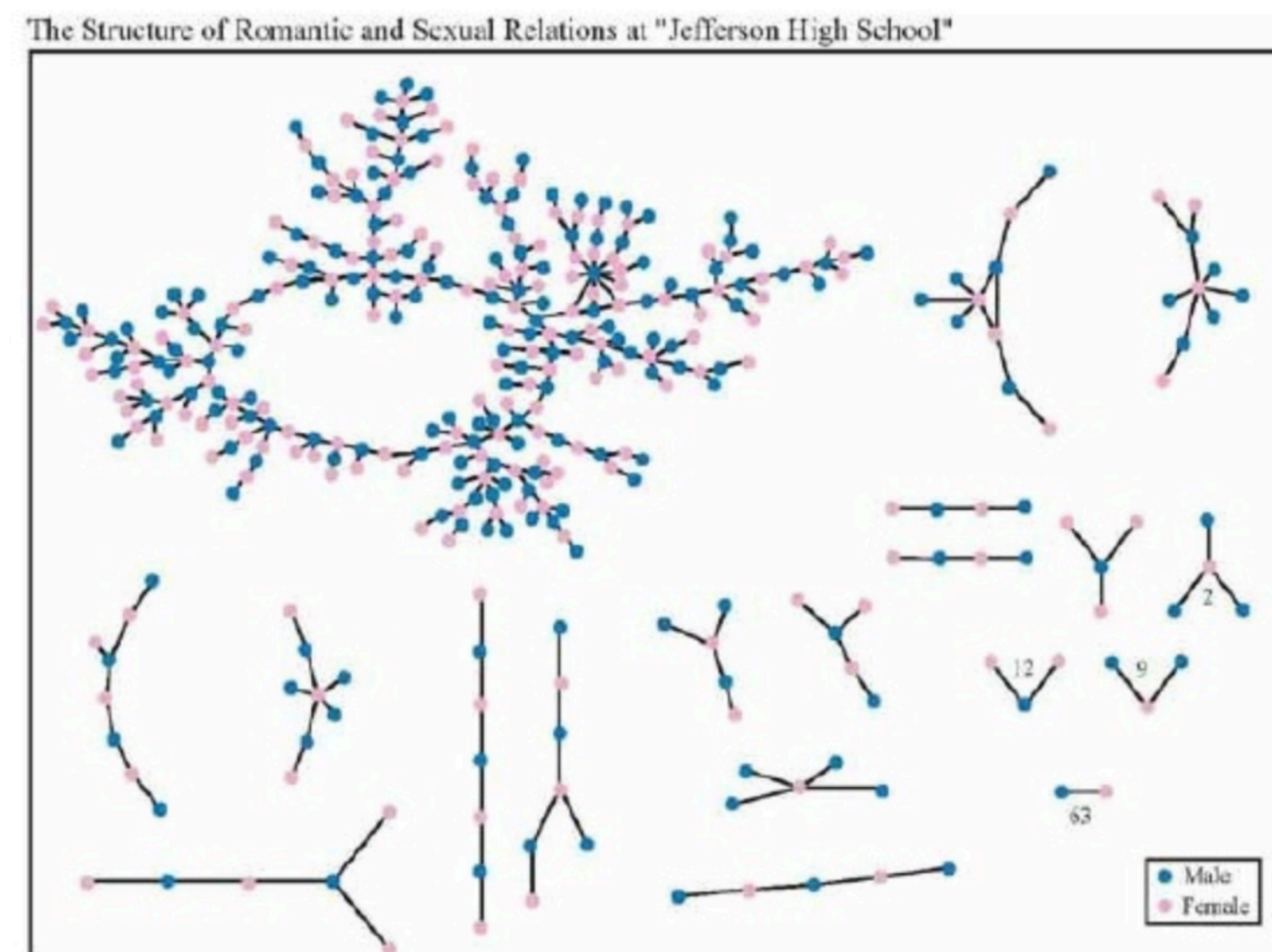


# Transient Contacts & Concurrency

**Small changes in times** can produce **large differences in global epidemic spread**

There are **rich classes of network models** incorporating transience and concurrency

**It's not enough to just know the structure**





# Oscillations

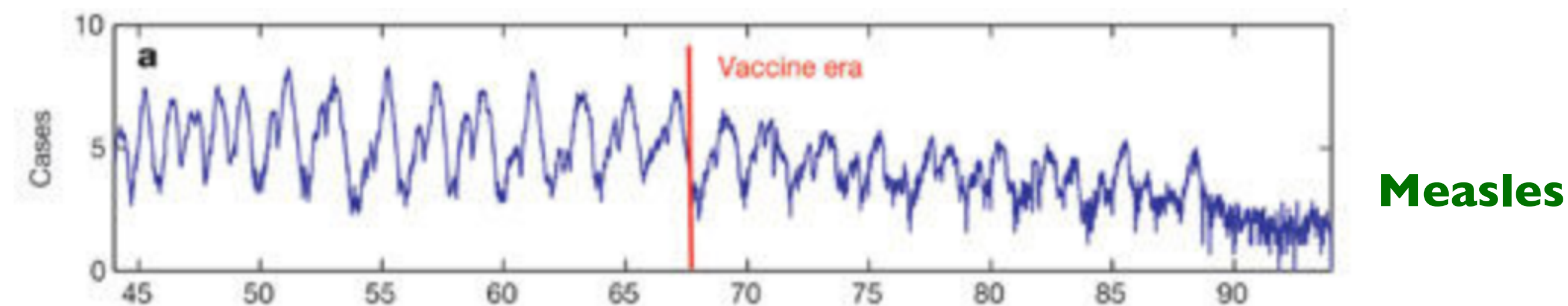
Diseases can be **cyclical** / have **oscillations** (like measles and syphilis)

To model this, vary the model so nodes have **temporary immunity**

**SIRS:** Susceptible, Infected for I steps, Recovered for R steps, then Susceptible again

This can **produce oscillations in very localized** parts of the network

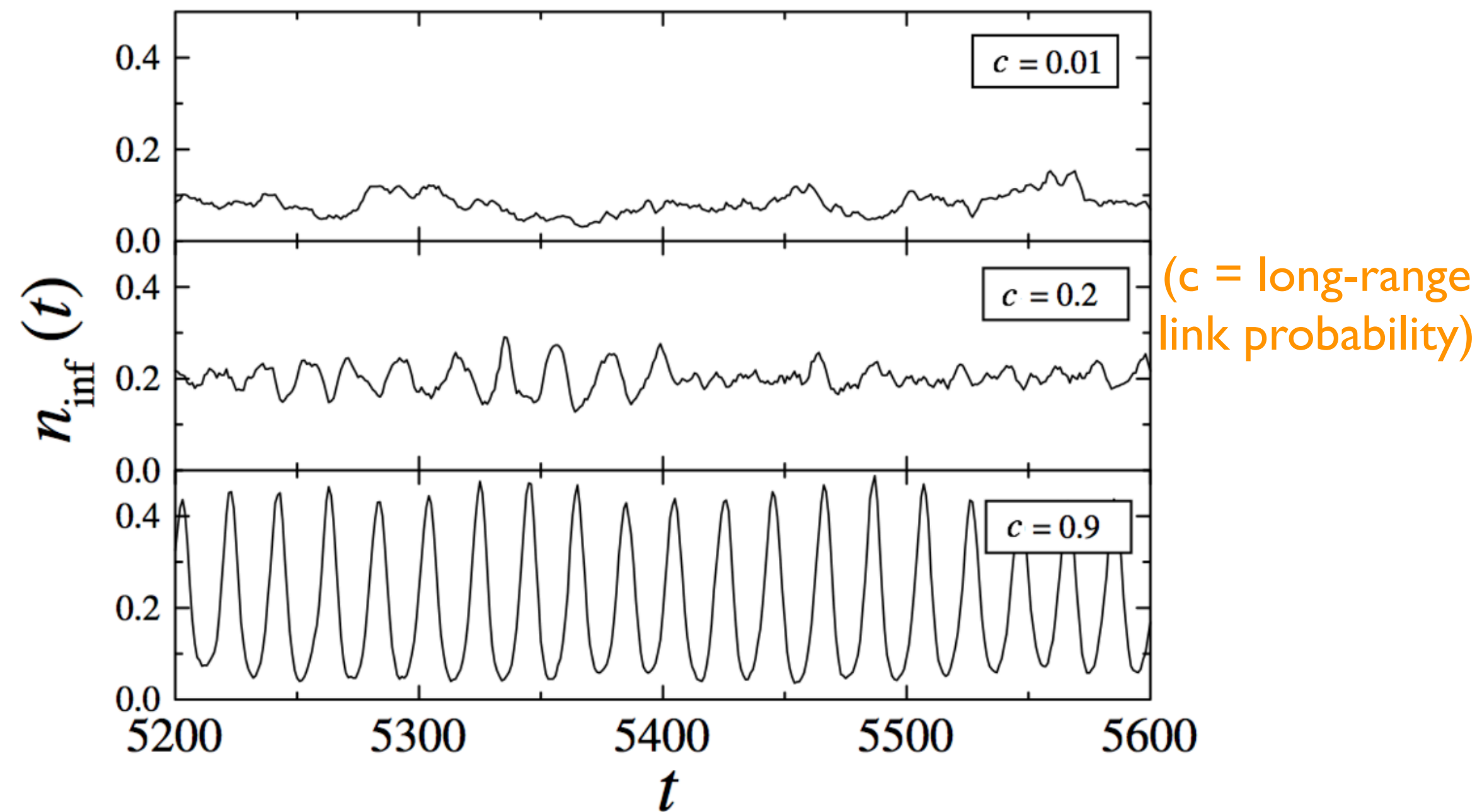
But for large fluctuations at the global network level, need **small-world structure** (random long-range contacts)



# Oscillations

Diseases can be **cyclical** / have **oscillations** (like the flu)

But for large fluctuations at the global network level, need **small-world structure** (random long-range contacts)



# Epidemics vs. Behaviour

In epidemic models, nodes get infected from **one** particular other node

To model information spread, people often use epidemic models ("viral diffusion")



But many social phenomena (behaviours, beliefs, practices, etc.) are **complex**: costly, risky, uncertain, etc.

# Epidemics vs. Behaviour

When a behaviour is **risky, costly, or uncertain**, you **may not do it just because one of your friends is** (but this is what epidemic diffusion looks like)

Social movements, health technologies, political activism, etc.

E.g.: PrEP medication is the best latest in HIV prevention → one pill a day gives 90% prevention.

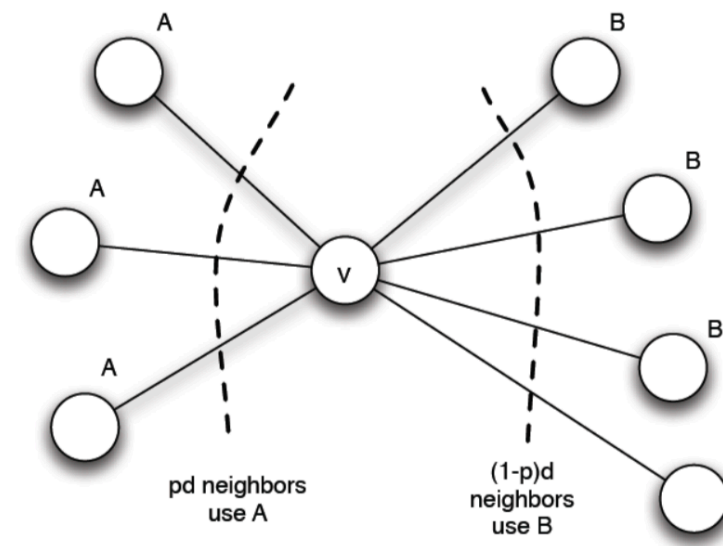
But in two trials in sub-Saharan Africa, it didn't work ... because no one was taking it! (fears of discrimination, etc.)

How do you get behaviour to diffuse?

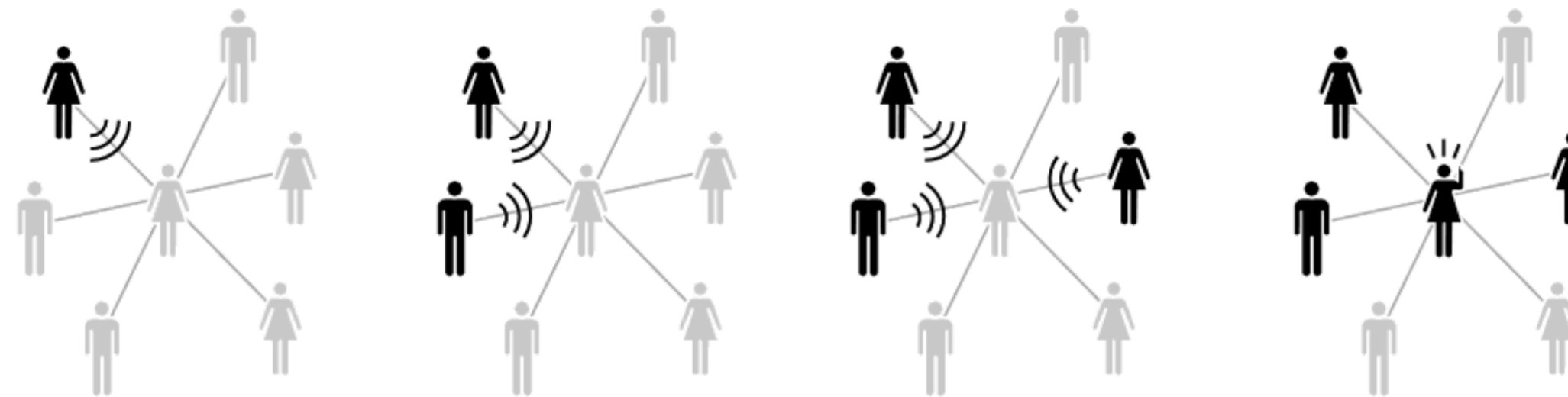


# Epidemics vs. Behaviour

Previously we saw a model of behaviour diffusion based on utility



This is an example of *complex* diffusion: in general, need more than one neighbour to adopt before you adopt a behaviour.

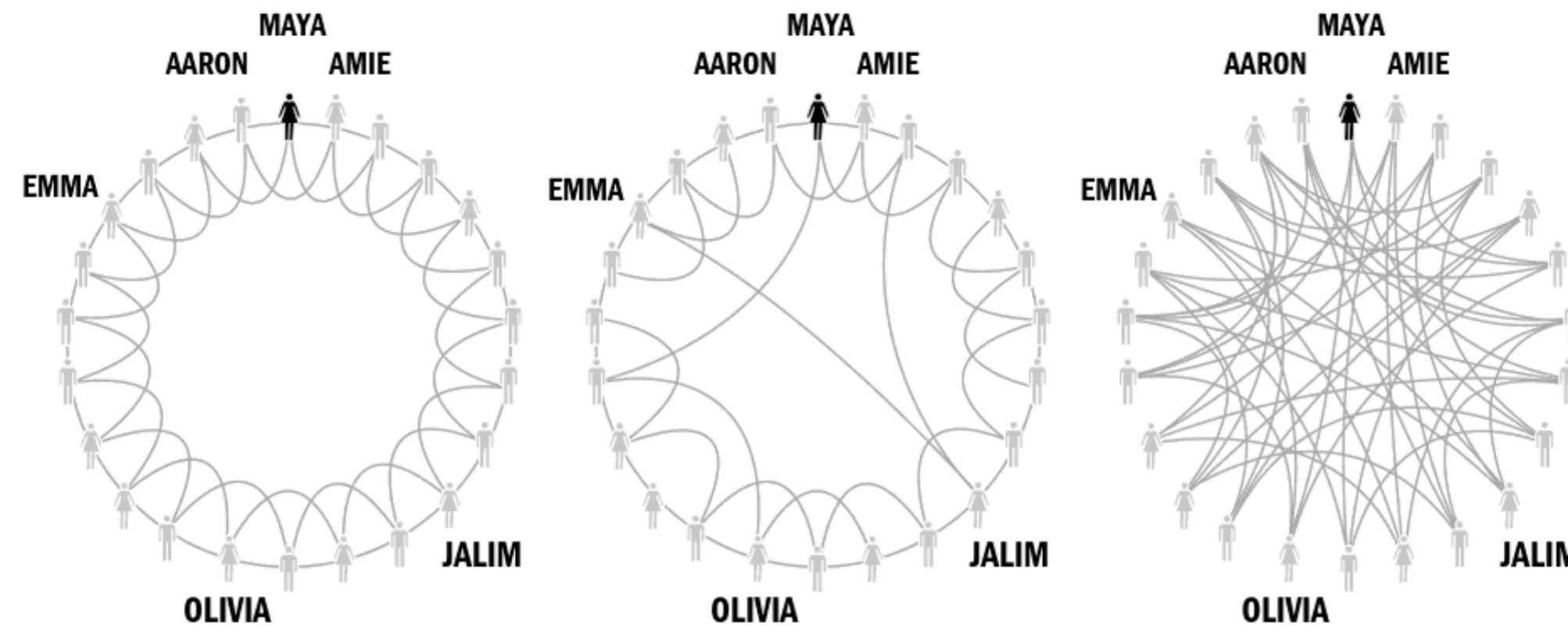


# Epidemics vs. Behaviour

Simple vs. complex diffusion  
Epidemics vs. behaviour

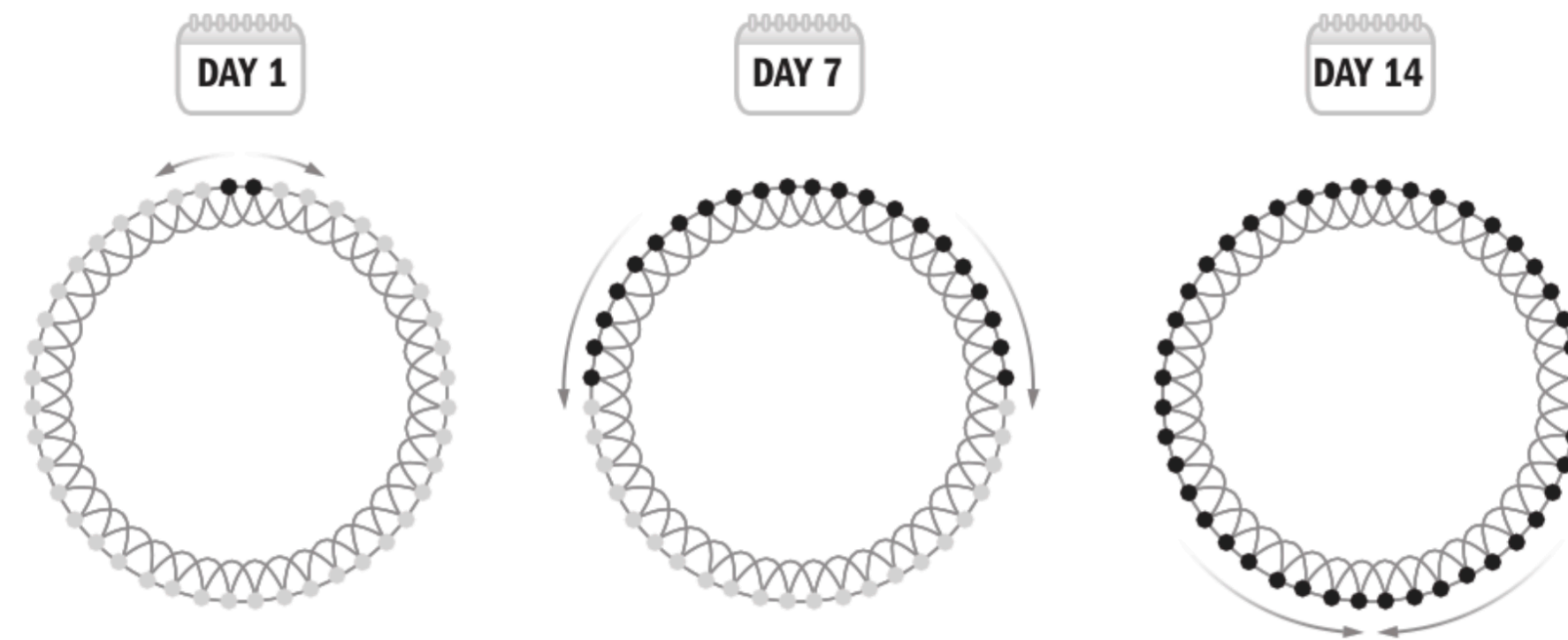
What's the difference?

**Recall the small-world model**

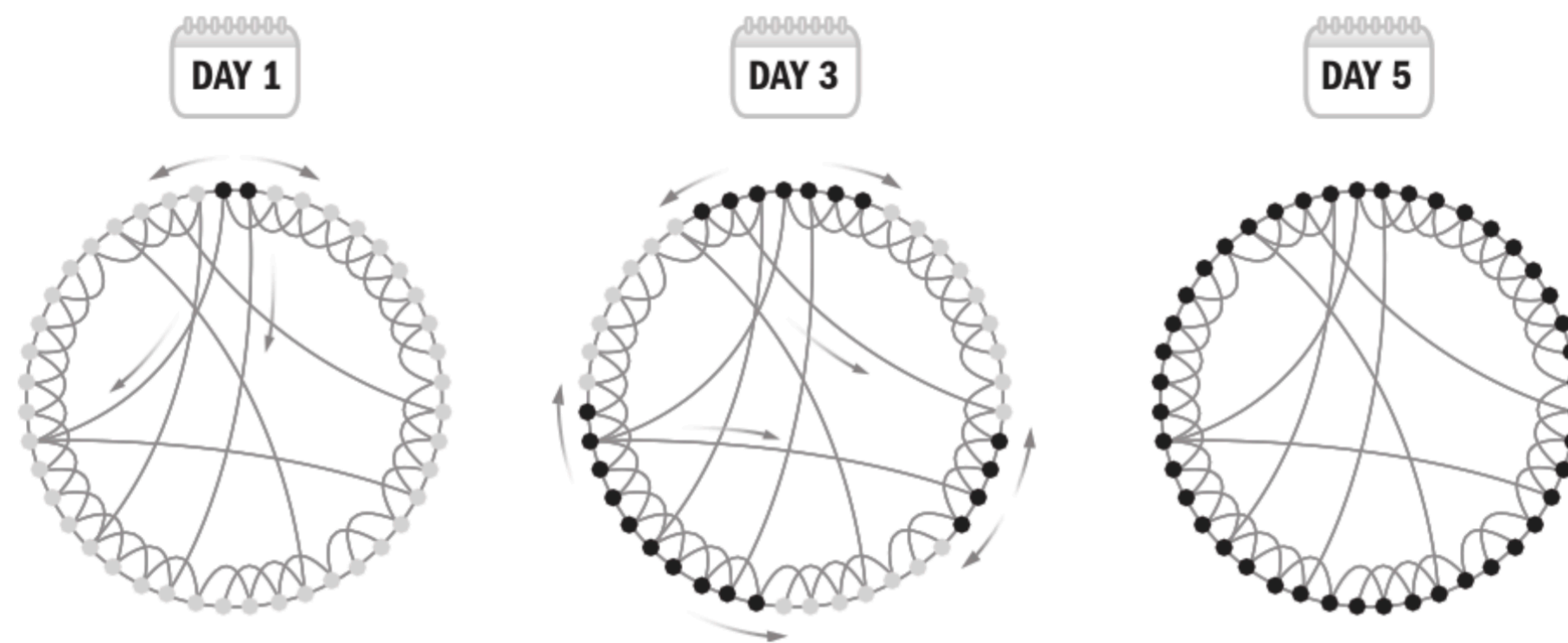


# Simple Diffusion

Large world:

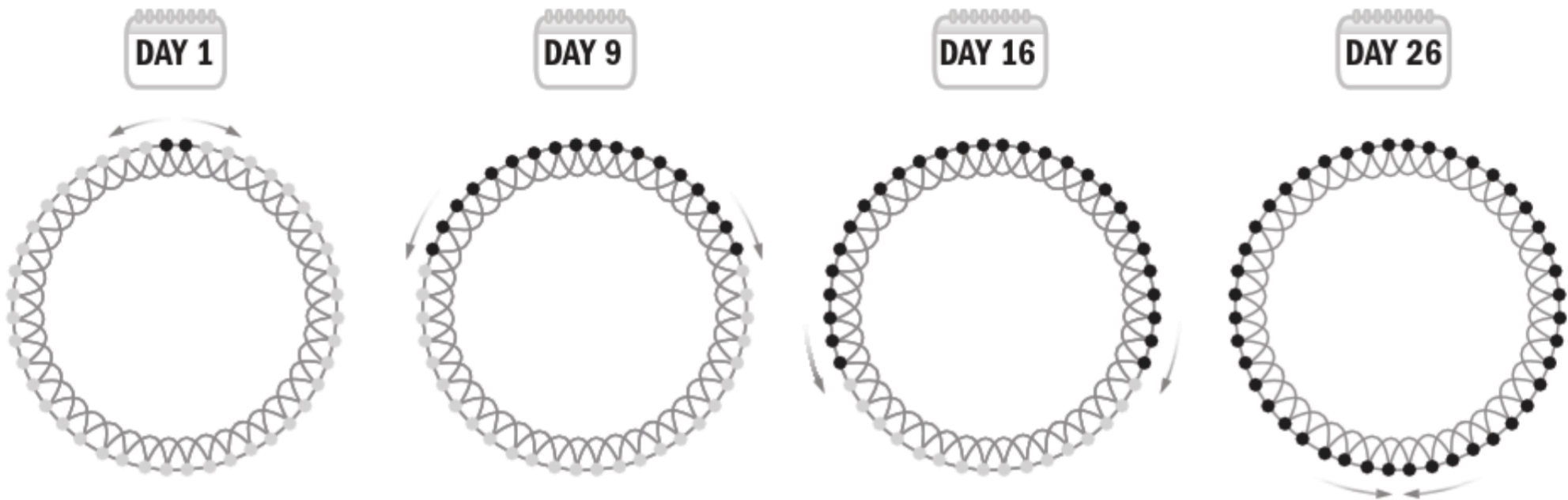


Small world:

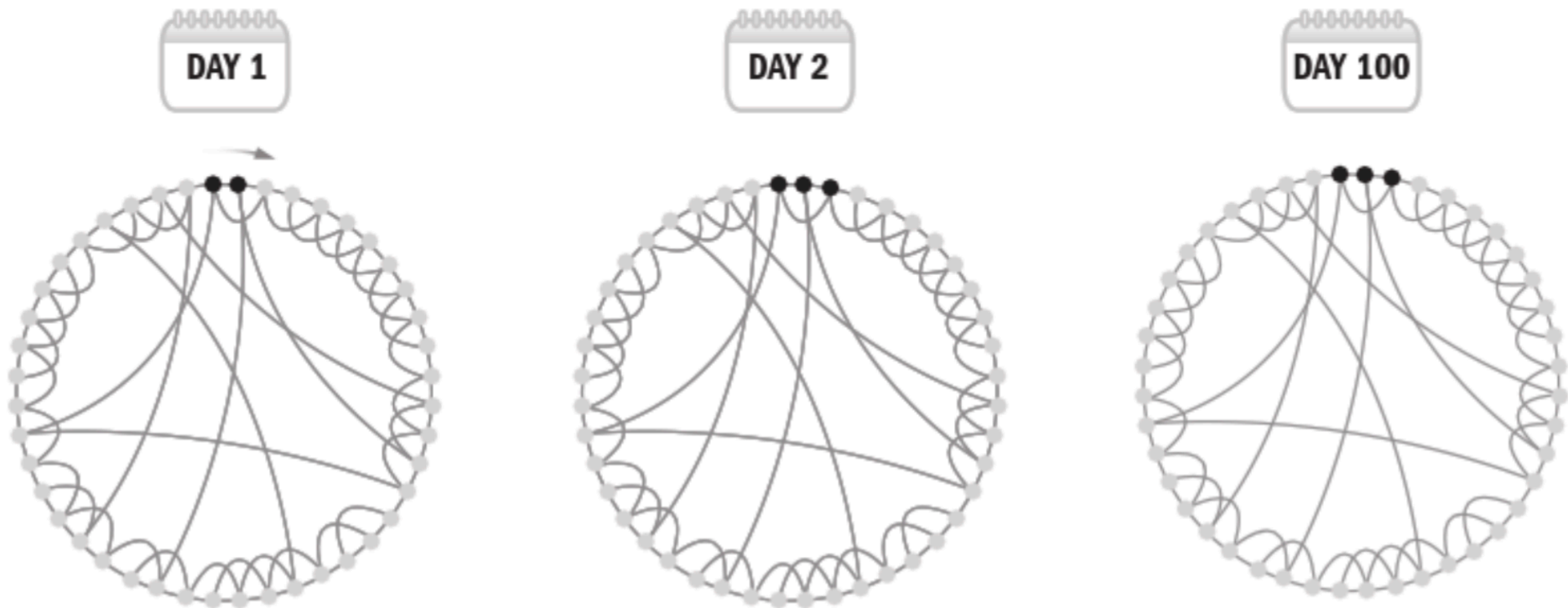


# Complex Diffusion

Large world:



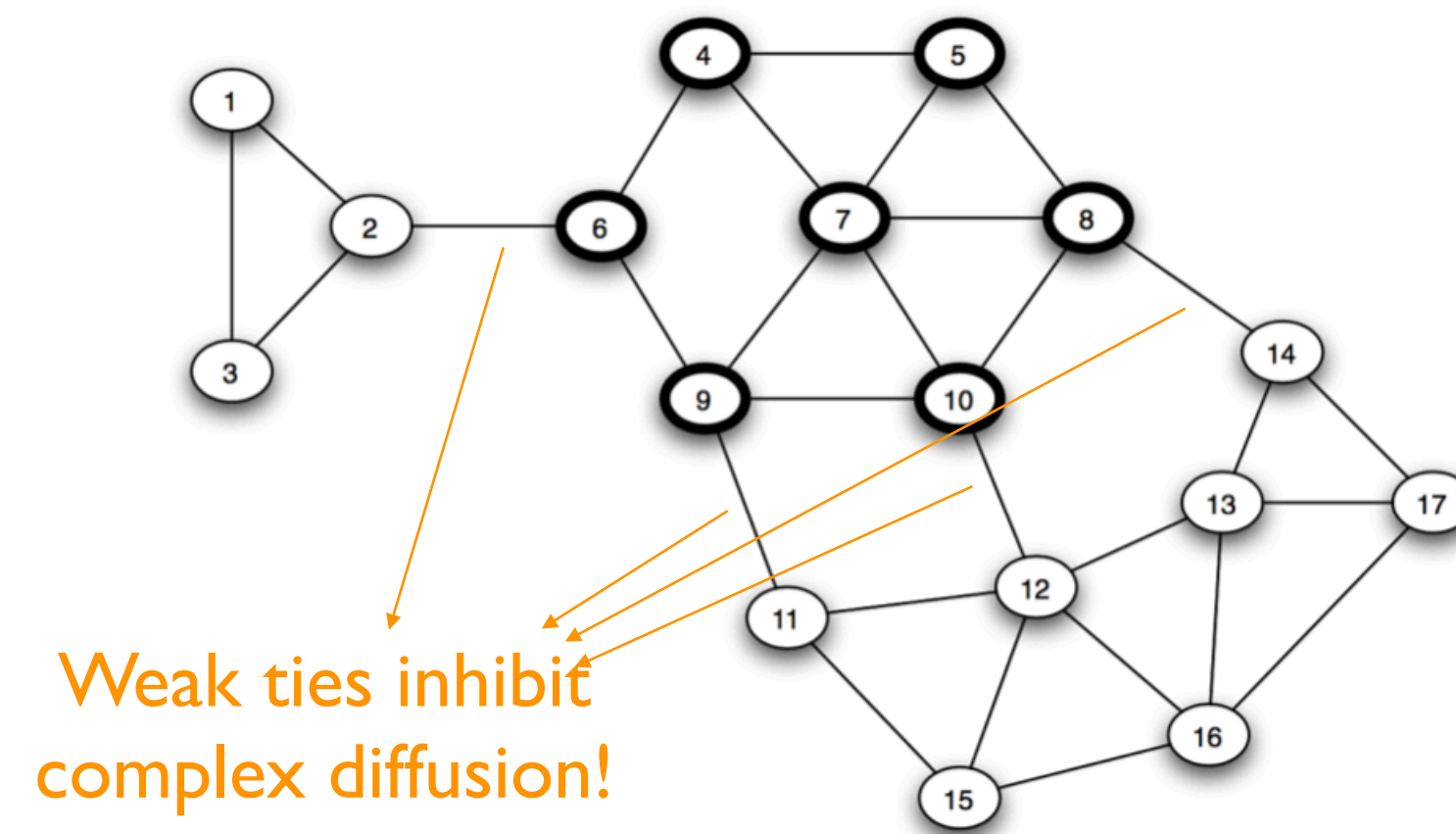
Small world:





# Simple vs. Complex Diffusion

Weak ties are extremely **useful for simple diffusion** and contagion, but they **inhibit complex diffusion!**



# Voting

# Voting

Why have voting?

**Synthesize the preferences of a group**

Aggregate information, preferences, beliefs, decisions

Voting on:

Candidates

Laws

Verdicts for trials

Awards



# Simple example

Say you want to pick the fairest outcome for the group

**Everyone has their preferred number (e.g. price)**

What should you do?

Easy...take the average

**Why fair? Minimizes the squared loss**





# Why voting is hard

But in many situations there is no natural **“average”**!

Voting on:

Candidates

Laws

Verdicts for trials

Awards

**Averaging fails here...**



# Why voting is hard

Often need to pick a **single winner** that becomes **binding for the group**

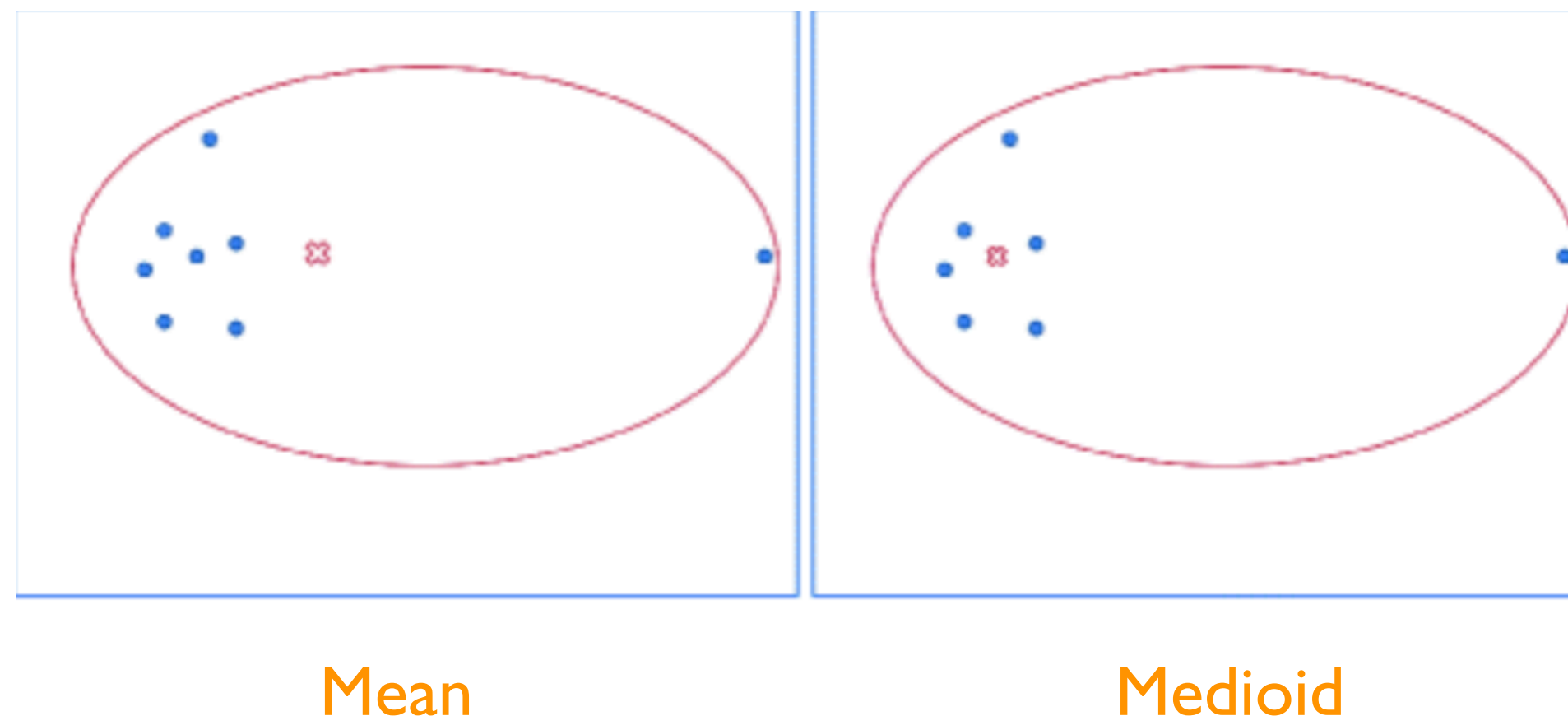
President

Award-winner

Policy decision

Voting as **group decision making**

**Parallels to clustering:** finding the centre vs finding the “medioid”—the best representative element





# Individual preferences

We want to **aggregate many individuals' preferences**

What are individual preferences?

Setup: a group of  $k$  **people** are evaluating a finite set of possible *alternatives*





# Individual preferences

The people want to produce a single **group ranking** that orders the alternatives from best to worst

The ranking should **reflect the collective opinion of the group**

The challenge: how do we define what it means to reflect multiple, potentially contradictory opinions?



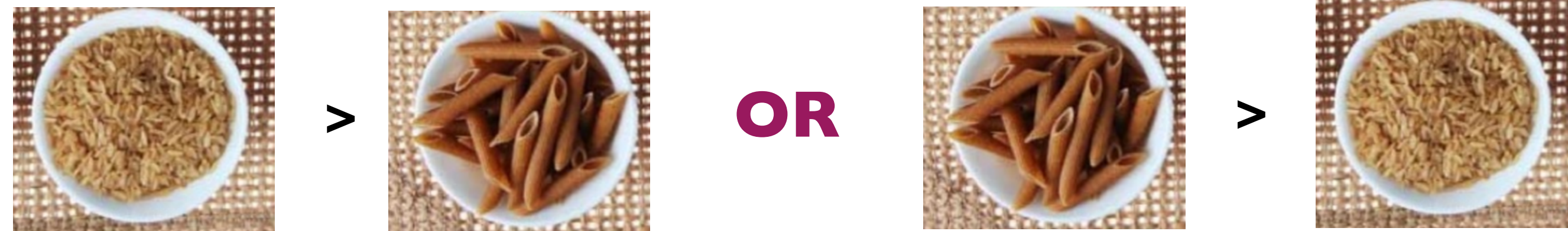


# Individual preferences

Every person has a **preference relation** over the alternatives, denoted  $\succ_i$  for player  $i$

Must satisfy two properties:

**Complete:** all pairs of distinct alternatives  $X$  and  $Y$ , either  $X \succ_i Y$  or  $Y \succ_i X$



**Transitive:** if  $X \succ_i Y$  and  $Y \succ_i Z$  then  $X \succ_i Z$



# Individual preferences

A way to think about preference relations: as a **graph**

**Nodes:** alternatives

**Directed edges:**  $Y \rightarrow X$  if  $X \succ_i Y$



**(complete and transitive example)**



# Individual preferences

Another way of expressing preferences: ranked list

**For example:**



Ranked list  $\rightarrow$  preference relation

Obviously complete and transitive

Preference relation  $\rightarrow$  ranked list

Less obvious but still true

# Individual preferences

Claim: Ranked list  $\rightarrow$  Preference relation

Proof:

A ranked list is **complete**, since for any pair of alternatives  $X$  and  $Y$ , either  $X > Y$  or  $Y > X$

A ranked list is **transitive**, since if  $X$  is higher than  $Y$  and  $Y$  is higher than  $Z$ , then  $X$  is also higher than  $Z$ .



# Individual preferences

Claim: Preference relation  $\rightarrow$  ranked list

Proof:

Identify the alternative  $X$  that wins the most pairwise comparisons

Claim:  $X$  actually beats **every** other alternative

Why? Suppose  $Y \succ_i X$ . Then  $Y$  would beat everything  $X$  beats (by transitivity), and also  $X$ . Therefore beats more than  $X$ . **Contradiction!**

**Put  $X$  at the top of the list, remove it from the set of alternatives, and recurse**

Relation is **still complete and transitive** over remaining alternatives

Construct a list by **repeatedly finding the alternative** that beats everyone else

# Individual preferences

Summary:

Preference relation  $\rightarrow$  Ranked list

Ranked list  $\rightarrow$  Preference relation

Therefore preference relations and ranked lists are equivalent!

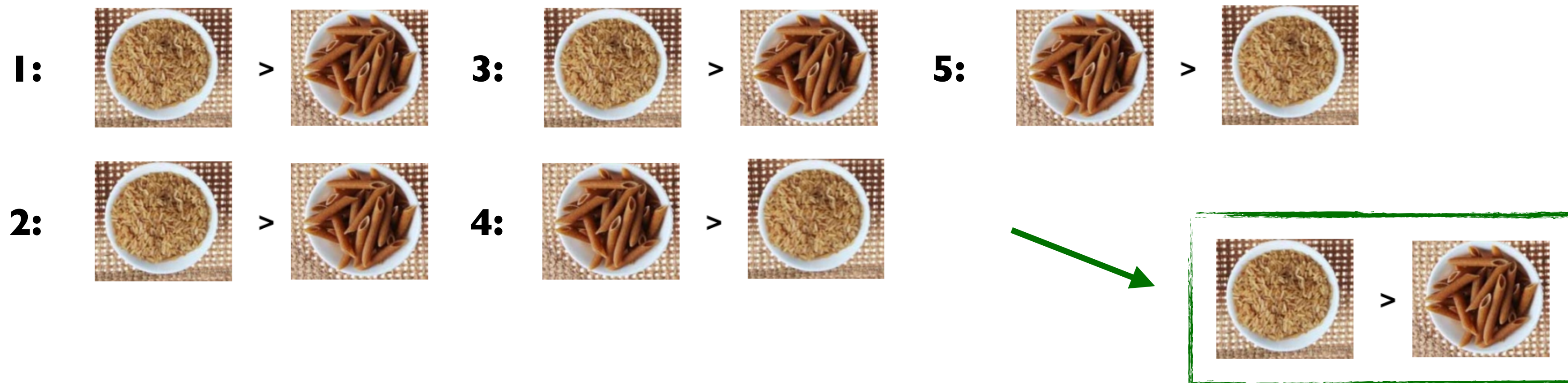
# Voting Systems

Voting system: a **method** that takes a set of **complete** and **transitive** individual preference relations (or ranked lists) and **outputs a group ranking**

When there's only two alternatives, what should we do?

**Majority Rule**: whoever is preferred by a majority of the voters wins, other one is second

(let  $k$  be odd to avoid ties)



# Majority Rule

Easy enough, what about majority rule with more than two alternatives?

**What's a natural way to extend it?**

Majority rule on every pair of alternatives:  $X > Y$  if a majority of voters have  $X >_i Y$

**Is this complete?**

Everyone has a preference for every pair, and there's always a majority (assume  $k$  is odd). So this is **complete**

**Is this transitive?**



# Majority Rule

Is majority rule on at least 3 alternatives transitive?

1:



$>_1$



$>_1$



2:



$>_2$



$>_2$



3:



$>_3$



$>_3$



What does majority rule do here?



# Majority Rule

Is majority rule on at least 3 alternatives transitive?

1:



$>_1$



$>_1$



2:



$>_2$



$>_2$



3:



$>_3$



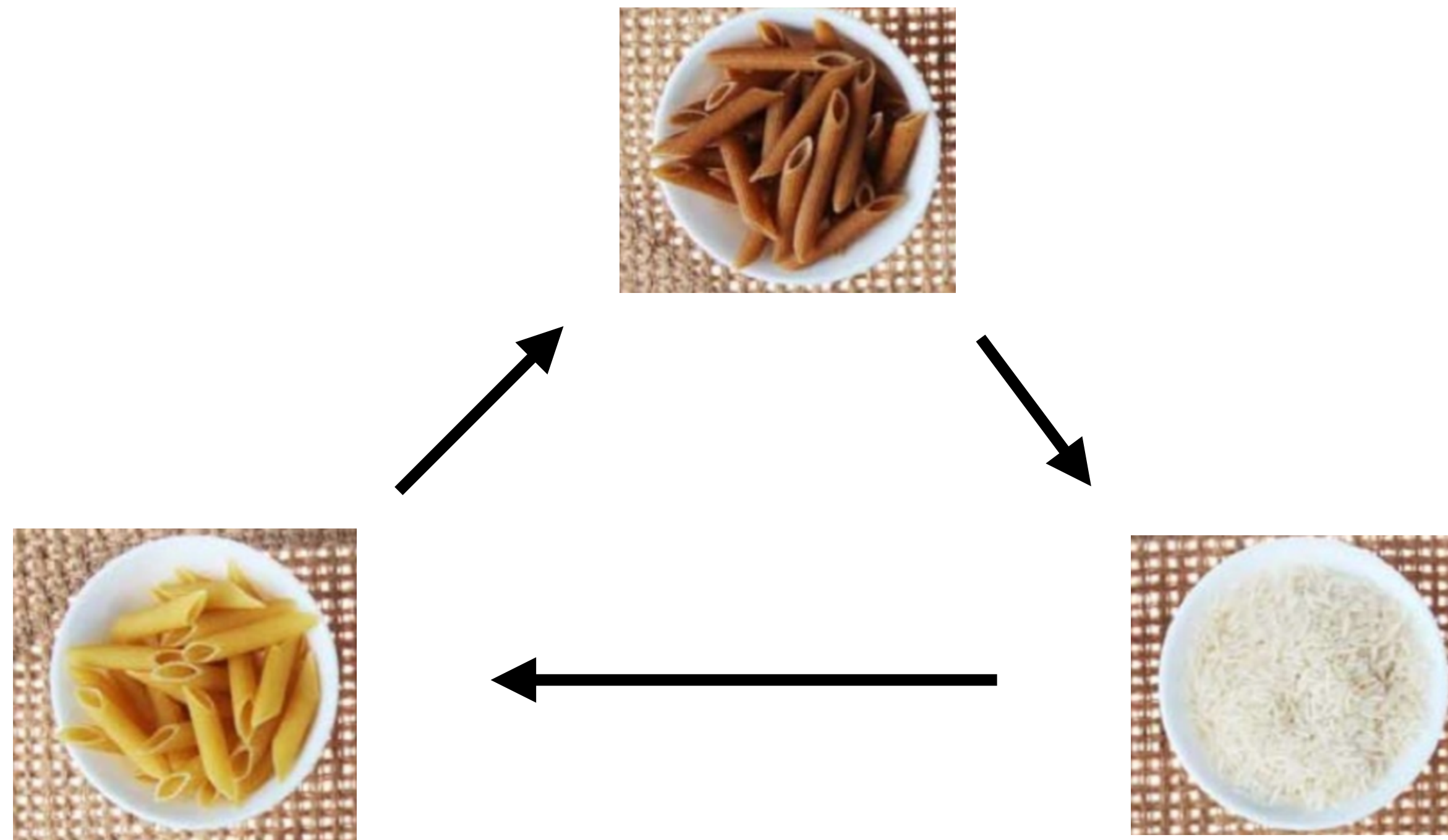
$>_3$



Y pasta  $>$  B pasta, B pasta  $>$  rice, rice  $>$  Y pasta!

# Majority Rule

Majority rule with at least three alternatives can produce a *non-transitive* group ranking



**Cycle on preferences => non transitive => bad!**



# Condorcet Paradox

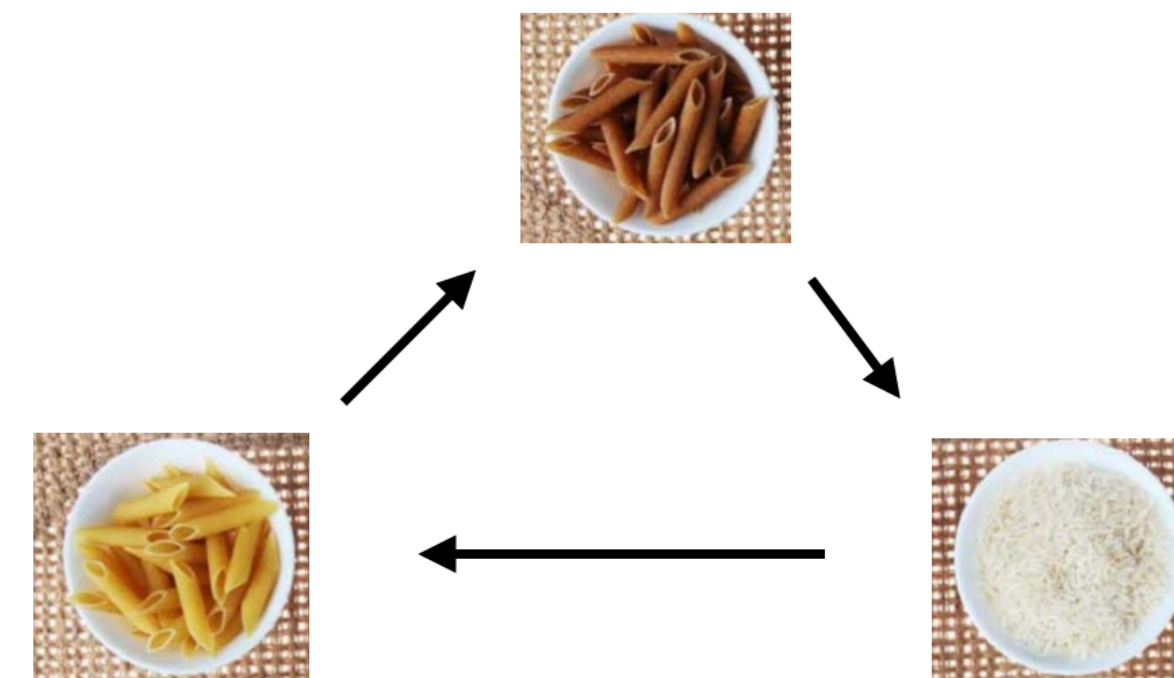
Majority rule with at least three alternatives can produce a *non-transitive* group ranking

Called the “Condorcet Paradox”

Really bad news!

Everyone had **perfectly plausible preferences**

But they **behave incoherently** as a group, can't even decide on a favourite





# Condorcet Paradox

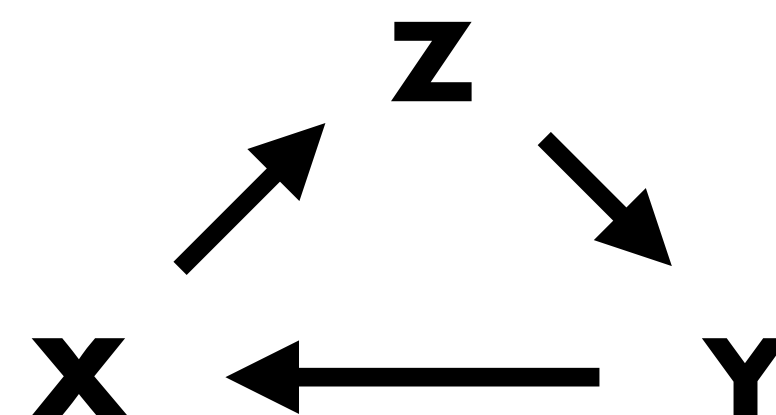
**Condorcet Paradox can even happen within a single individual person**

Consider a student deciding which college to attend

Wants a highly-ranked college, a small average class size, and maximum scholarship money

Plans to decide between pairs by **favouring the one does better on the most criteria**

College	National Ranking	Average Class Size	Scholarship Money Offered
X	4	40	\$3000
Y	8	18	\$1000
Z	12	24	\$8000



# Majority Rule: Other Ideas

## What about using majority rule another way?

Iterative approach: find a winner, remove from the list, and **recurse**

One idea: **arrange** alternatives in some order, then **compare** by majority vote, compare the winner to the third alternative, and so on.

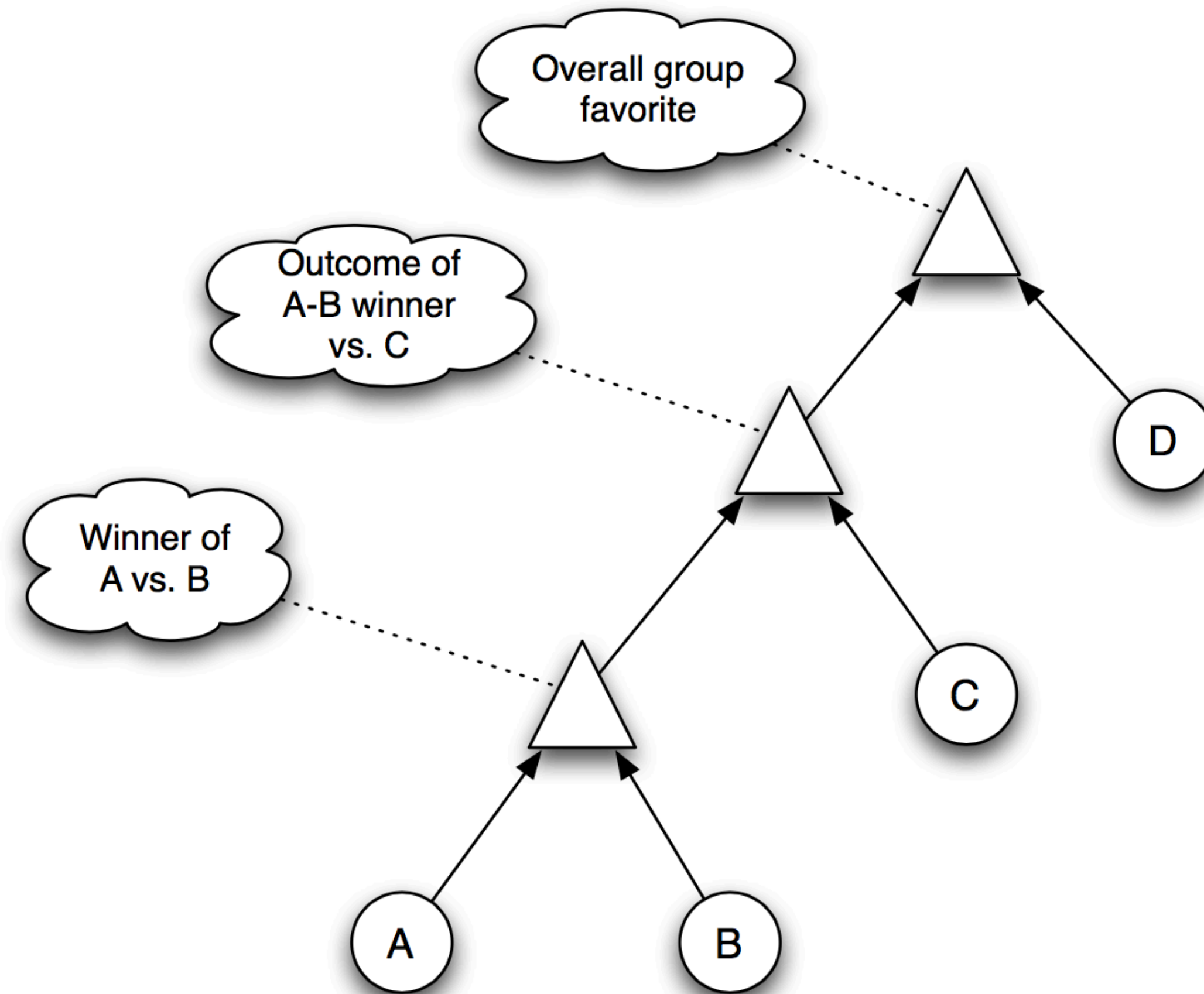
**Winner of the final comparison is the group favourite**

**More generally, we can **schedule any kind of elimination tournament** to determine the favourite**

→ Then recurse!

# Majority Rule: Other Ideas

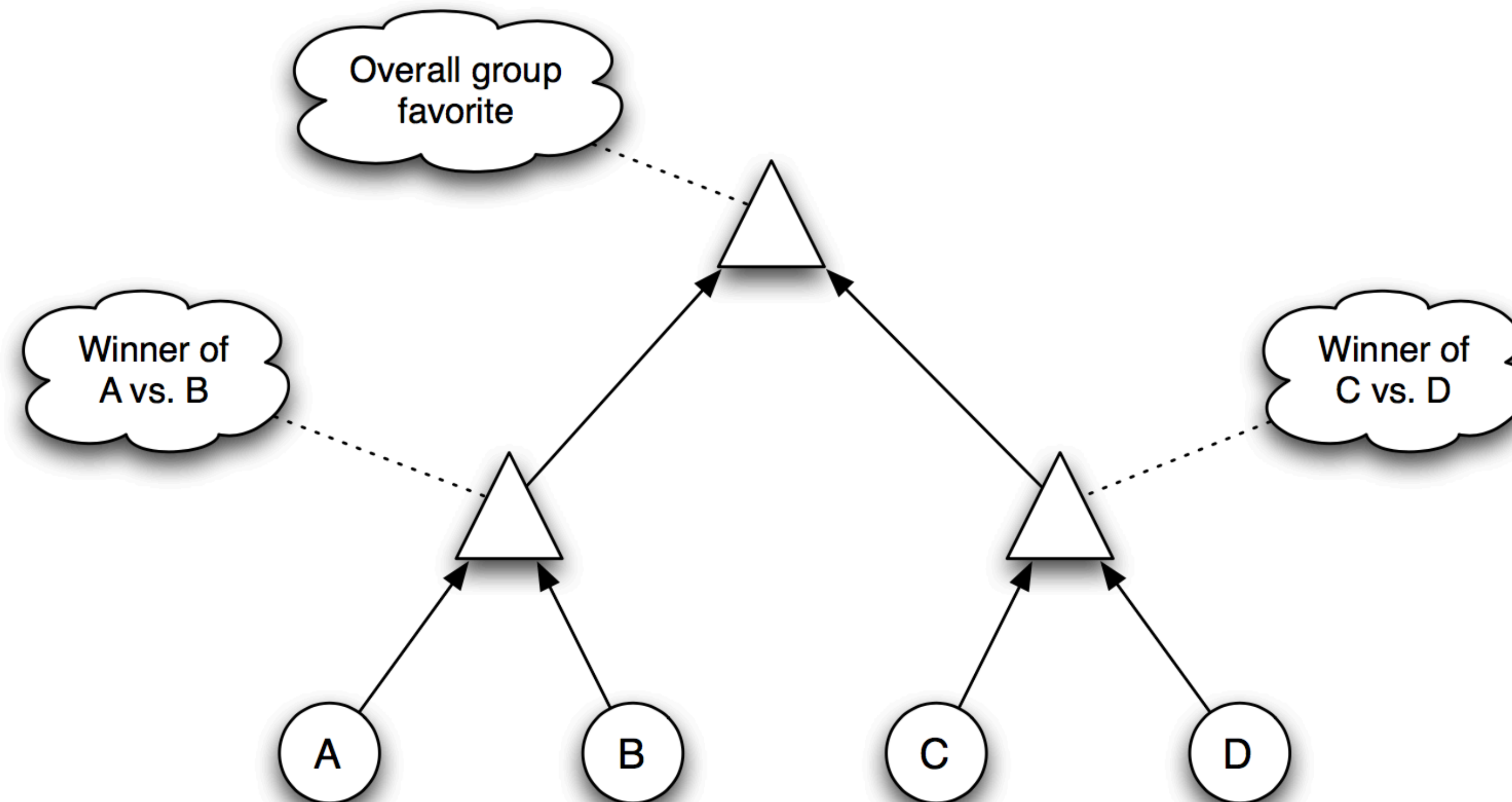
Graphically:





# Majority Rule: Other Ideas

**Other kind of elimination tournament:**

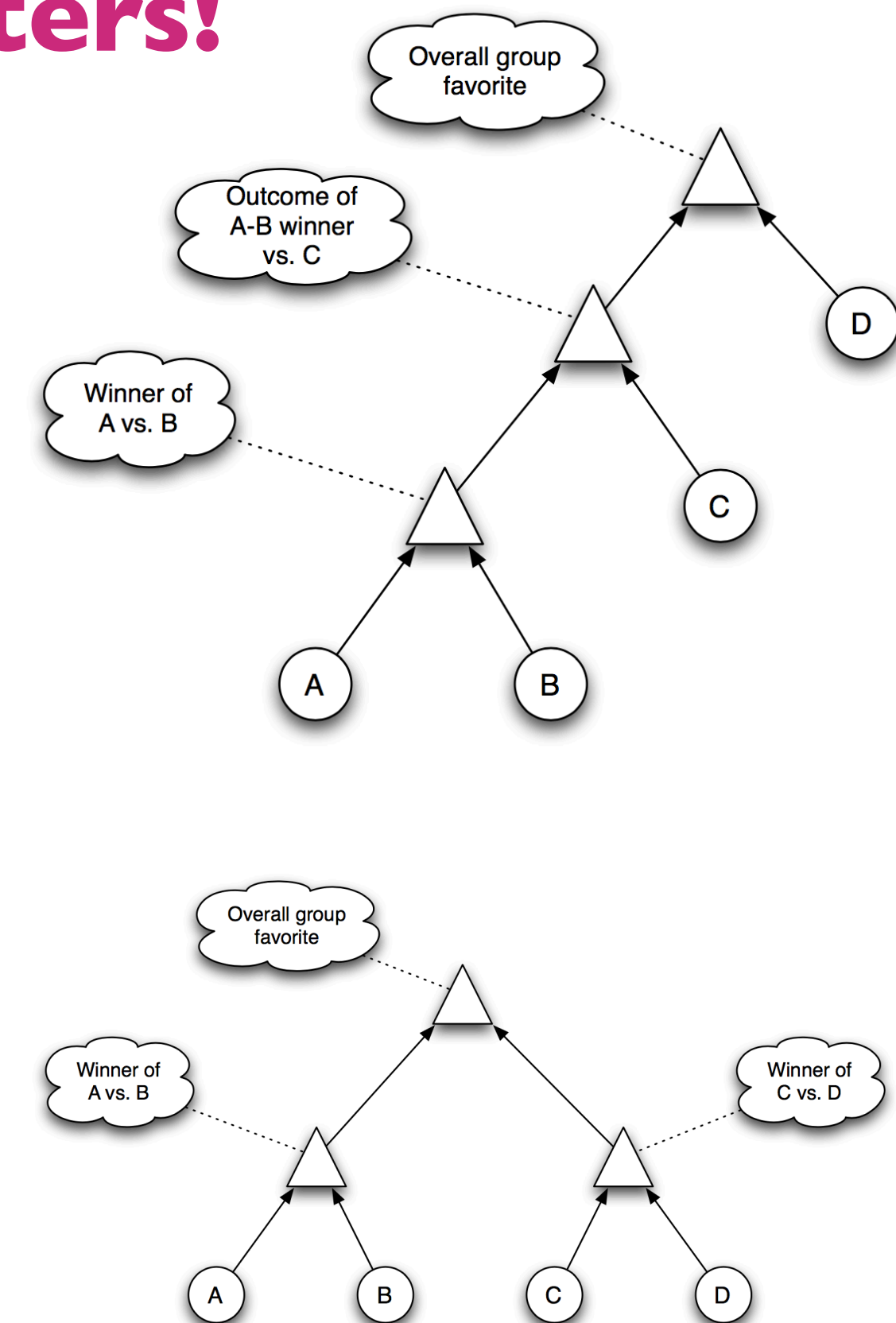
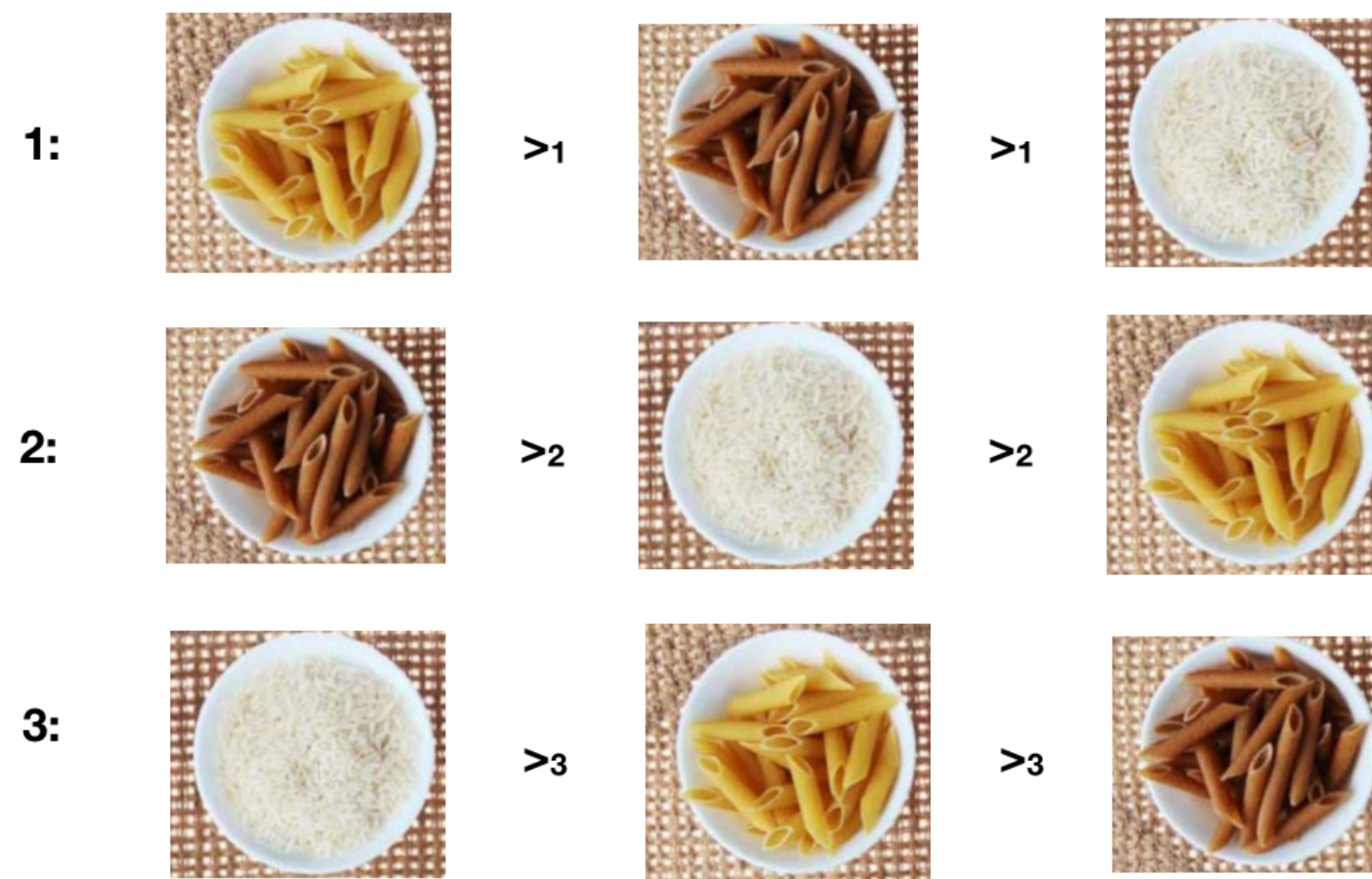


# Majority Rule: Other Ideas

What's wrong with this?

**Strategic agenda setting: order matters!**

Consider example from before:

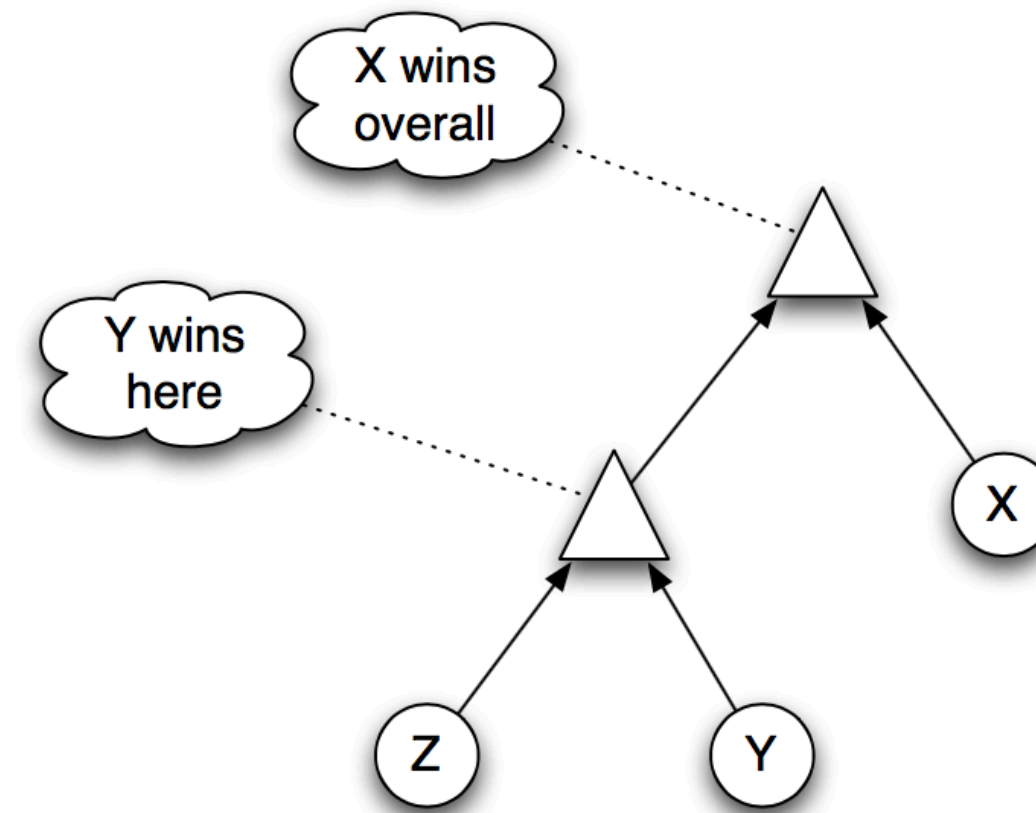
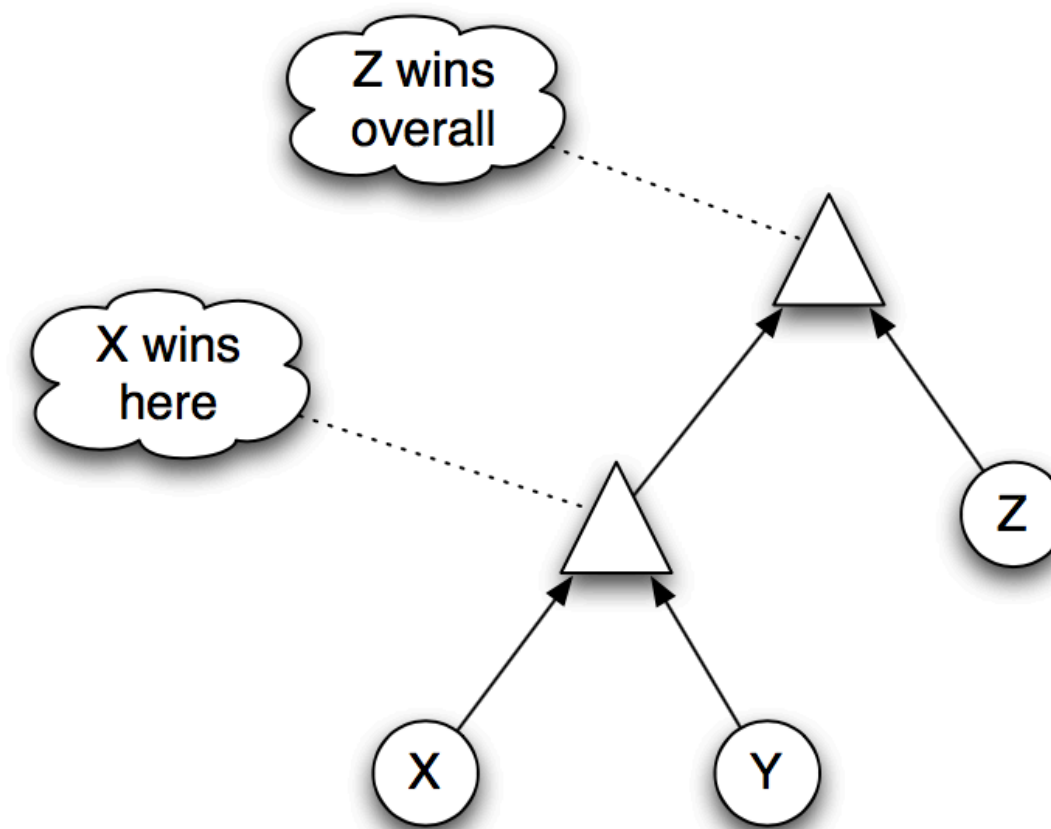


**In what order do we evaluate the alternatives?**

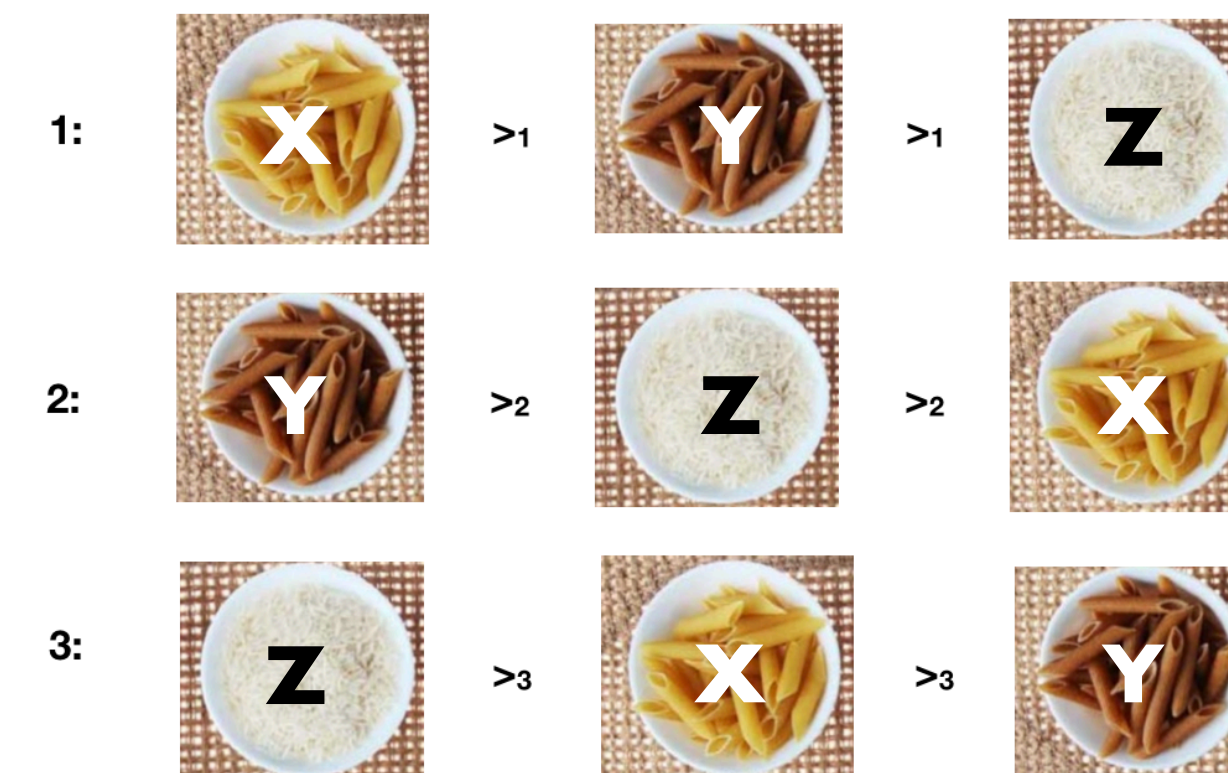


# Majority Rule: Other Ideas

In what order do we evaluate the alternatives?



Entire ranking is entirely determined by the order in which we evaluate!





# Other systems?

Majority rule led to some **bad outcomes**

What about other strategies?

**Positional voting:** produce a group ranking directly from the individual rankings

Forget pairwise comparisons

Each alternative receives a certain **weight** based on its positions in all the individual rankings

# Borda count

Heisman trophy in college football (and NBA MVP, etc.) all use the following method: get weight 0 for being picked last, 1 for being second last, ...,  $k-1$  for being picked first

**Repeat for each voter, tally up the scores, and rank**

**Example: two voters, four alternatives**

Voter 1:  $A >_1 B >_1 C >_1 D$

Voter 2:  $B >_2 C >_2 A >_2 D$

A:  $3 + 1 = 4$

B:  $2 + 3 = 5$

C:  $1 + 2 = 3$

D:  $0 + 0 = 0$

Group ranking:  $B > A > C > D$

**Called the “Borda Count”**



# Borda count

You can create your own variants (and many have) by changing the number of points per position

Example: if only top 3 matter, you could assign 3 for first place, 2 for second place, 1 for third place, and 0 otherwise

Any such system is a “**positional voting system**”

**Ignoring ties, Borda Count always produces a complete, transitive ranking!**





# Borda count

But the Borda Count **has its own problems**

Magazine tries to rank greatest movie of all time, asks five film critics to rank Citizen Kane and The Godfather

Three prefer CK, two prefer TG =>  $CK > TG$  => **all good!**

At the last second, they want to inject some modernity into the discussion, so they include **Frozen**

First three only like old movies, so they vote:

$CK >_i TG >_i F$

Critics 4 and 5 only like past 40 years, so:

$TG >_i F >_i CK$

What is the Borda Count now?



# Borda count

First three only like old movies, so they vote:

$CK >_i TG >_i F$

Critics 4 and 5 only like past 40 years, so:

$TG >_i F >_i CK$

Borda:

$CK: 6, TG: 7, F: 2 \Rightarrow TG > CK > F$

But before Frozen was introduced it was  $CK > TG$ !

**TG and CK flip because of Frozen??**

**Both TG and CK beat Frozen head-to-head**

Yet still Frozen influenced  $CK > TG$



# Borda count

Borda Count is susceptible to “irrelevant alternatives”

What voters think of Frozen **should be irrelevant** to how they feel about relative ranking of TG and CK

**But it isn't**

This gives rise to another problem: voters can **strategically misreport their preferences**

For example, say voters 4 and 5 actually had the true ranking  $TG > CK > F$

1,2,3:  $CK >_i TG >_i F$

4,5:  $TG >_i CK >_i F$

Borda:  $CK >_i TG >_i F$

By lying and reporting  $TG >_i F >_i CK$ , they get TG to win





# Irrelevant Alternatives in Politics

These problems with “irrelevant alternatives” and strategic misreporting have happened in elections around the world

Most vote with **plurality voting**: the candidate ranked at the top by most voters wins

Q: **is this a positional voting system?**

A: **Yes: 1 for winner, 0 otherwise**

“Third-party effects”/“spoiler effects”: if very few people favour some candidate, this can swing outcome of two leading contenders

In response, some people strategically misreport their preferences

# What's The Deal?

Voting is one society's **most important institutions**

On its face, seems like a relatively simple problem

**But we can't find a system that doesn't have horrible pathologies!**

**Is there any system that is free of pathologies?**

# What's The Deal?

Is there any system that is free of pathologies?

Let's define "Free of pathologies"

- Criterion 1 **"Unanimity"**: if there is a pair  $X$  and  $Y$  for which  $X >_i Y$  for every  $i$ , then  $X > Y$
- Criterion 2 **"Independence of Irrelevant Alternatives" (IIA)**: the ordering of  $X$  and  $Y$  should only depend on the relative positions  $X$  and  $Y$  in individual rankings

If we have a bunch of rankings that produces a group ranking with  $X > Y$

Then we move some  $Z$  around in the individual rankings

**It should still be the case that  $X > Y$**

- Criterion 3 **"Non-Dictatorship"**: the group ranking should not just always be what one particular voter thinks



# Independence of Irrelevant Alternatives



# Good Voting Systems

**What satisfies Unanimity and IIA and non-dictatorship?**

With two alternatives, majority rule clearly satisfies all

**Arrow's Theorem** [Arrow 1953]: With at least three alternatives, **no voting system** satisfies Unanimity, IIA, and Non-dictatorship

**In general, there is no good voting system!**

In practice, this means that there will always be **inherent tradeoffs we have to choose from**





# What Do We Do Now?

**How do we vote, how do we decide on things in the presence of Condorcet's Paradox and Arrow's Theorem?**

If you're faced with an impossibility result, you don't just give up

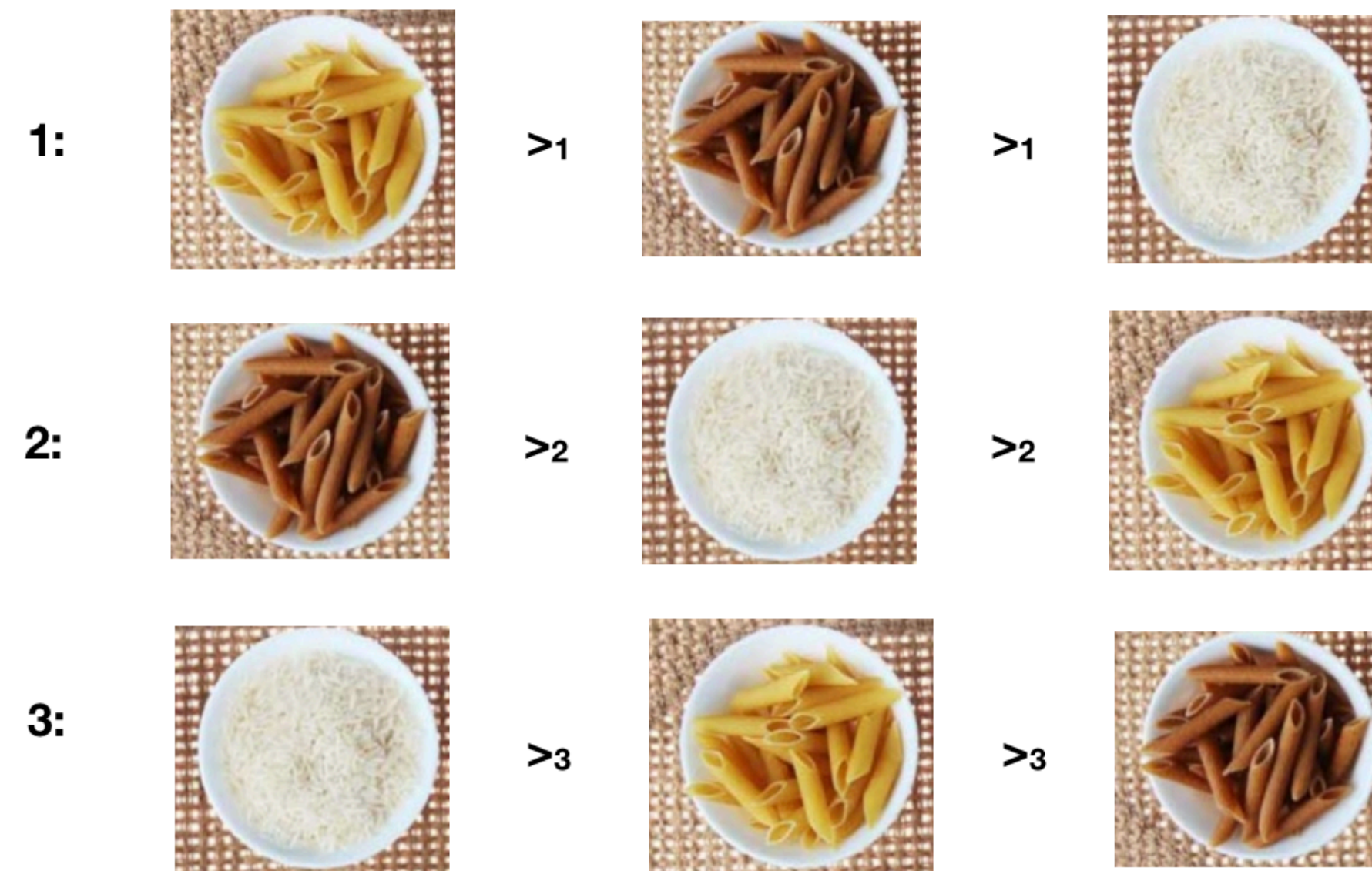
One common technique is to **look for important special cases**

Arrow's Theorem is a **general result**, so it doesn't necessarily apply if we **make some additional assumptions**



# What Do We Do Now?

Go back to original Condorcet problem



Replace food with choices about how much money to spend on education

# What Do We Do Now?

Go back to original Condorcet problem with money now:

1:  $X >_1 Y >_1 Z$

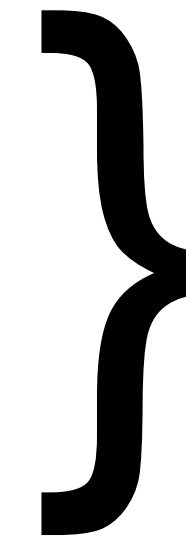
2:  $Y >_2 Z >_2 X$

3:  $Z >_3 X >_3 Y$

X: small

Y: medium

Z: a lot



Amount to  
spend on  
education

Voter 1's preferences "make sense"

Voter 2's preferences do too: prefer between Y and Z, so say Y then Z then X

Voter 3's preferences are **harder to justify**

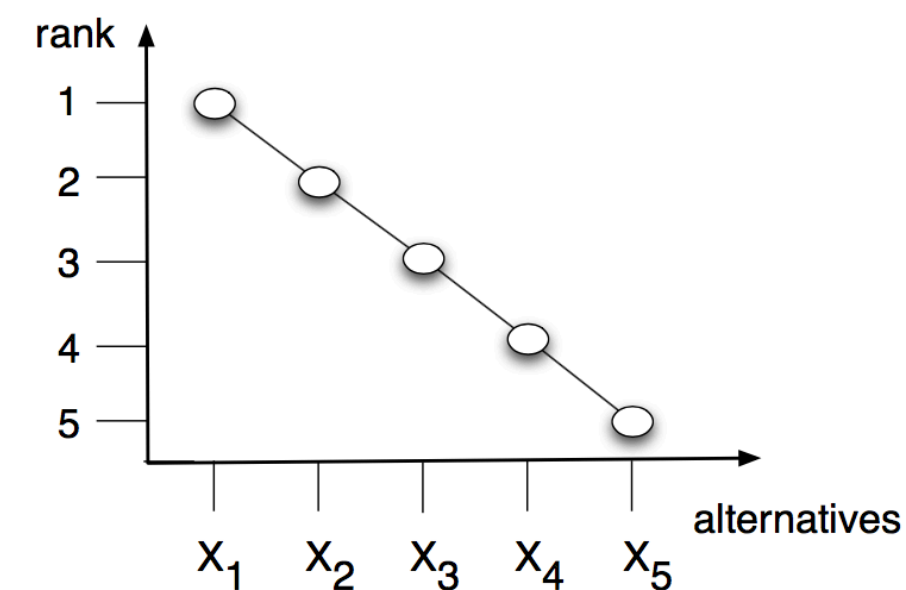
Not impossible, but they're more unusual

# Ideal Points

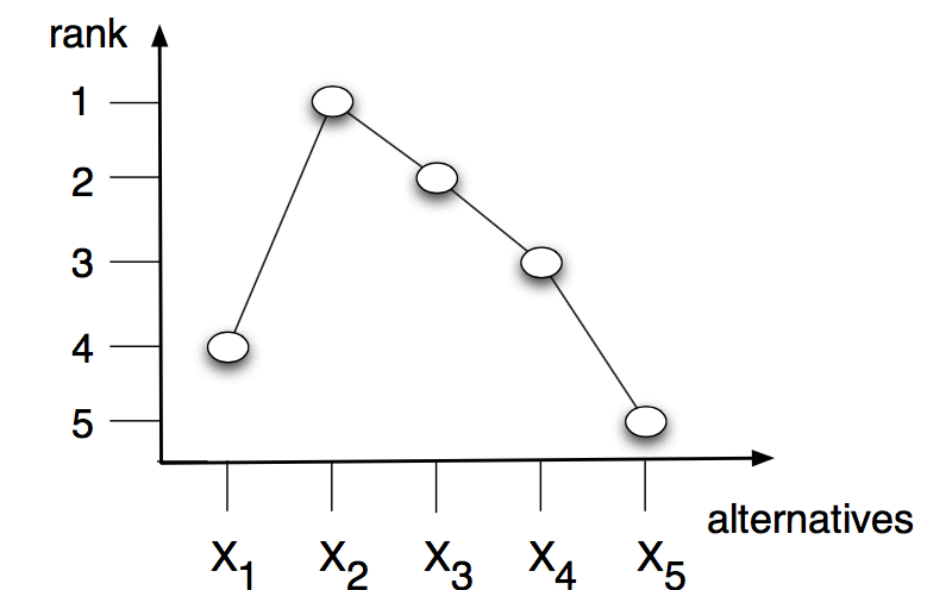
**Assume the preferences lie on a one-dimensional spectrum, and each voter has an “ideal point” on the spectrum**

They evaluate alternatives by proximity to this ideal point

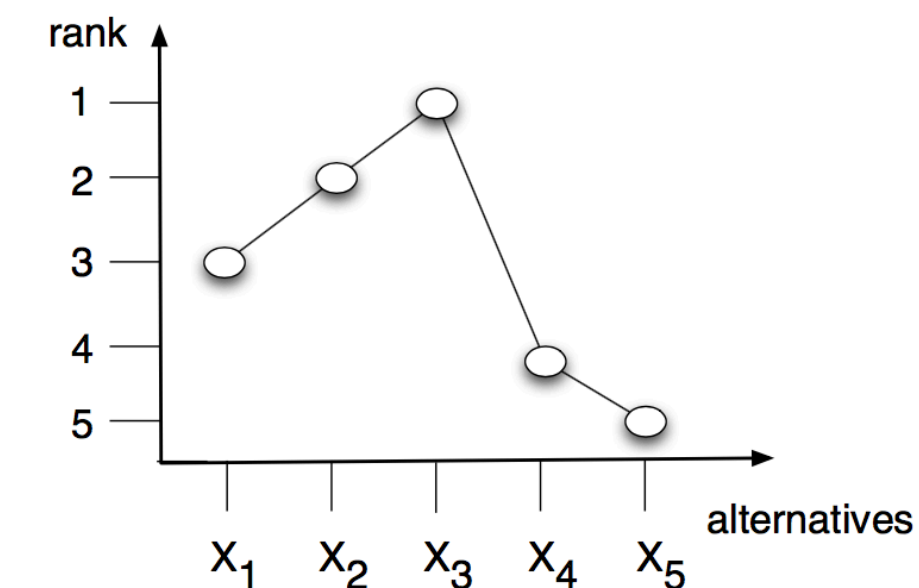
Actually we can assume something weaker: each voter's preferences “fall away” consistently on both sides of their favourite alternative



(a) Voter 1's ranking.



(b) Voter 2's ranking.



(c) Voter 3's ranking.

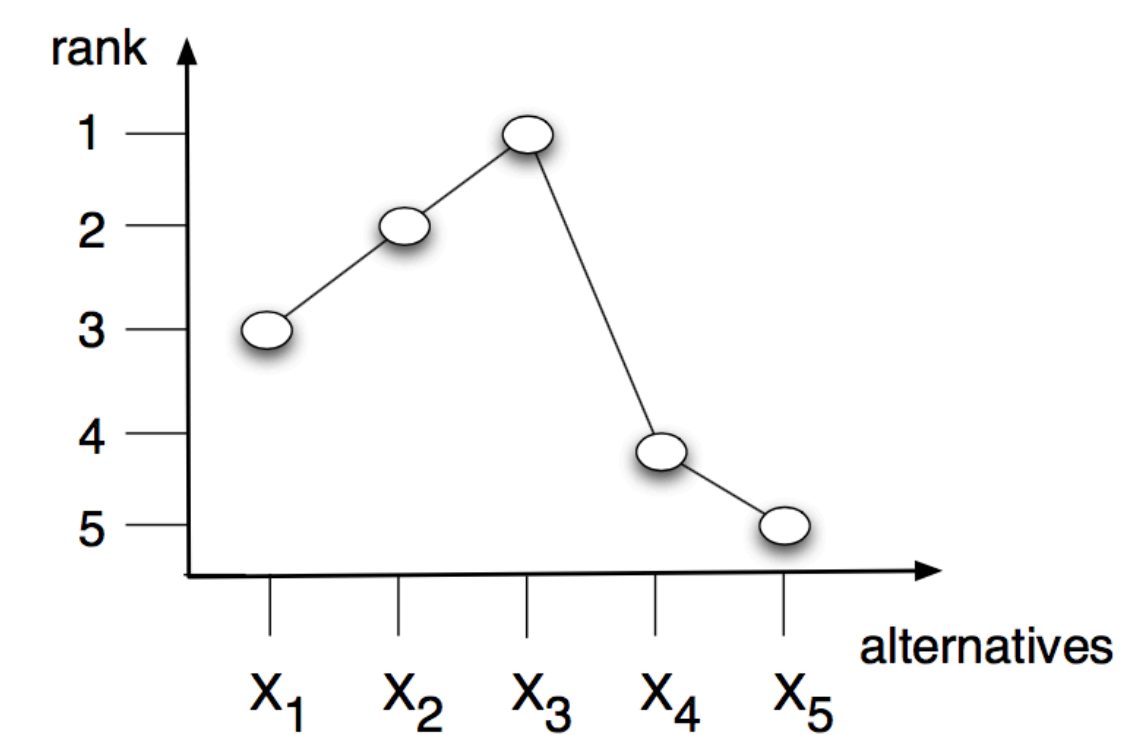
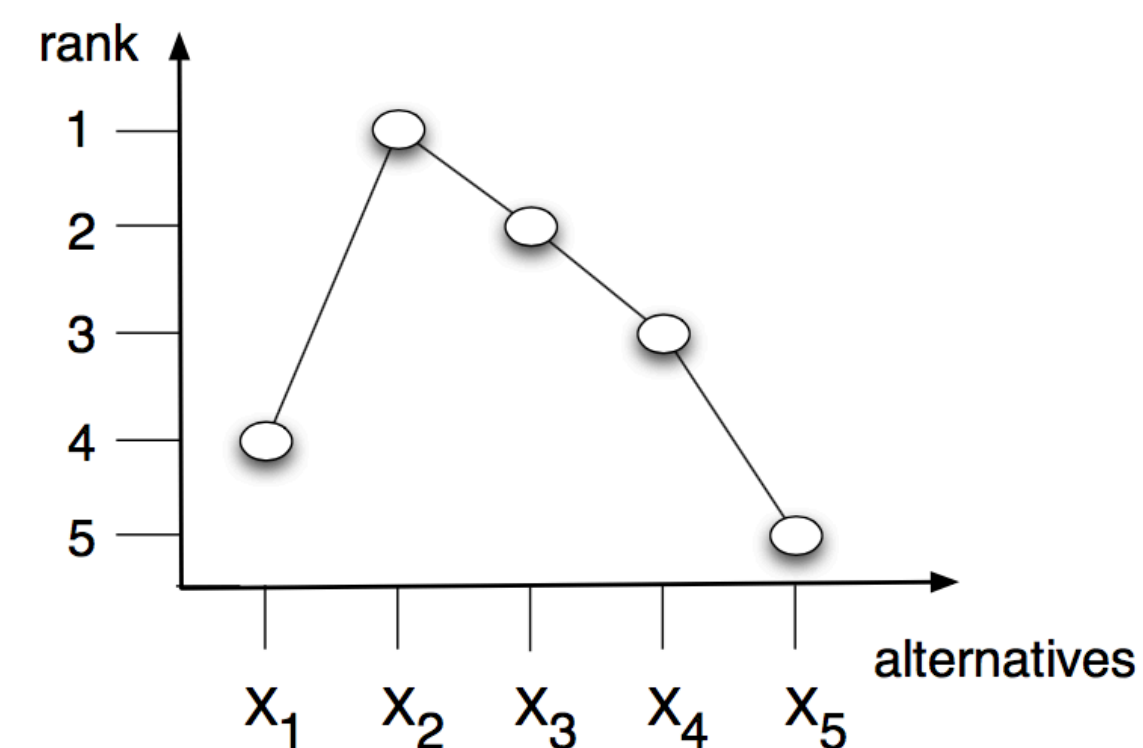
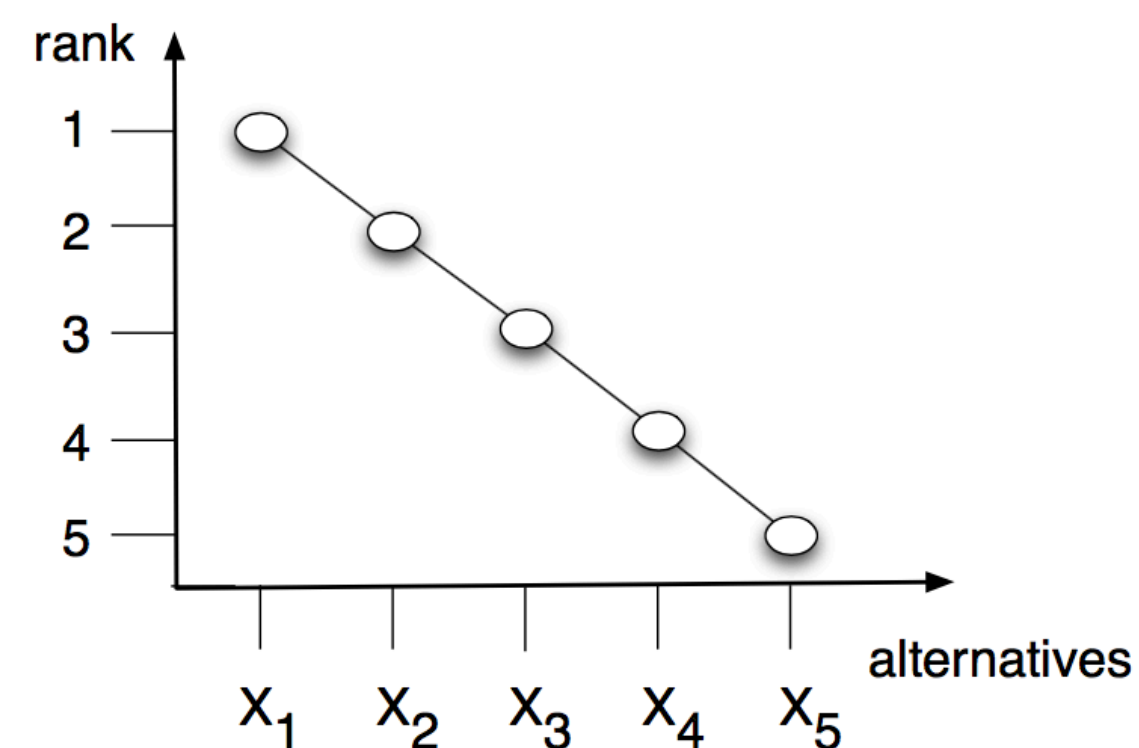


# Single-Peaked Preferences

**Definition:** a voter has “single-peaked preferences” if there is no alternative  $X_s$  for which both neighbouring alternatives  $X_{s-1}$  and  $X_{s+1}$  are ranked above  $X_s$

**Equivalent to:** every voter  $i$  has a top-ranked option  $X_t$ , and her preferences fall off on both sides of  $t$ :

$$X_t \succ_i X_{t+1} \succ_i X_{t+2} \succ_i \dots \quad \text{and} \quad X_t \succ_i X_{t-1} \succ_i X_{t-2} \succ_i \dots$$



# Single-Peaked Preferences

**Majority rule** with single-peaked preferences

Recall majority rule: compare every pair of alternatives  $X$  and  $Y$ , and decide  $X > Y$  or  $Y > X$  by the majority of voters

**Claim:** If all individual rankings are single-peaked, then majority rule applied to all pairs of alternatives produces a group preference relation that is **complete** and **transitive**.

In other words, **majority rule works!**

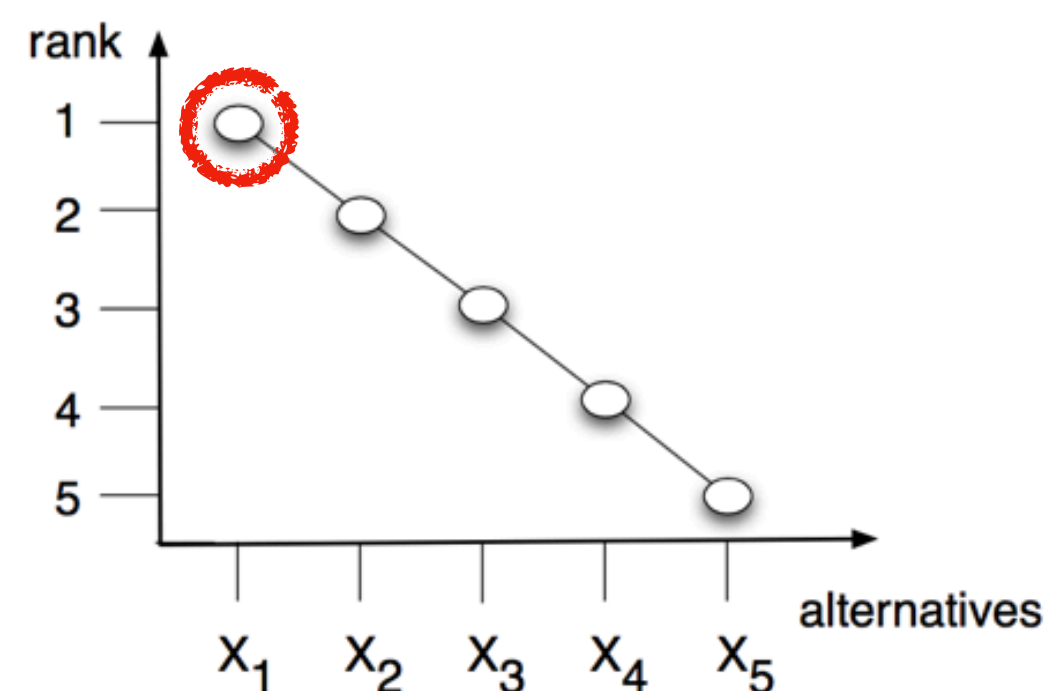
# Proof

Start off by trying to find a group favourite, then proceed by recursion on the rest of the alternatives

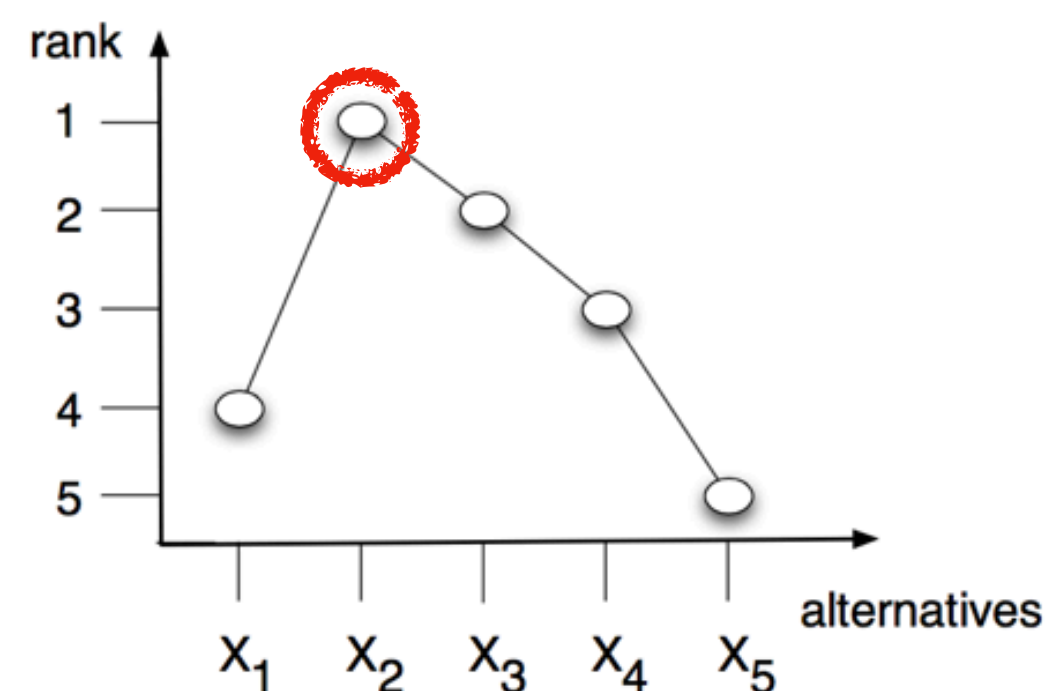
Consider **every voter's top-ranked alternative** — their peak — and **sort this set of favourites** from left to right along the spectrum

A popular alternative can show up many times

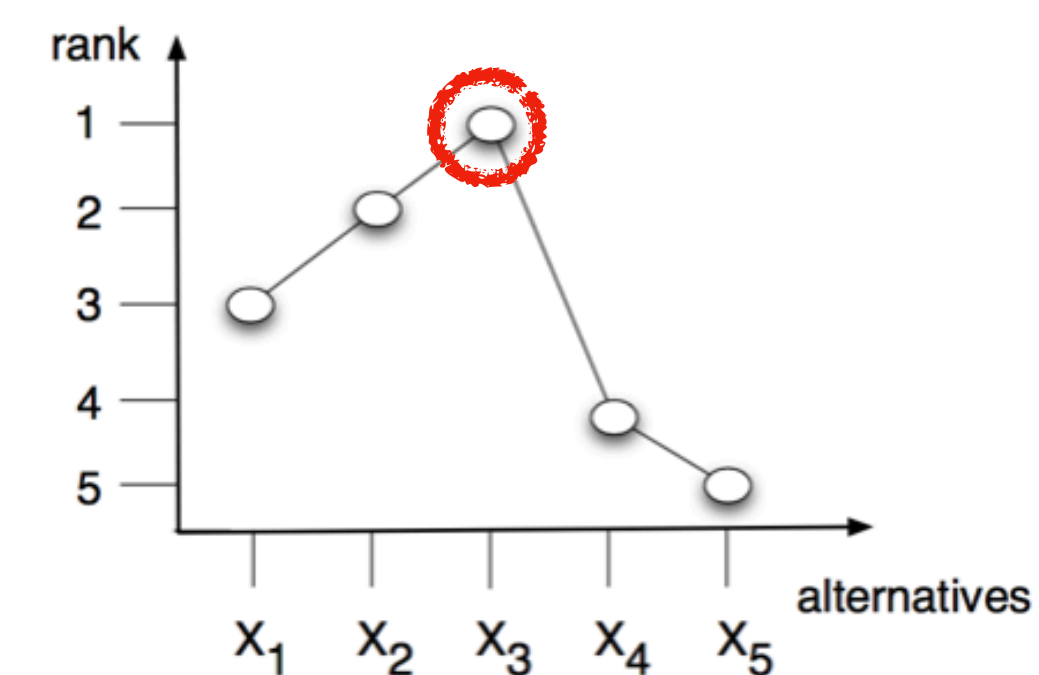
Now consider the **median** of these favourites



Favourites: X<sub>1</sub>, X<sub>2</sub>, X<sub>3</sub>



Median: X<sub>2</sub>





# Proof

The median individual favourite is a natural candidate for potential group favourite

Strikes a compromise between more extreme favourites on either side

**Median Voter Theorem:** With single-peaked preferences, the median individual favourite defeats every other alternative in a pairwise majority vote.

Let  $X_m$  be the median individual favourite, and  $X_t$  some random

Assume  $t > m$ , and sort the voters in order of their favourites

# Proof

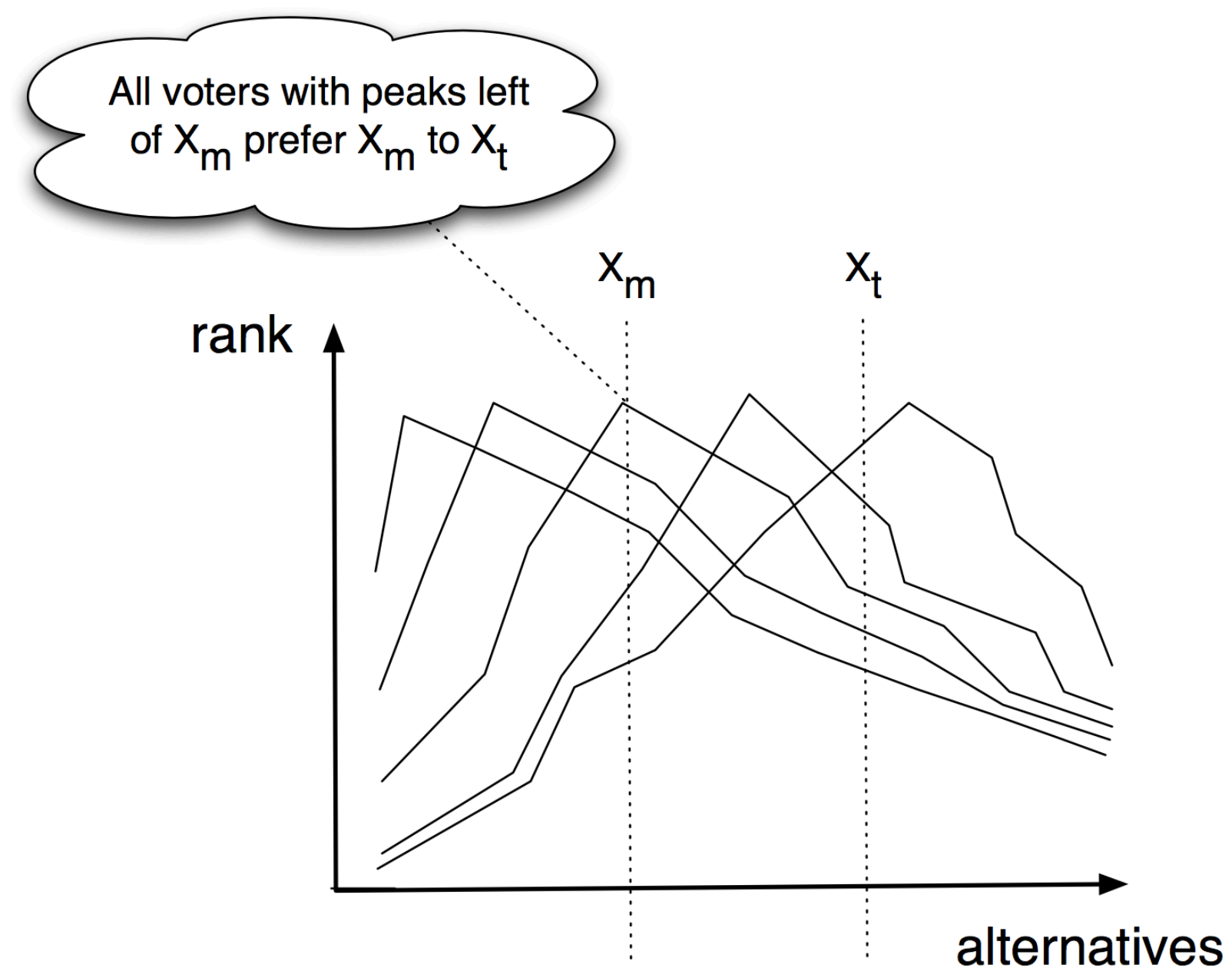
[remember  $k$  is odd]

By definition  $X_m$  is in position  $(k+1)/2$  of sorted favourites

In the first  $(k+1)/2$  positions, everyone either prefers  $X_m$  or their favourite is to the left of  $X_m$

For everyone in the latter group, both  $X_m$  and  $X_t$  on the right-hand “down-slope” of preferences, but  $X_m$  is higher than  $X_t$

**So  $(k+1)/2$  people prefer  $X_m$  to  $X_t$**  — which is a strict majority!



# Proof

So  $X_m$  can gather a majority of support against any other alternative  $X_t$  (case to the left of  $X_m$  is completely symmetric)

Build up group favourites one at a time: start by finding global favourite, then remove from individual rankings and recurse

Notice that removing it from the rankings still maintains the single-peaked property of everyone's preferences!

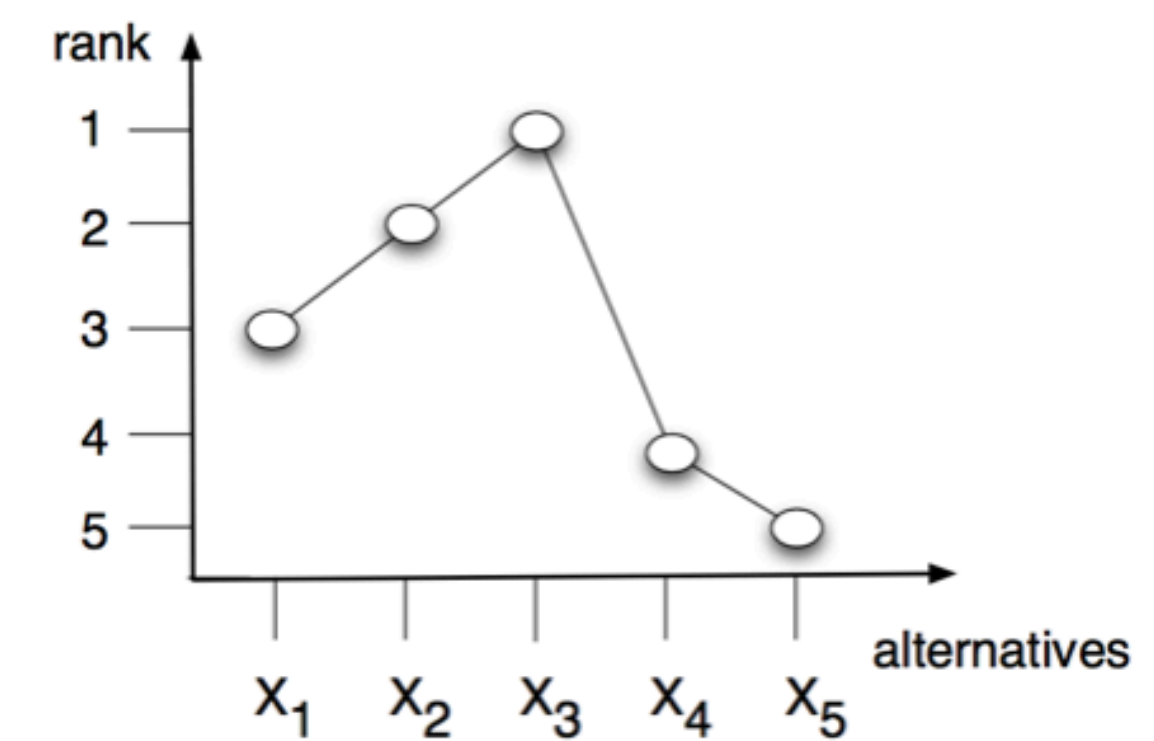
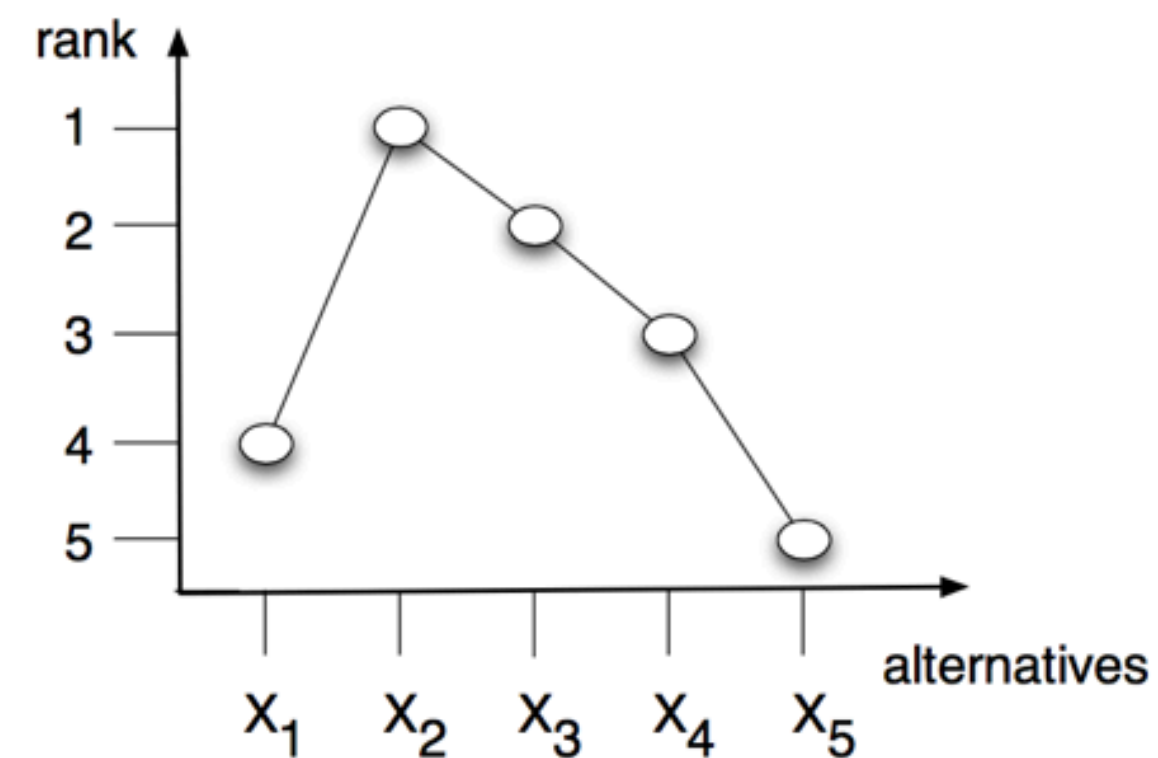
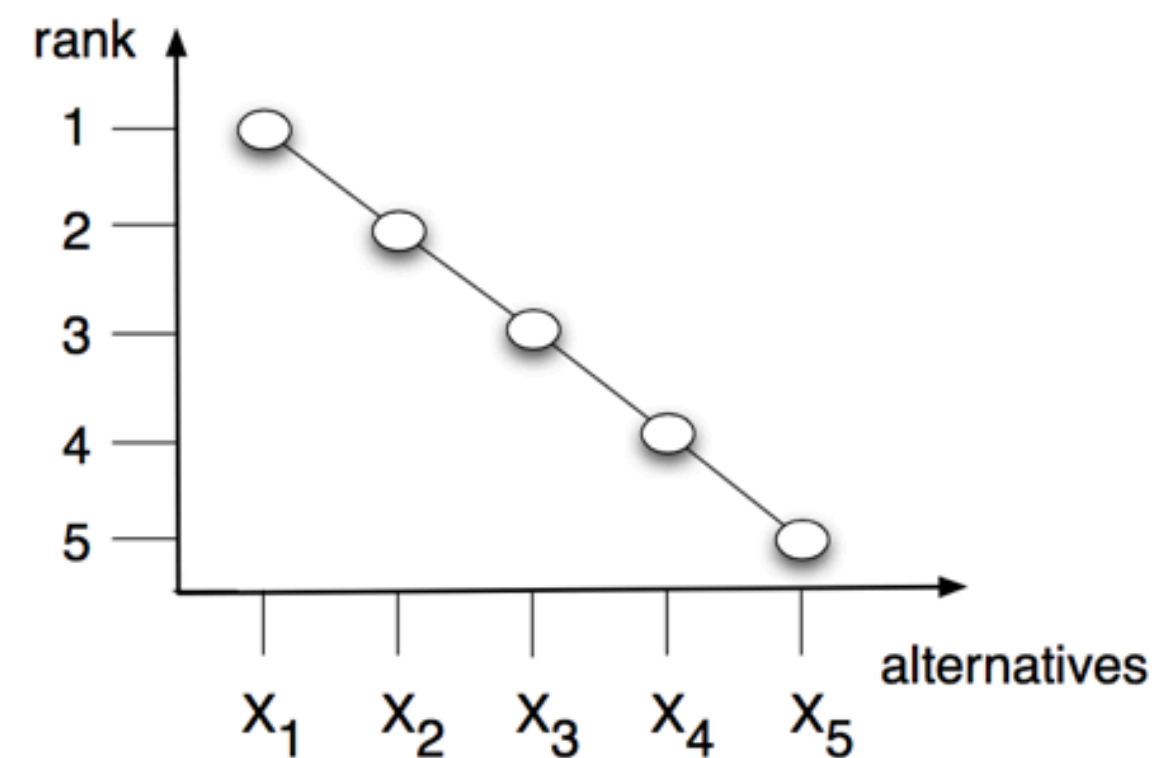
It just chops the peak off

# Example

$X_2$  is global median favourite

Then favourites are  $X_1, X_3, X_3 \Rightarrow X_3$  median favourite

Eventually we get  $X_2 > X_3 > X_1 > X_4 > X_5$





# Voting as Information Aggregation

So far, trying to come up with **methods for people who have different preferences**

**Sometimes there is a “true” underlying ranking** and people with different information are trying to uncover it

Examples:

Jury deliberation

Board of advisors to a company

# Simple Case: Simultaneous, Sincere Voting

Simple setting, two alternatives  $X$  and  $Y$

One is genuinely the best choice, each voter casts vote on what she thinks the right choice is

**Assume everyone votes sincerely**

**Model: similar to information cascades**

Prior probability that  $X$  is best is  $1/2$

Each voter gets a private independent signal on which is best, prob of getting right signal is  $q$  ( $> 1/2$ )

With probability  $q$ , voter should vote for what her signal says

**Condorcet Jury Theorem:** as the number of voters increases, probability of the majority choosing correct decision goes to 1

Oldest “wisdom of crowds” argument

# Simple Case: Simultaneous, Sincere Voting

Formal Bayes argument

Recall Bayes Rule:  $P[A|B] = P[B|A]P[A]/P[B]$

We want to compute  $P[X \text{ is best} | X\text{-signal}]$

Given:  $P[X \text{ is best}] = 1/2$  and  $P[X\text{-signal} | X \text{ is best}] = q$

Voter's strategy: evaluate  $P[X \text{ is best} | X\text{-signal}]$  then vote  $X$  if this probability  $> 1/2$

$P[X \text{ is best} | X\text{-signal}] = P[X\text{-signal} | X \text{ is best}]P[X \text{ is best}]/P[X\text{-signal}]$

$X$ -signal can be observed if  $X$  is best or if  $Y$  is best:

$P[X\text{-signal}] = P[X \text{ is best}] * P[X\text{-signal observed} | X \text{ is best}] + P[Y \text{ is best}] * P[X\text{-signal observed} | Y \text{ is best}]$

$P[X\text{-signal observed} | Y \text{ is best}] = 1/2q + 1/2(1-q) = 1/2$

So overall:  $P[X \text{ is best} | X\text{-signal}] = (1/2)q / (1/2) = q$

**Voter favours the alternative that is reinforced by her signal**

# Insincere Voting

We just assumed sincere voting

But there are **very natural situations** where a voter should actually **lie**, even though her goal is to **maximize the probability that the group chooses the right alternative!**

Example, information cascades-style:

Experimenter has two urns, 10 marbles each

One urn has 10 white marbles (“**pure**”) and the other has 9 green and one white (“**mixed**”)

Three people privately draw one marble and guess what urn it is, and all win money if the majority of them are right



# Insincere Voting

Suppose you draw a white marble

→ Way more likely that urn is **pure** than **mixed**

If you draw a green marble

→ Know for sure it's **mixed**

**But what should you guess?**

**First, when will your guess actually matter?**

If the two others agree, then **your guess doesn't change anything!**

Only case where it matters is if they're split

If they're split, someone said mixed, so they know it's mixed!

**Then you should guess mixed** to break the tie the right way!

Assuming others vote sincerely, you have an incentive to vote insincerely =>  
everyone voting sincerely is **not** a Nash equilibrium

# Insincere Voting

This is very naturally thought of as a game

Voters are **players**, guesses are **strategies**, and they result in certain **payoffs**

This is **highly stylized setting** so we can **see what's going on**

**But it happens in the real world too**

# Jury Deliberations

Consider a jury deliberating on a verdict: **guilty** or **innocent**

**There is a “best” answer — whether the defendant is actually guilty or innocent**

Compare with Condorcet Jury Theorem setup:

1. Juries require a **unanimous** vote. **Guilty** only if everyone says guilty
2. In Condorcet, evaluate alternatives just by picking most likely one (if  $> 1/2$  sure, pick it). Here, only pick guilty if sure beyond a reasonable doubt:

$$\Pr[\textit{defendant is guilty} \mid \textit{all available information}] > z \quad \textbf{for some large } z$$

# Jury Deliberations

Each juror gets an independent private signal: **guilty signal** (G-signal) or **innocent signal** (I-signal)

**They usually get the right signal:**  $P[\text{G-signal} \mid \text{defendant guilty}] = P[\text{I-signal} \mid \text{defendant innocent}] = q, q > 1/2$

Assume prior probability of guilt of  $1/2$ , but doesn't matter

**What should a juror do?**



# Jury Deliberations

- **What should a juror do?**
- Say you receive an I-signal
  - At first it seems obvious that you should vote to acquit
  - But: conviction criterion is  $\Pr[\text{defendant is guilty} \mid \text{available information}] > z$  so **if all the other jurors received G-signals you might still be above that threshold**
  - Second, ask yourself key question from before: **when does my vote actually matter?**
  - **Like before, your vote only changes the outcome if everyone except you is voting guilty!**
    - **If you vote guilty, defendant is found guilty**
    - **If you vote to acquit, defendant is found innocent**

# Jury Deliberations

- **If everyone but you is voting guilty, what is the probability of defendant being guilty?**

$$\begin{aligned} & \Pr[\text{defendant is guilty} \mid \text{you have the only } I\text{-signal}] \\ &= \frac{\Pr[\text{defendant is guilty}] \cdot \Pr[\text{you have the only } I\text{-signal} \mid \text{defendant is guilty}]}{\Pr[\text{you have the only } I\text{-signal}]}. \end{aligned}$$

$$\begin{aligned} & \Pr[\text{you have the only } I\text{-signal}] \\ &= \Pr[\text{defendant is guilty}] \cdot \Pr[\text{you have the only } I\text{-signal} \mid \text{defendant is guilty}] + \\ & \quad \Pr[\text{defendant is innocent}] \cdot \Pr[\text{you have the only } I\text{-signal} \mid \text{defendant is innocent}] \\ &= \frac{1}{2} \cdot q^{k-1}(1-q) + \frac{1}{2}(1-q)^{k-1}q. \end{aligned}$$

# Jury Deliberations

- If everyone but you is voting guilty, what is the probability of defendant being guilty?

$$\begin{aligned} & \Pr[\text{defendant is guilty} \mid \text{you have the only I-signal}] \\ &= \frac{\Pr[\text{defendant is guilty}] \cdot \Pr[\text{you have the only I-signal} \mid \text{defendant is guilty}]}{\Pr[\text{you have the only I-signal}]} \end{aligned}$$

$$\begin{aligned} \Pr[\text{defendant is guilty} \mid \text{you have the only I-signal}] &= \frac{\frac{1}{2}q^{k-1}(1-q)}{\frac{1}{2}q^{k-1}(1-q) + \frac{1}{2}(1-q)^{k-1}q} \\ &= \frac{q^{k-2}}{q^{k-2} + (1-q)^{k-2}}, \end{aligned}$$

- Since  $q > 1/2$ ,  $(1-q)^{k-2}$  is super small, so the probability goes to 1
- **In only case where your vote to acquit matters, you should vote guilty despite your I-signal!**

# Jury Deliberations

- Intuitively: because of the unanimity rule, you only affect the outcome when everyone else holds the opposite opinion
- Assuming everyone else is as informed as you, and **assuming independence** (remember information cascades!), then the conclusion is that they're probably collectively right
- The result is: assuming everyone else votes **sincerely**, you have an incentive to vote **insincerely**
  - **All-sincere voting is not an equilibrium**
- **What is the equilibrium?**
  - There are several
  - Most interesting is a mixed equilibrium (randomly disregard I-signal some fraction of the time to correct for possibility that it's wrong)
  - **In this equilibrium, probability of convicting an innocent defendant does not go to zero as #jurors goes to infinity!**



# Jury Decisions

- Why do we get such a bad outcome?
- **Unanimity is a very harsh constraint.**
  - If we relax to only requiring a certain fraction  $f$  saying guilty, then the probability that we convict an innocent defendant goes to 0

# Summary

- **Voting**: synthesizing the preferences of many people into a single group preference
- Many fundamental issues:
  - **Condorcet paradox**: most natural method (majority rule) can turn a set of reasonable preference relations into an unreasonable one
  - **Arrow's Theorem**: **no general voting system** simultaneously satisfies unanimity, IIA, and non-dictatorship.
- Special case: single-peaked preferences
  - Median Voter Theorem says we can get good outcomes
- Jury deliberations: insincere voting can be incentivized



# Arrow's Theorem

- “Voting profile” = set of  $k$  individual rankings
- Criterion 1 **“Unanimity”**: if there is a pair  $X$  and  $Y$  for which  $X >_i Y$  **for every  $i$** , then  $X > Y$
- Criterion 2 **“Independence of Irrelevant Alternatives” (IIA)**: the ordering of  $X$  and  $Y$  should **only depend on the relative positions  $X$  and  $Y$  in individual rankings**
  - If we have a bunch of rankings that produces a group ranking with  $X > Y$
  - Then we move some  $Z$  around in the individual rankings
  - **It should still be the case that  $X > Y$**
  - **Another way to say it**: voting profile restricted to  $X$  and  $Y$  deletes everything except  $X$  and  $Y$ . Voting system should order  $X$  and  $Y$  same way on profile as on profile restricted to  $X$  and  $Y$
- Criterion 3 **“Non-Dictatorship”**: the group ranking should not just always be what one particular voter thinks