Heterogeneity, contact patterns and modeling options

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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

1 Homogeneous disease models

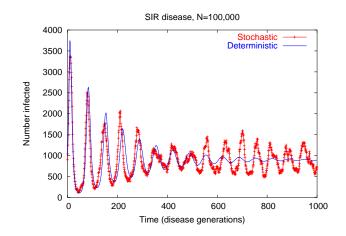
Expanding our models

- 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



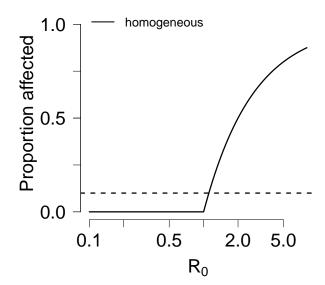
Equilibrium analysis

• \mathcal{R}_{eff} is the number of people who would be infected by an infectious individual in a general population.

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- $\mathcal{R}_{\text{eff}} = \mathcal{R}_0 \frac{S}{N} = pcD \frac{S}{N}$
- At equilibrium: $\mathcal{R}_{\text{eff}} = \mathcal{R}_0 \frac{S}{N} = 1$.
- Thus: $\frac{S}{N} = 1/R_0$.
- Proportion 'affected' is $V = 1 S/N = 1 1/R_0$.

endemic equilibrium

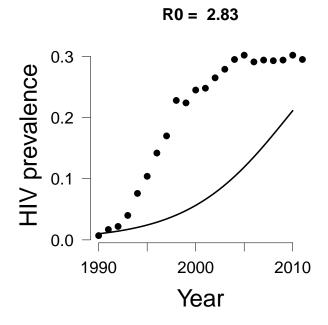


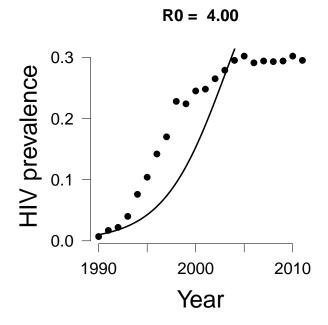
- Threshold value
- Sharp response to changes in factors underlying transmission
- Works sometimes

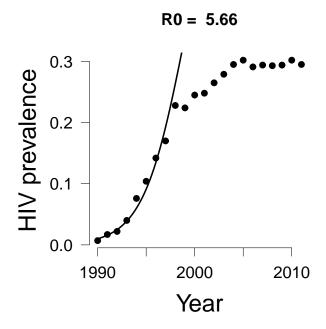
Proportion affected and disease prevalence

- ullet 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
 - $-\bar{P} = 0.95 \times (2\text{wk}/60\text{yr}).$

Homogeneous assumptions







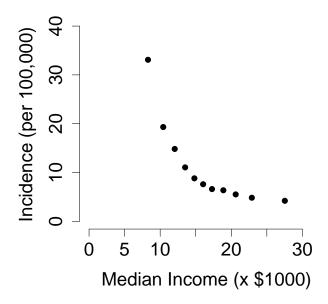
2 The importance of heterogeneity

Beyond homogeneity

- Flavors of heterogeneity
 - among hosts
 - spatial
 - $-\,$ demographic (discreteness of indviduals)
 - temporal
 - others

Heterogeneity in TB

Tuberculosis Notifications in USA, 1980s



• 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

- 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.

3 Effects of heterogeneity

Equilibrium calculations

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

Equilibrium calculations with heterogeneity

- τD applies to infectious individuals $\to \tau_I D_I$
- σ applies to susceptible individuals $\to \sigma_S$
- 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?
 - $-c_S<\bar{c};c_I\to\bar{c}.$

Simpson's paradox

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

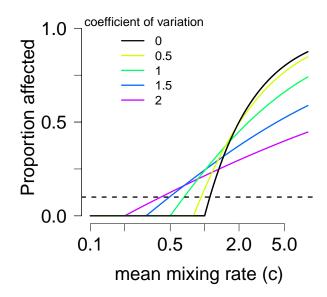
The basic reproductive number

- 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
- ullet \mathcal{R}_0 is typically greater than you would expect from a homogeneous model

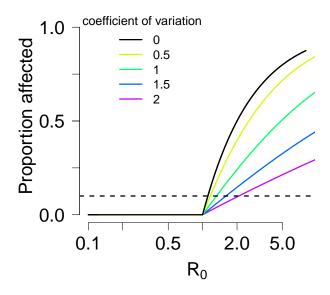
Equilibrium analysis

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

endemic equilibrium



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

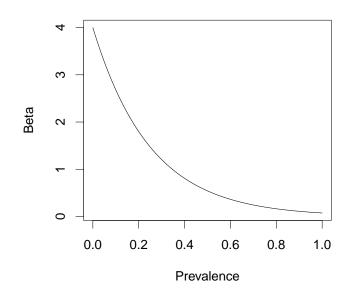
- Diseases that invade have high values of \mathcal{R}_0
- \mathcal{R}_{eff} must be 1 at equilibrium
 - Potentially infectious contacts are wasted
 - * Many potential contacts are not susceptible (affected by disease)
 - * Those not affected less susceptible than average
 - Infectious pool less infectious

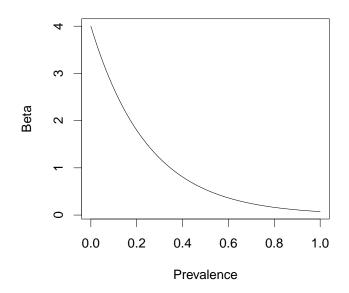
Spatial and network models

- 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

4 Modeling approaches

Phenomenological



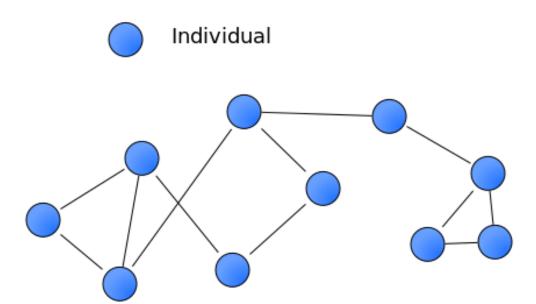


- Simply make β go down with prevalence, $\beta = B \times$:
 - $-e^{-\alpha P}$
 - $-(1-P)^{s}$
 - $-(1-P/s)^{\alpha s}$

Multi-group models

- $\bullet\,$ 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

- People are heterogeneous in many ways
 - $-\ldots$ 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