

Heterogeneity, contact patterns and modeling options

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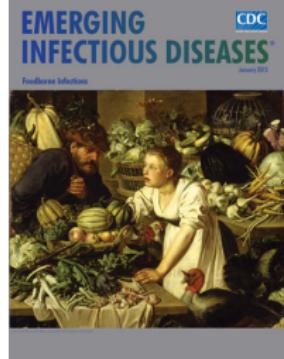
MMED 2017

Goals

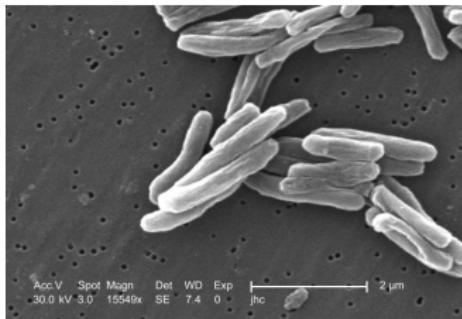
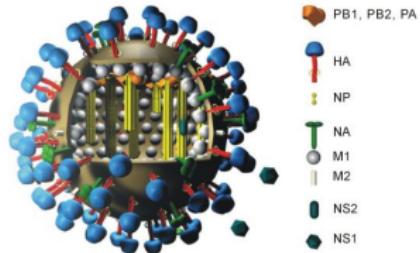
- ▶ Explain the importance of heterogeneity on patterns of disease spread
 - ▶ Focus on different types of human heterogeneity
- ▶ Discuss ways in which homogeneous models fail to match observed dynamics
- ▶ Use simple models to explore qualitative effects of heterogeneity on modeling conclusions
- ▶ Briefly introduce some methods that are used to incorporate heterogeneity in models

The resilience of infectious disease

1967: It's time to close the book on infectious diseases



Pathogen evolution



Human heterogeneity



Human heterogeneity



Human heterogeneity



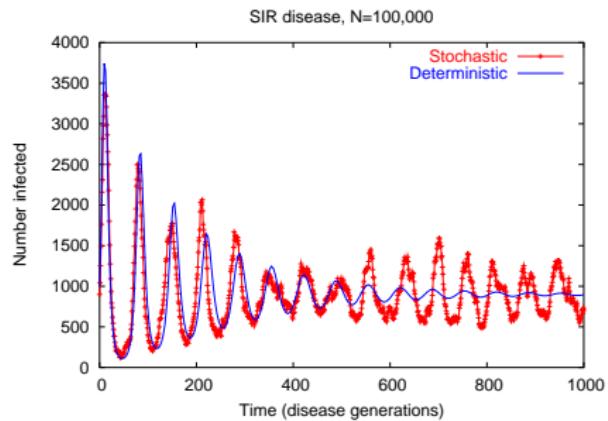
Expanding our models

- ▶ **Homogeneous** models assume everyone has the same:
 - ▶ disease characteristics (e.g. susceptibility, tendency to transmit)
 - ▶ mixing rate
 - ▶ probability of mixing with each person
- ▶ **Heterogeneous** models allow people to be different

The basic reproductive number

- ▶ \mathcal{R}_0 is the number of people who would be infected by an infectious individual *in a fully susceptible population*.
- ▶ $\mathcal{R}_0 = \beta/\gamma = \beta D = (cp)D$
 - ▶ c : Contact Rate
 - ▶ p : Probability of transmission (infectivity)
 - ▶ D : Average duration of infection
- ▶ A disease can invade a population if and only if $\mathcal{R}_0 > 1$.

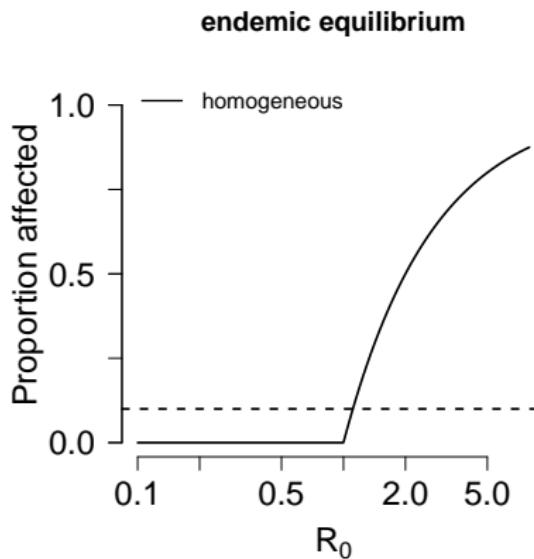
Equilibrium



Equilibrium analysis

- ▶ \mathcal{R}_{eff} is the number of people who would be infected by an infectious individual *in a general population*.
- ▶
$$\mathcal{R}_{\text{eff}} = \mathcal{R}_0 \frac{S}{N} = pcD \frac{S}{N}$$
- ▶ At equilibrium:
$$\mathcal{R}_{\text{eff}} = \mathcal{R}_0 \frac{S}{N} = 1.$$
- ▶ Thus:
$$\frac{S}{N} = 1/R_0.$$
- ▶ Proportion ‘affected’ is $V = 1 - S/N = 1 - 1/R_0.$

Homogeneous endemic curve

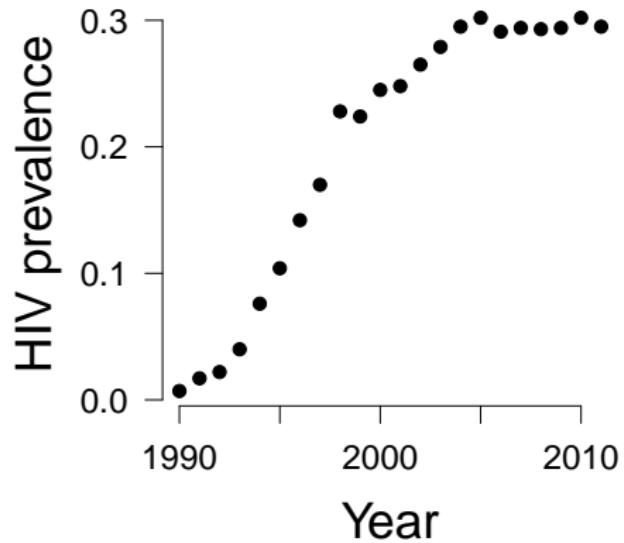


- ▶ Threshold value
- ▶ Sharp response to changes in factors underlying transmission
- ▶ Works – sometimes

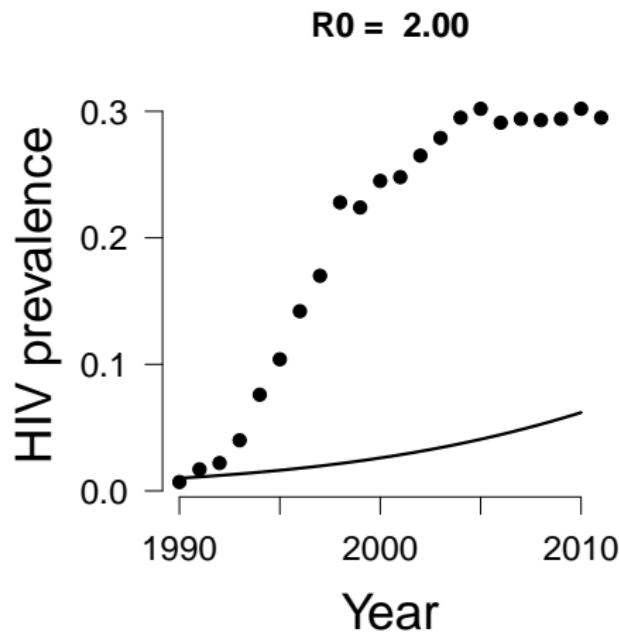
Proportion affected and disease prevalence

- ▶ For diseases with no recovery, V is the disease prevalence
- ▶ For other diseases, the equilibrium value of $P = I/N$ will be equal to V times the ratio of time spent sick to the time spent immune.
- ▶ Example: measles before vaccination
 - ▶ $V = 0.95$
 - ▶ $\bar{P} = 0.95 \times (2\text{wk}/60\text{yr})$.

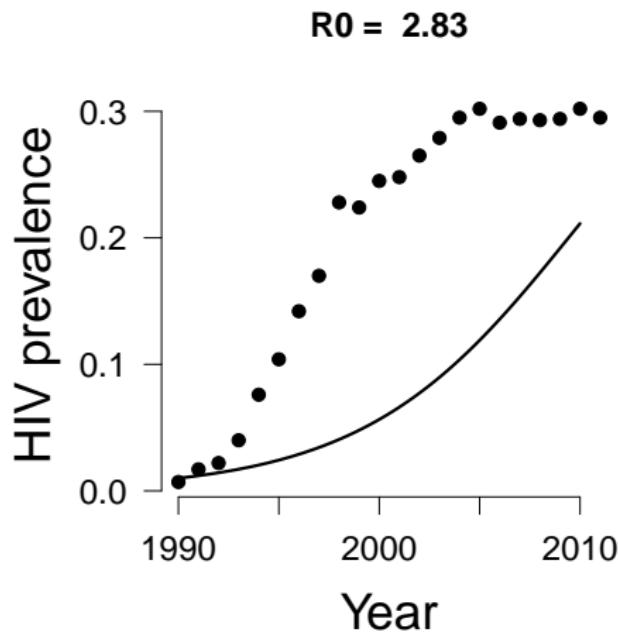
Disease dynamics



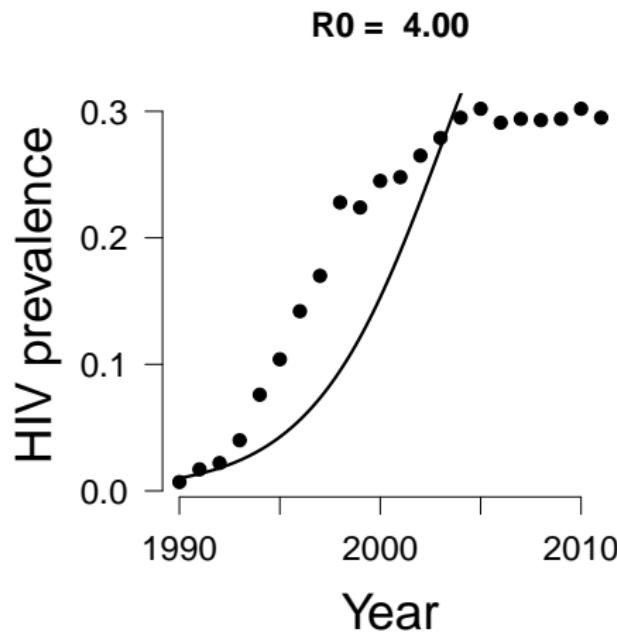
Homogeneous assumptions



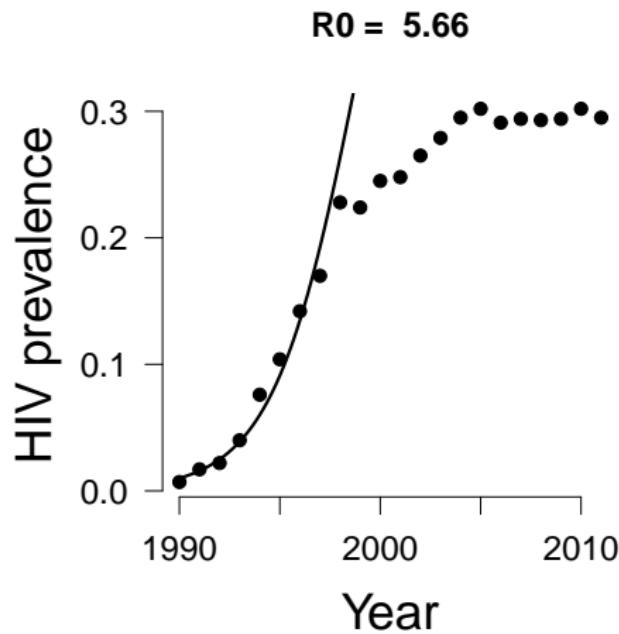
Homogeneous assumptions



Homogeneous assumptions



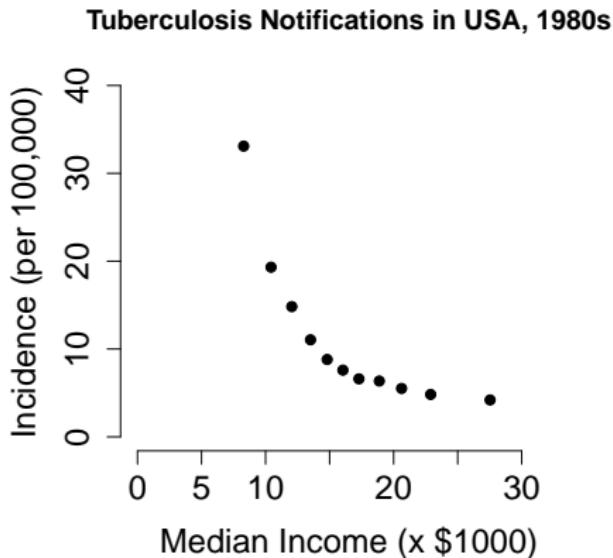
Homogeneous assumptions



Beyond homogeneity

- ▶ Flavors of heterogeneity
 - ▶ among hosts
 - ▶ spatial
 - ▶ demographic (discreteness of individuals)
 - ▶ temporal
 - ▶ others

Heterogeneity in TB



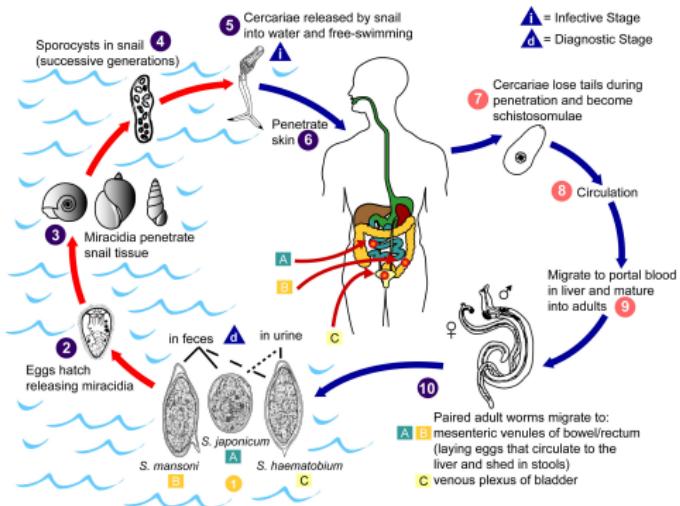
- ▶ **Progression:**
Nutrition, stress
- ▶ **Contact:**
Overcrowding, poor ventilation
- ▶ **Cure:** Access to medical care

Heterogeneity in other diseases

- ▶ **STDs:** Sexual mixing patterns, access to medical care
- ▶ **Influenza:** Crowding, nutrition
- ▶ **Malaria:** Attractiveness to biting insects, geographical location, immune status
- ▶ **Every disease!**

Large-scale heterogeneity

Schistosomiasis



- For schistosomiasis, the worldwide average $\mathcal{R}_0 < 1$
- Disease persists because of specific populations with $\mathcal{R}_0 > 1$.
- This effect operates at many scales.

Equilibrium calculations

- ▶ Assume $p = \sigma\tau$ has a susceptibility component and a transmission component:
 - ▶ $\mathcal{R}_0 = \sigma\tau cD$
 - ▶ $\mathcal{R}_{\text{eff}} = \sigma\tau cDS/N$

Equilibrium calculations with heterogeneity

- ▶ τD applies to infectious individuals $\rightarrow \tau_I D_I$
- ▶ σ applies to susceptible individuals $\rightarrow \sigma_S$
- ▶ c is complicated $\rightarrow c_S c_I / \bar{c}$

Example

- ▶ Imagine a disease spread by people who differ only in their effective mixing rates
- ▶ If the disease has just started spreading in the population, how do c_S and c_I compare to \bar{c} ?
 - ▶ $c_S \approx \bar{c}; c_I > \bar{c}$.
- ▶ If the disease is very widespread in the population?
 - ▶ $c_S < \bar{c}; c_I \rightarrow \bar{c}$.

Simpson's paradox



- ▶ What happens when a peanut farmer is elected to the US Senate?
- ▶ The average IQ goes up in both places!

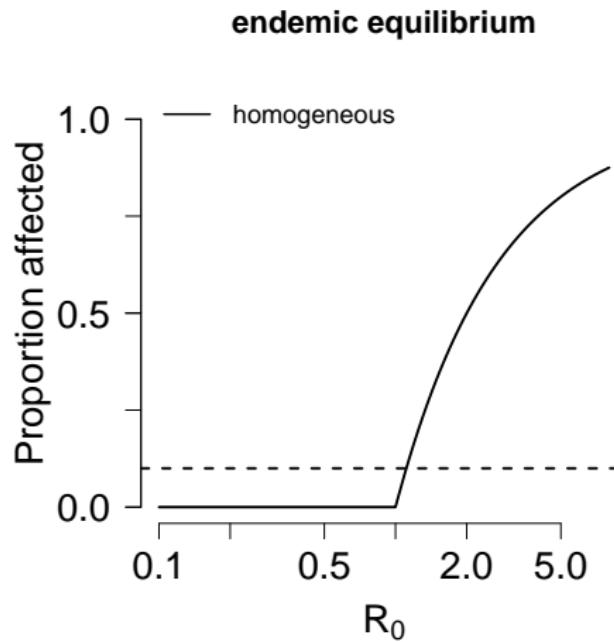
The basic reproductive number

- ▶ When the disease invades:
 - ▶ The susceptible population \approx the general population
 - ▶ The infectious population is likely to have higher values of c , D and/or τ
- ▶ \mathcal{R}_0 is typically greater than you would expect from a homogeneous model

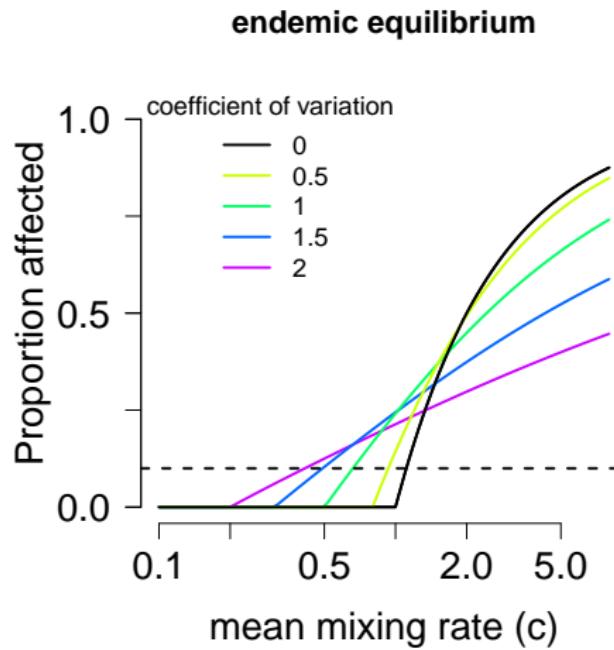
Equilibrium analysis

- ▶ As disease prevalence goes up:
 - ▶ Susceptible pool is the most resistant, or least exposed group
 - ▶ Infectious pool moves looks more like the general population.
- ▶ → lower proportion affected *for a given value of \mathcal{R}_0 .*

Homogeneous endemic curve

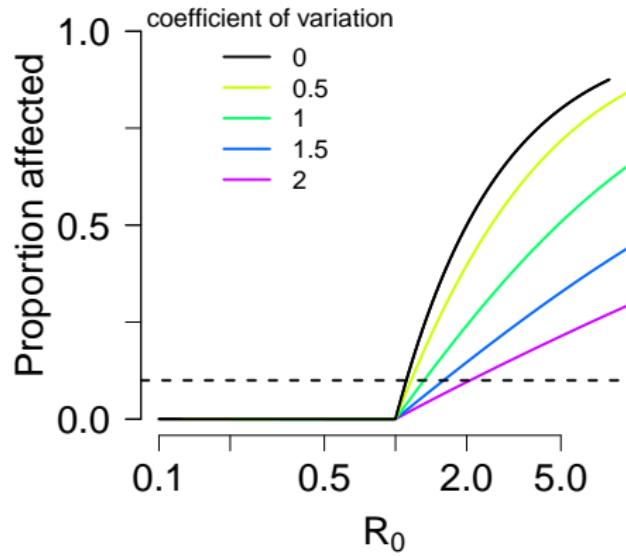


Heterogeneous endemic curves



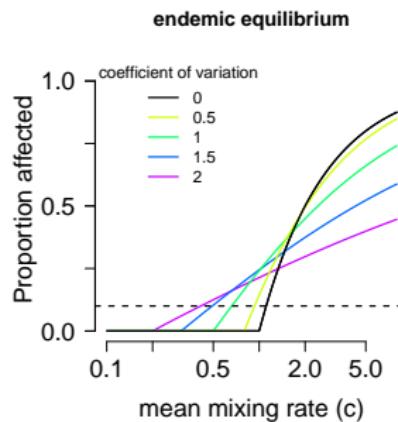
Heterogeneous endemic curves

endemic equilibrium



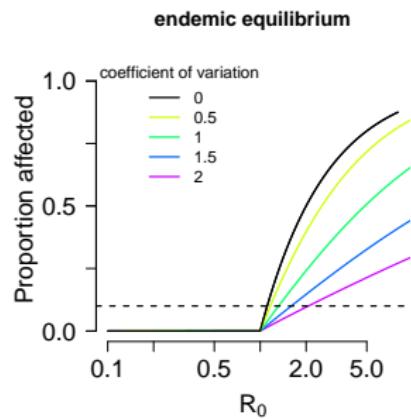
Heterogeneity and disease

- ▶ Heterogeneity has a double-edged effect
 - ▶ Effects of disease are *lower* for a given value of \mathcal{R}_0 .
 - ▶ But \mathcal{R}_0 is *higher* for given mean values of factors underlying transmission



Heterogeneous endemic curves

- ▶ Heterogeneity makes the endemic curve flatter
- ▶ Disease levels are more resistant to change



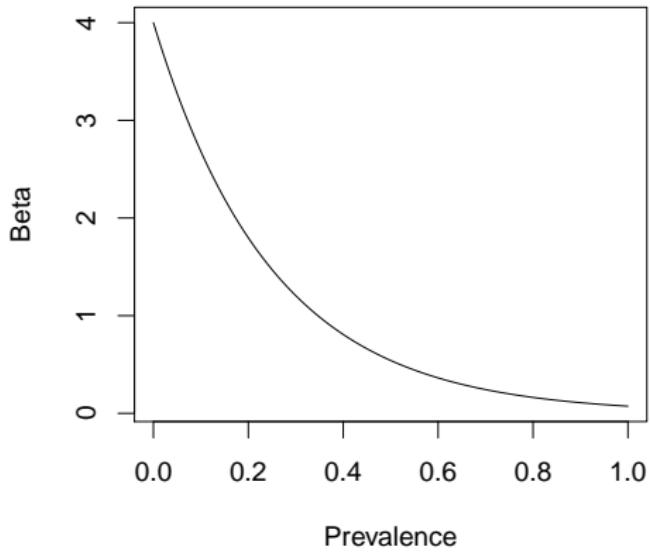
How diseases reach equilibrium

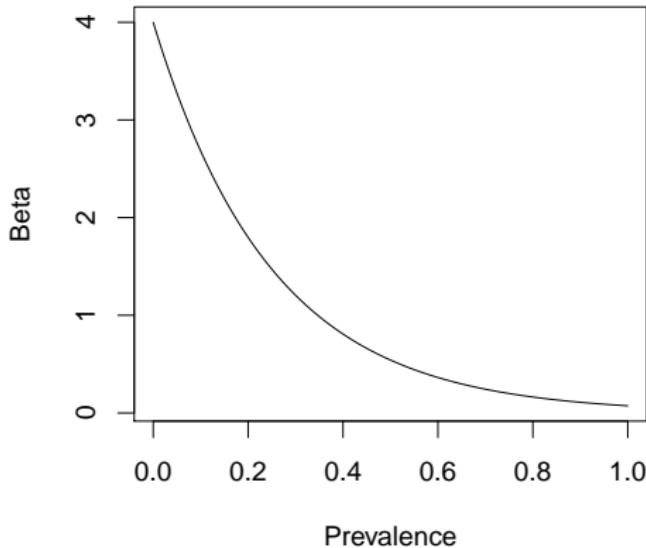
- ▶ Diseases that invade have high values of \mathcal{R}_0
- ▶ \mathcal{R}_{eff} must be 1 at equilibrium
 - ▶ Potentially infectious contacts are wasted
 - ▶ Many potential contacts are not susceptible (affected by disease)
 - ▶ Those not affected less susceptible than average
 - ▶ Infectious pool less infectious

Spatial and network models

- ▶ Individual-level, or spatial, heterogeneity also usually increases wasted contacts
- ▶ Infectious people meet:
 - ▶ people with similar social backgrounds
 - ▶ people with similar behaviours
 - ▶ people who are nearby geographically or in the contact network
- ▶ More wasted contacts further flatten the endemic curve

Phenomenological



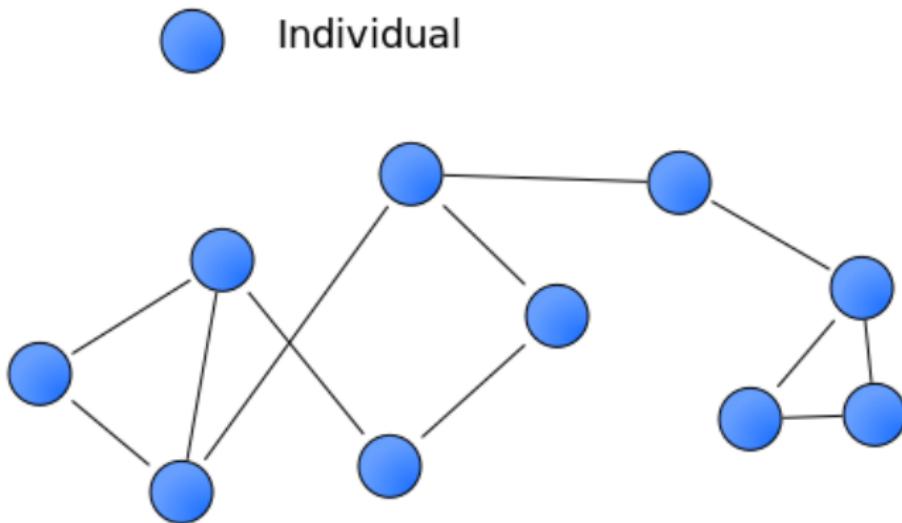


- ▶ Simply make β go down with prevalence, $\beta = B \times$:
 - ▶ $e^{-\alpha P}$
 - ▶ $(1 - P)^s$
 - ▶ $(1 - P/s)^{\alpha s}$

Multi-group models

- ▶ Divide the population into groups.
 - ▶ cities and villages
 - ▶ rich and poor
 - ▶ high and low sexual activity
 - ▶ age, gender
 - ▶ ...

Individual-based models

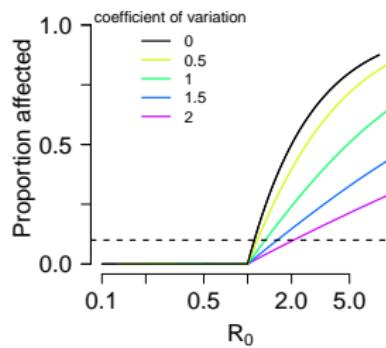


- ▶ Allow many possibilities:
 - ▶ vary individual characteristics
 - ▶ add a network of interactions
 - ▶ let the network change
- ▶ Individual-based approaches require stochastic models

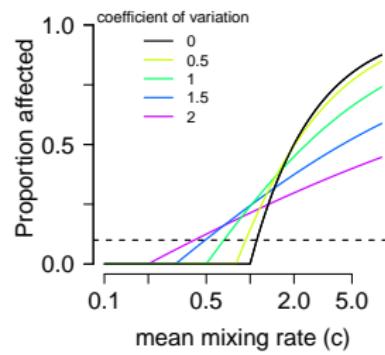
Summary



endemic equilibrium



endemic equilibrium



Summary



- ▶ People are heterogeneous in many ways
 - ▶ ... and on many scales
- ▶ Simple models give us important qualitative insights
 - ▶ Diseases in heterogeneous populations are likely to be more robust to change than expected from homogeneous models
- ▶ More complicated models can help address relevant detail



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