

Mathematical Models of Infectious Diseases

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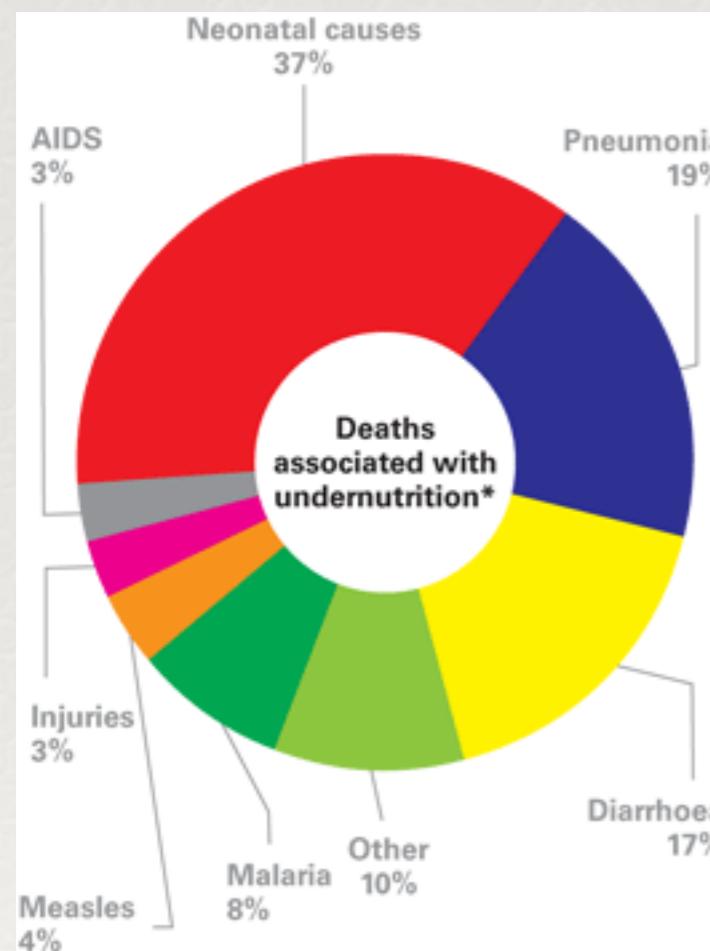
Infectious Disease University of Georgia

University of Georgia

Global causes of mortality

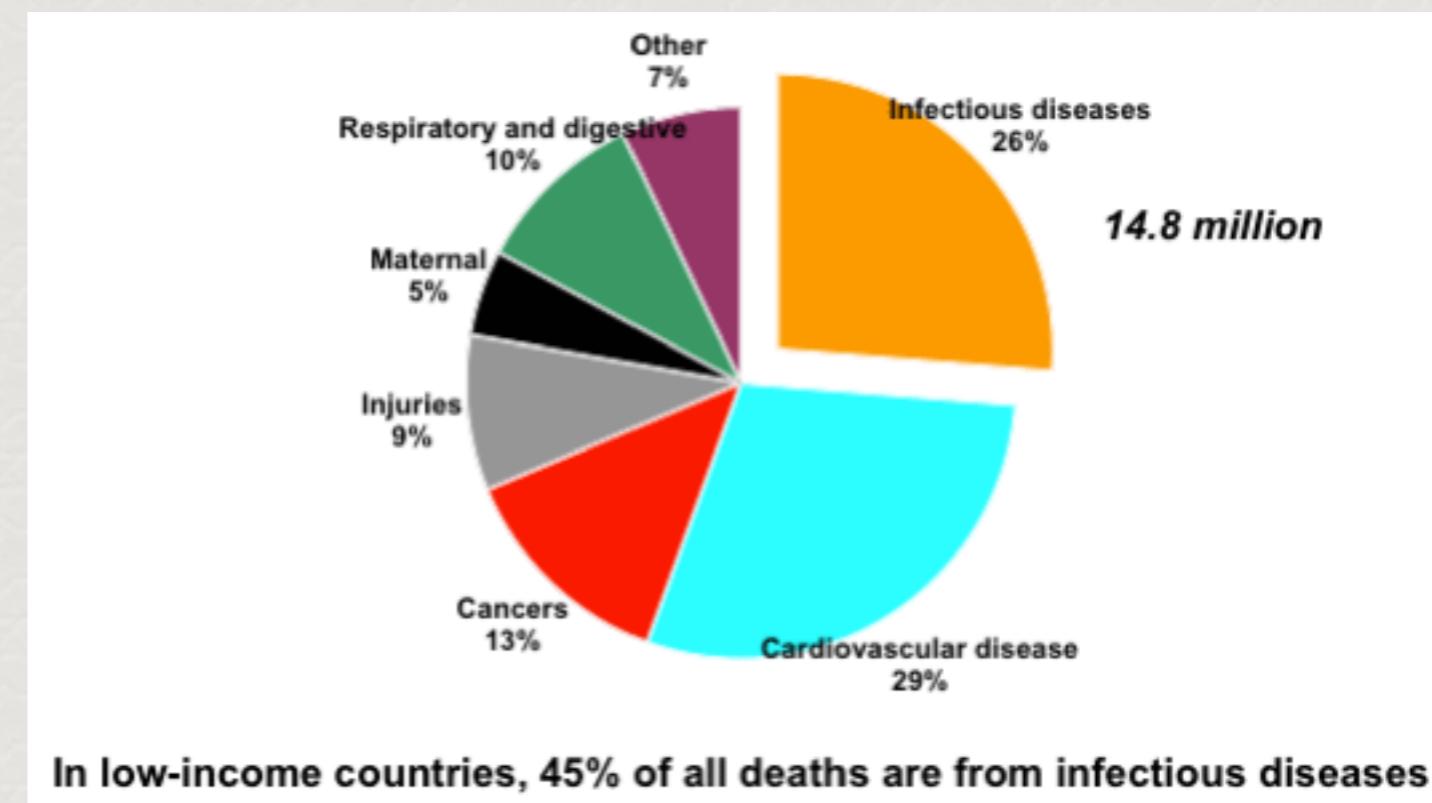


World Health Organization



* Undernutrition has been estimated to be an underlying cause in up to half of all under-five deaths. This estimate will be revised in 2008.

Measles & pertussis account for ~300,000 and ~200,000 annual deaths



Total mortality

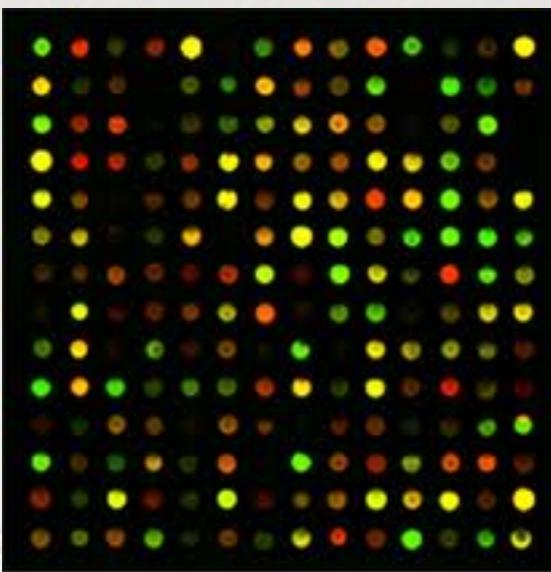
Infant mortality

Multifaceted approach to understanding infectious diseases

Medicine



Genomics

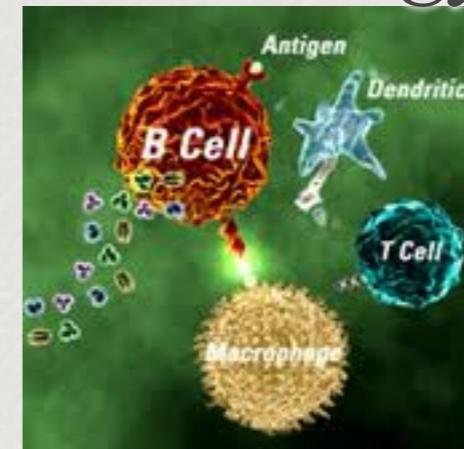


But these approaches don't address important questions at population level ...

Microbiology



Immunology



Vaccines & Drugs

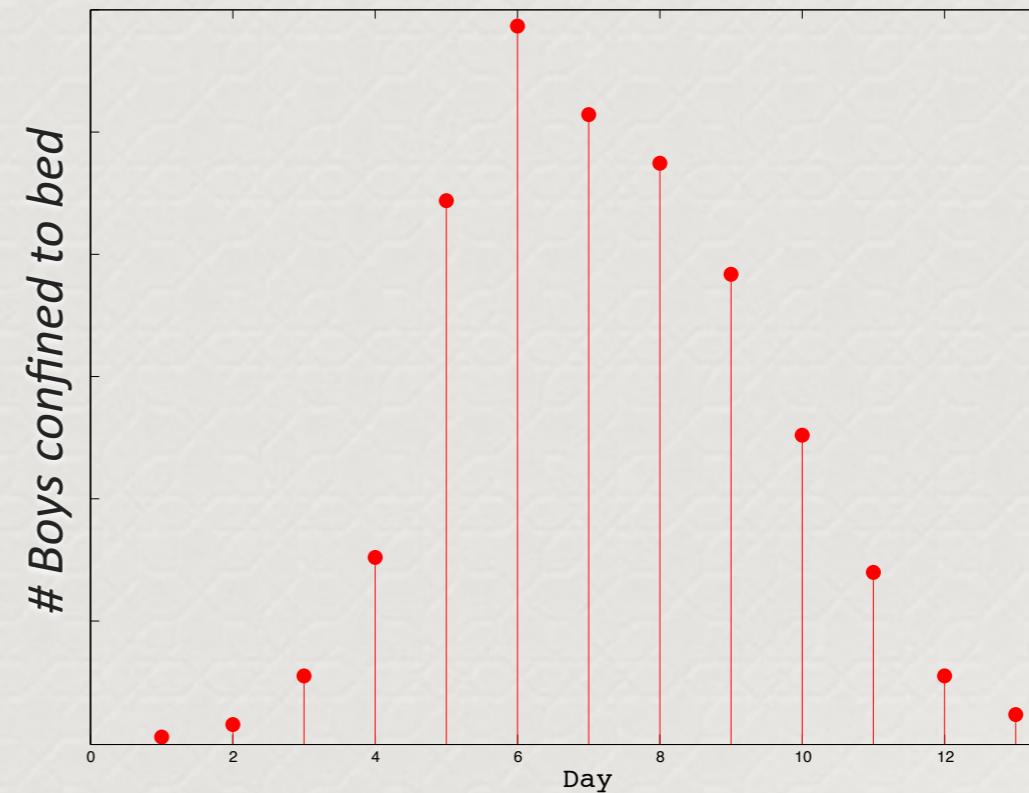


School outbreak

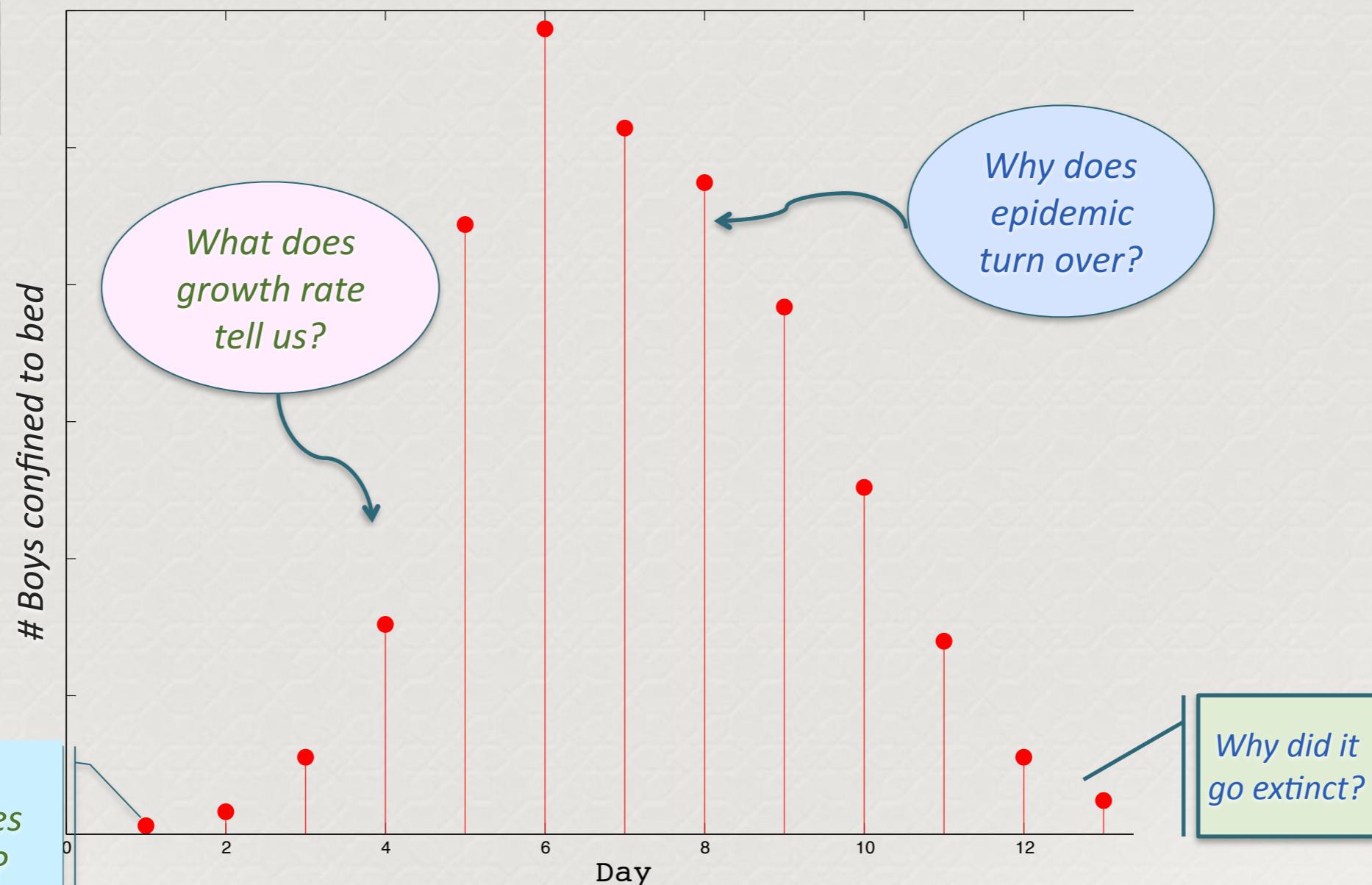


*Boarding School, England
Jan 1978*

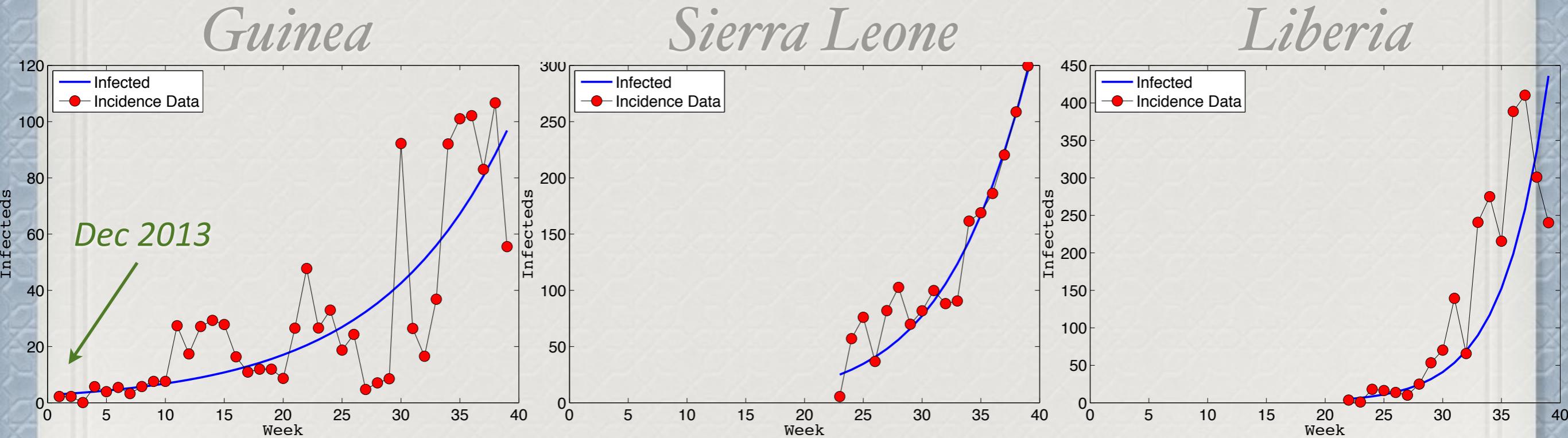
- Raises numerous questions:*
- What is etiological agent?*
- Is it novel?*
- Is a vaccine available?*



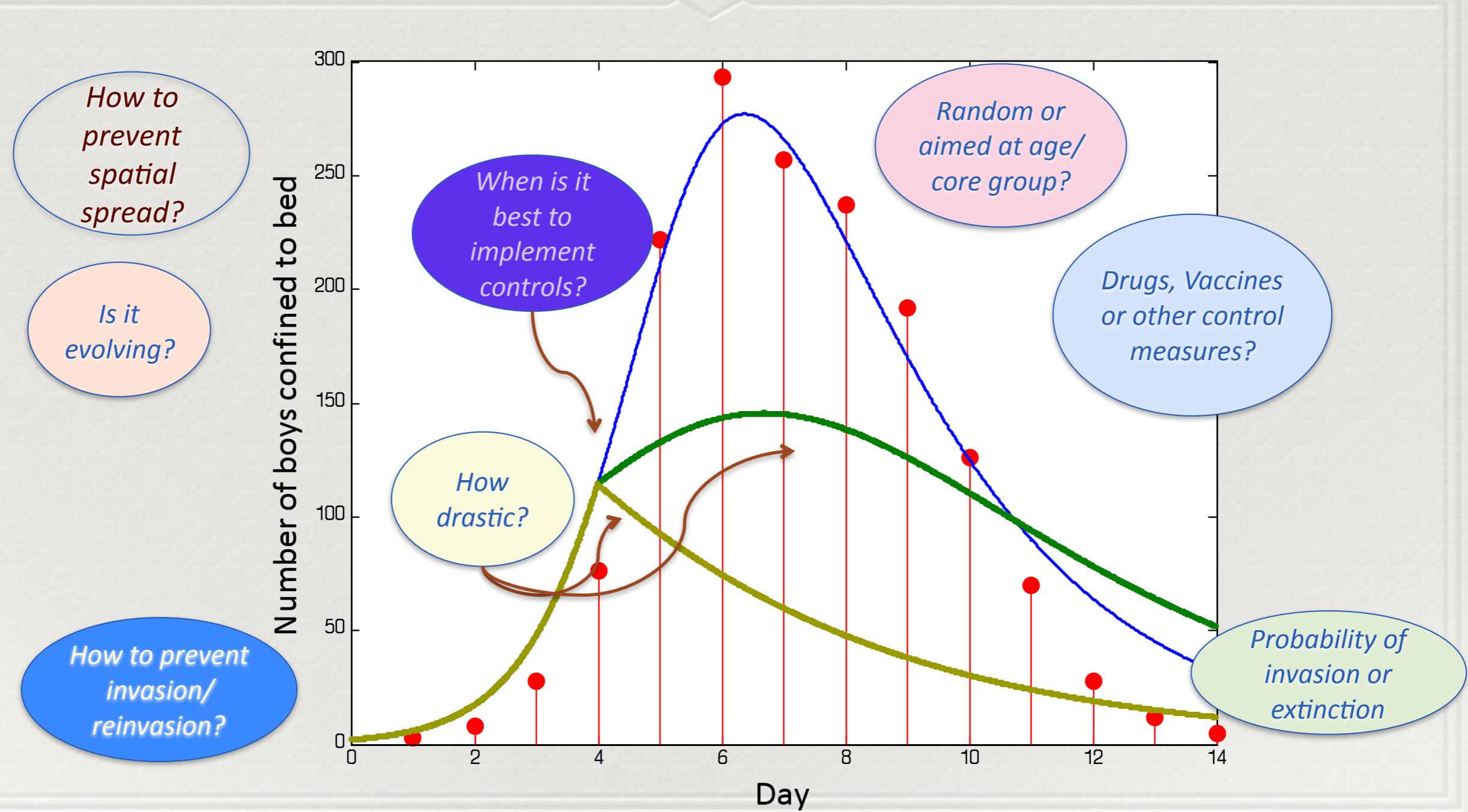
Modeling questions I. Basics



2014 Ebola outbreak in West Africa



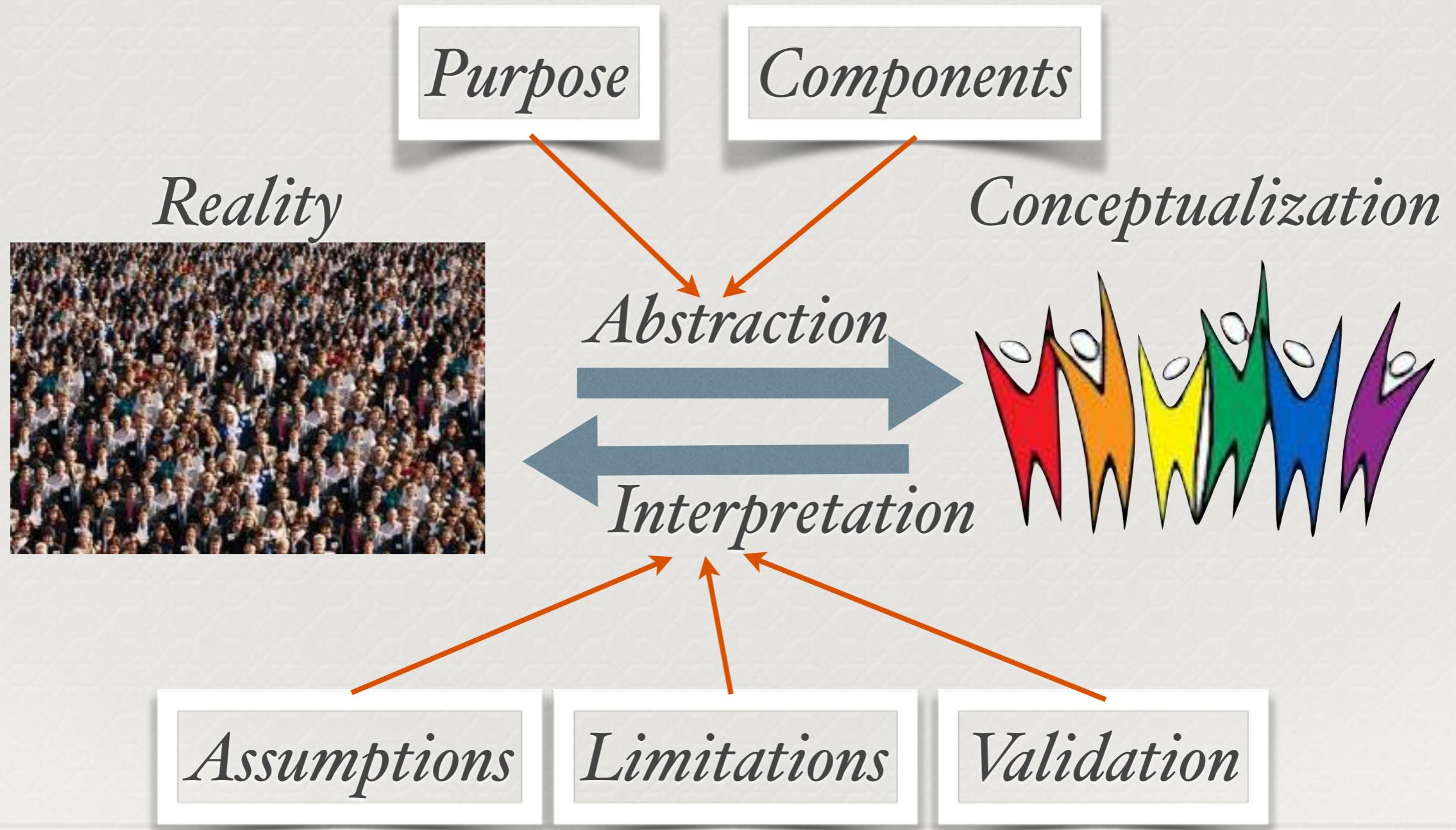
Modeling questions II. Control Implications



What is a model?

- ◆ *Different types of models:*
 - ◆ *A mathematical/computational model is an abstract model that uses mathematical language to describe the behaviour of a system*
 - ◆ *A Statistical model attempts to describe relationships between observed quantities and independent variables*
 - ◆ *Developing a model is different from statistical analyses of data*

Abstraction



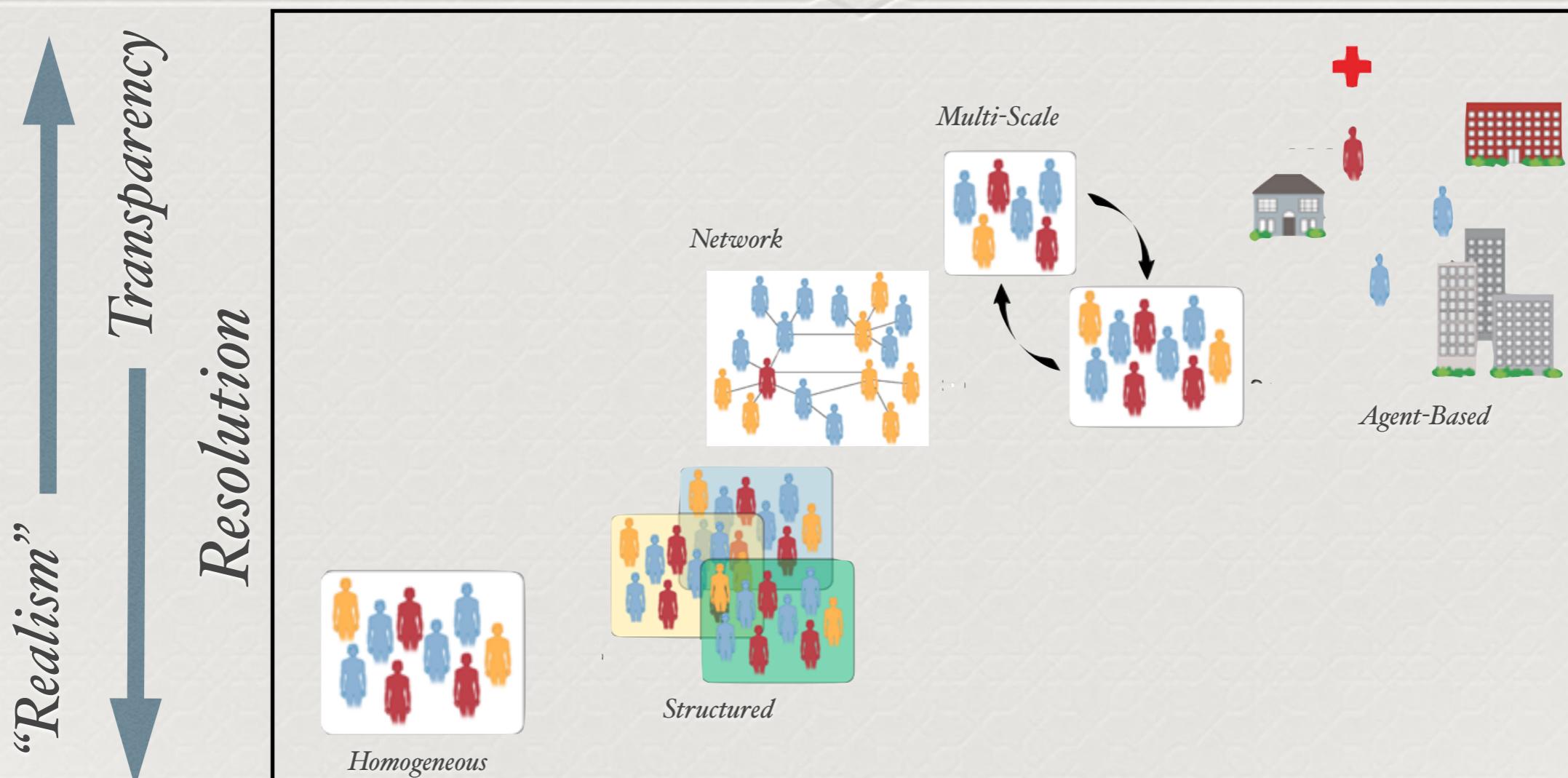
What's a 'Good' Model?

- ◆ Choice of model depends crucially on focal question and available data (hammer & chisel or pneumatic drill?)
- ◆ Use model principally for
 - ◆ understanding nature
 - ◆ making predictions

Judging a Model...

- ◆ Three fundamental features of models, often opposing forces:
 - ◆ **Accuracy**
 - ◆ *Capture observed patterns (qualitative or quantitative?) and make predictions*
 - ◆ *Increases with model complexity*
 - ◆ **Transparency**
 - ◆ *Ability to understand model components*
 - ◆ *Decreases with model complexity*
 - ◆ **Flexibility**
 - ◆ *How easily can model be adapted to new scenarios?*
 - ◆ *Decreases with model complexity*

Realism Vs Transparency



Solution tools



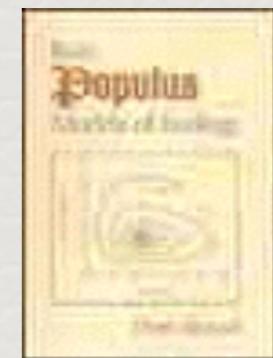
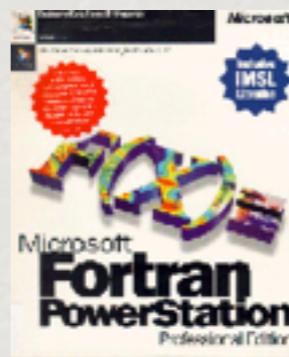
‘How’ do you Model?

Analytical Models

Concentrate on problems that can be expressed and analysed fully using analytical approaches.



The MathWorks



ModelMaker v.00

Problem-based Models

Construct most “appropriate” model and use whatever combination of methods for analysis and prediction.



Ready-Made Software

ModelMaker

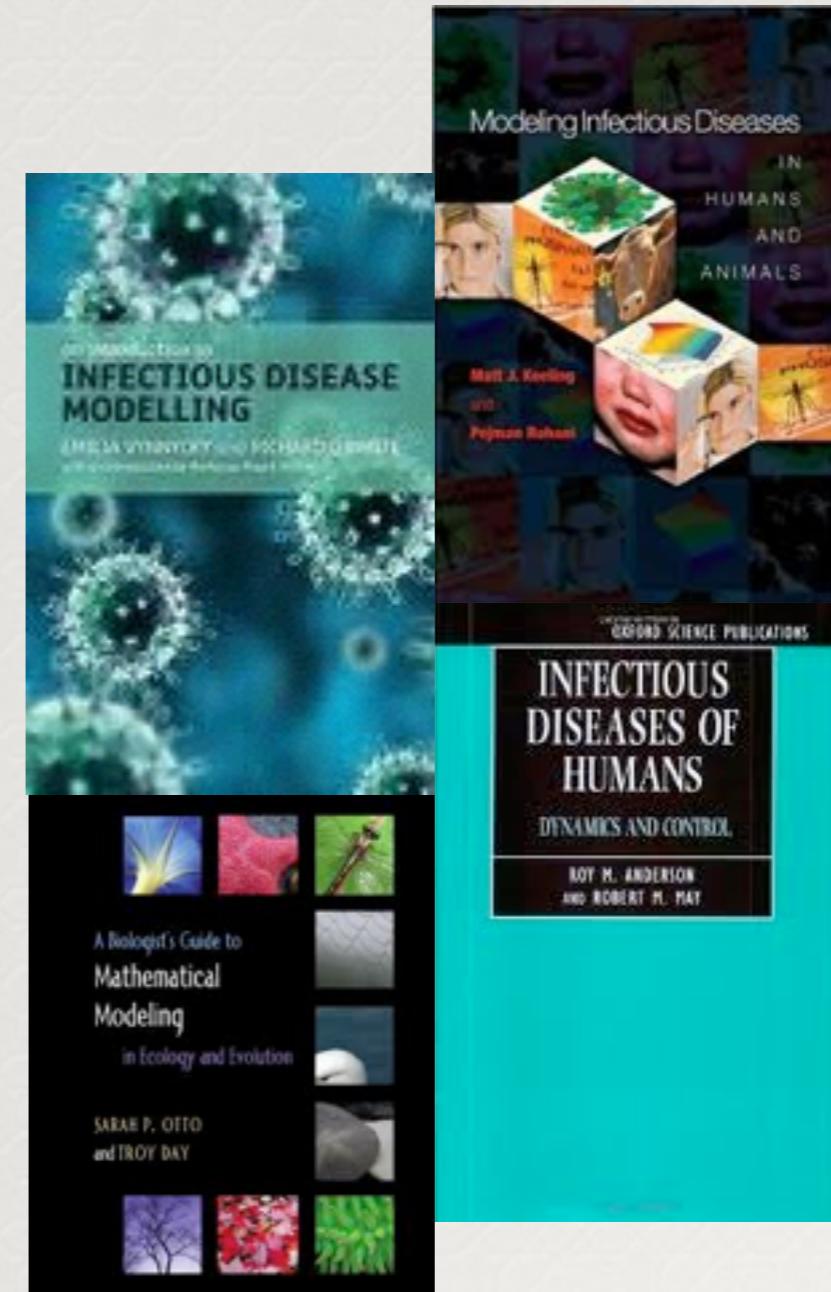
www.modelkinetix.com/modelmaker/modelmaker.html

Global simulators



Resource Materials

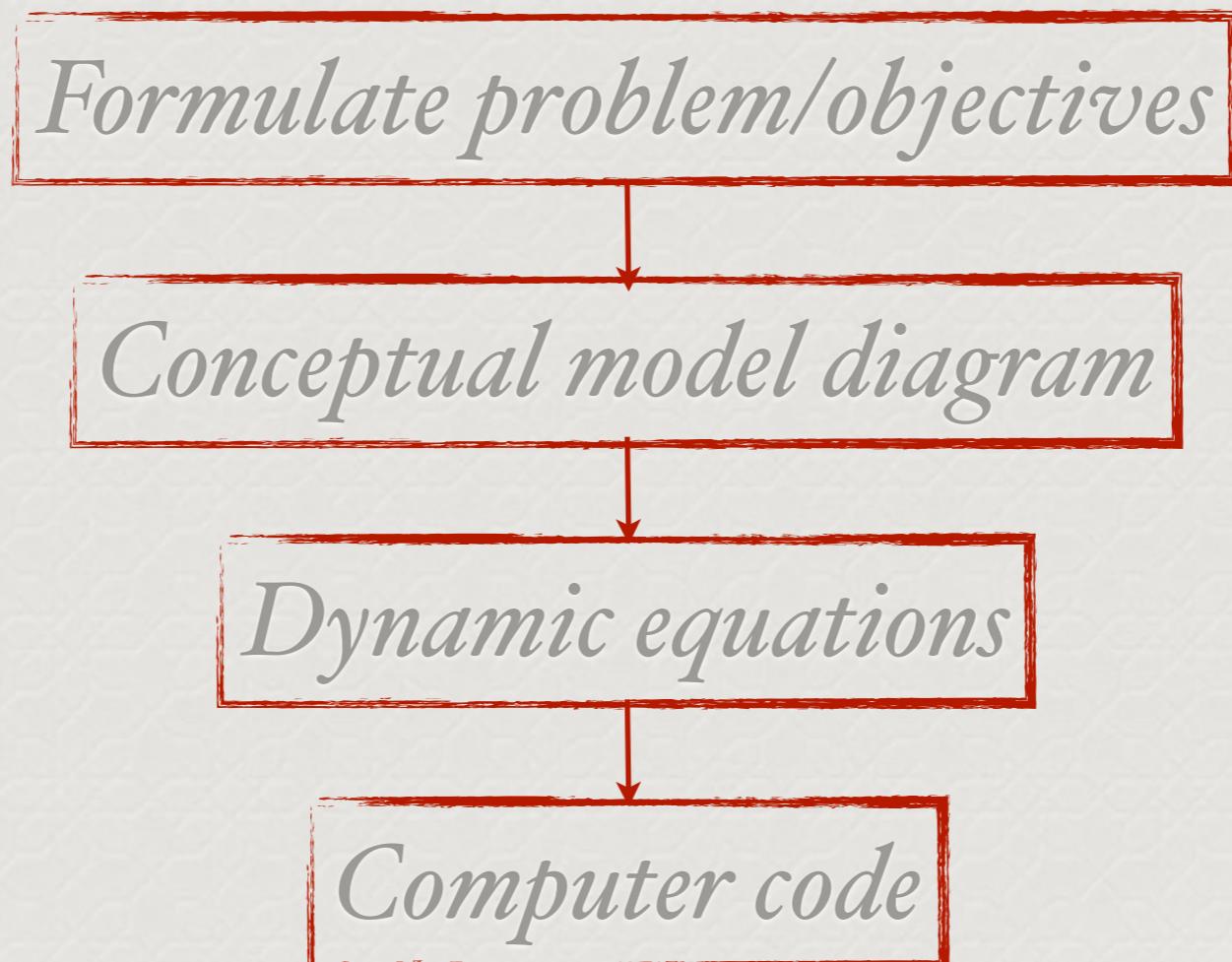
- ◆ *Keeling & Rohani (2008)*
- ◆ *Vynnycky & White (2010)*
- ◆ *Anderson & May (1991)*
- ◆ *Otto & Day (2007)*



Mathematical Modelling of Infectious Diseases

- ◆ Objective 1: Setting up simple models
 - ◆ Different transmission modes
 - ◆ Basic Reproduction Ratio (R_0), Simple Epidemics, Invasion threshold & extinction
 - ◆ Stability analysis
- ◆ Objective 2: Control
 - ◆ Infection management
- ◆ Objective 3: Statistical estimation
 - ◆ R_0 and other parameters
- ◆ Objective 4: Heterogeneities
 - ◆ Risk structure
 - ◆ Realistic pathogenesis
 - ◆ Seasonality
 - ◆ Age-structured transmission effects
- ◆ Objective 5: Sensitivity
 - ◆ Stochastic implementation
 - ◆ Parameter uncertainty

Steps in Developing a Model



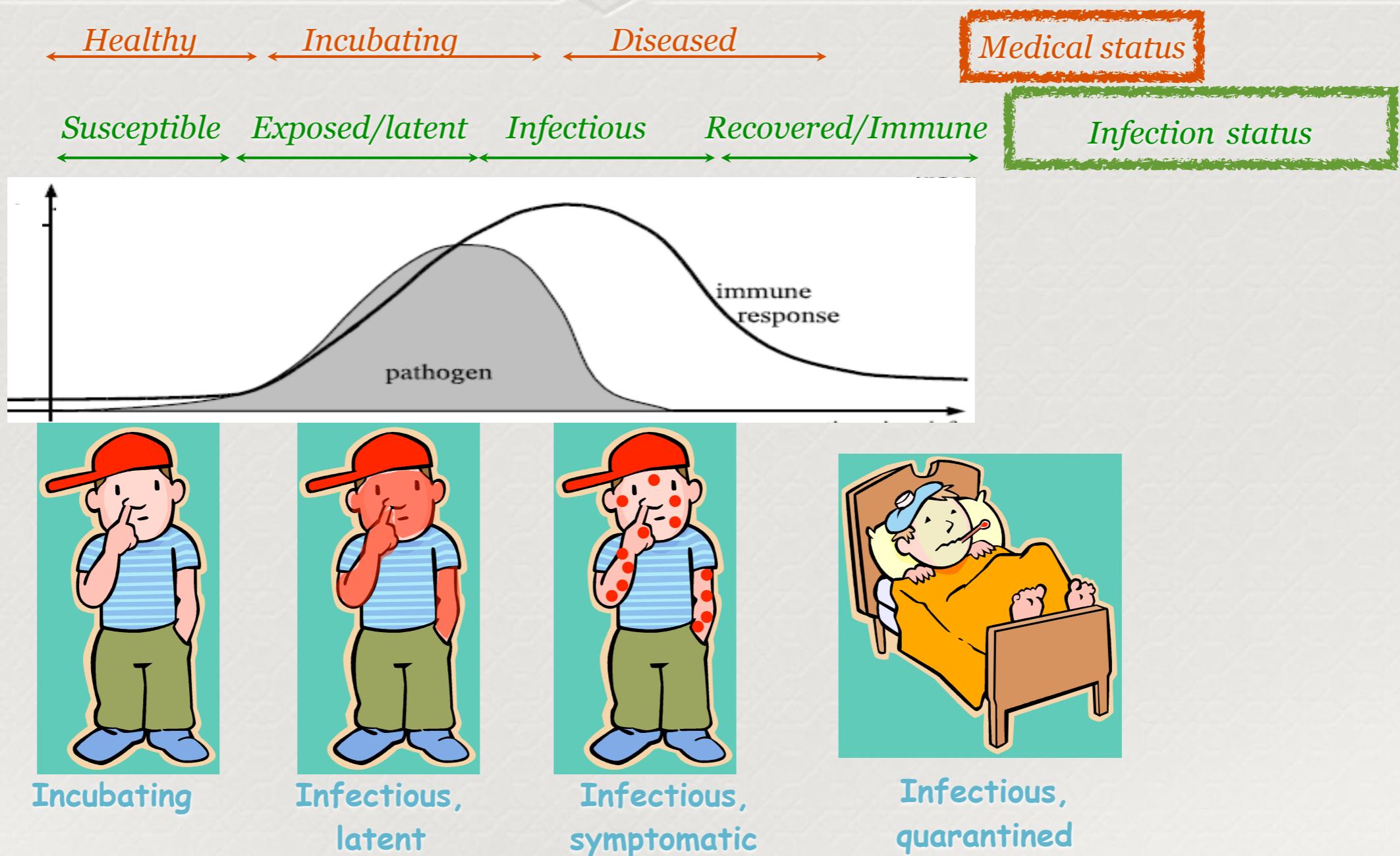
The simplest models

- ◆ Let's develop a model for Boarding School influenza outbreak
- ◆ Some **important** choices need to be made at outset

I. ***What do we want to keep track of?***

- ◆ *Amount of virus in population?*
- ◆ *Antibody titre of everyone in population (school)?*
- ◆ *Cities in which infected people have been found?*

Categorising individuals



The simplest models

- ◆ Pragmatic choice: categorise individuals in population according to their infection status, eg:



*These are our
“system variables”*

The simplest models

2. What model structure?

- Determined by pathogen biology



SI – signifies fatal infection



SIR – recovery after infection



SEIR – latency

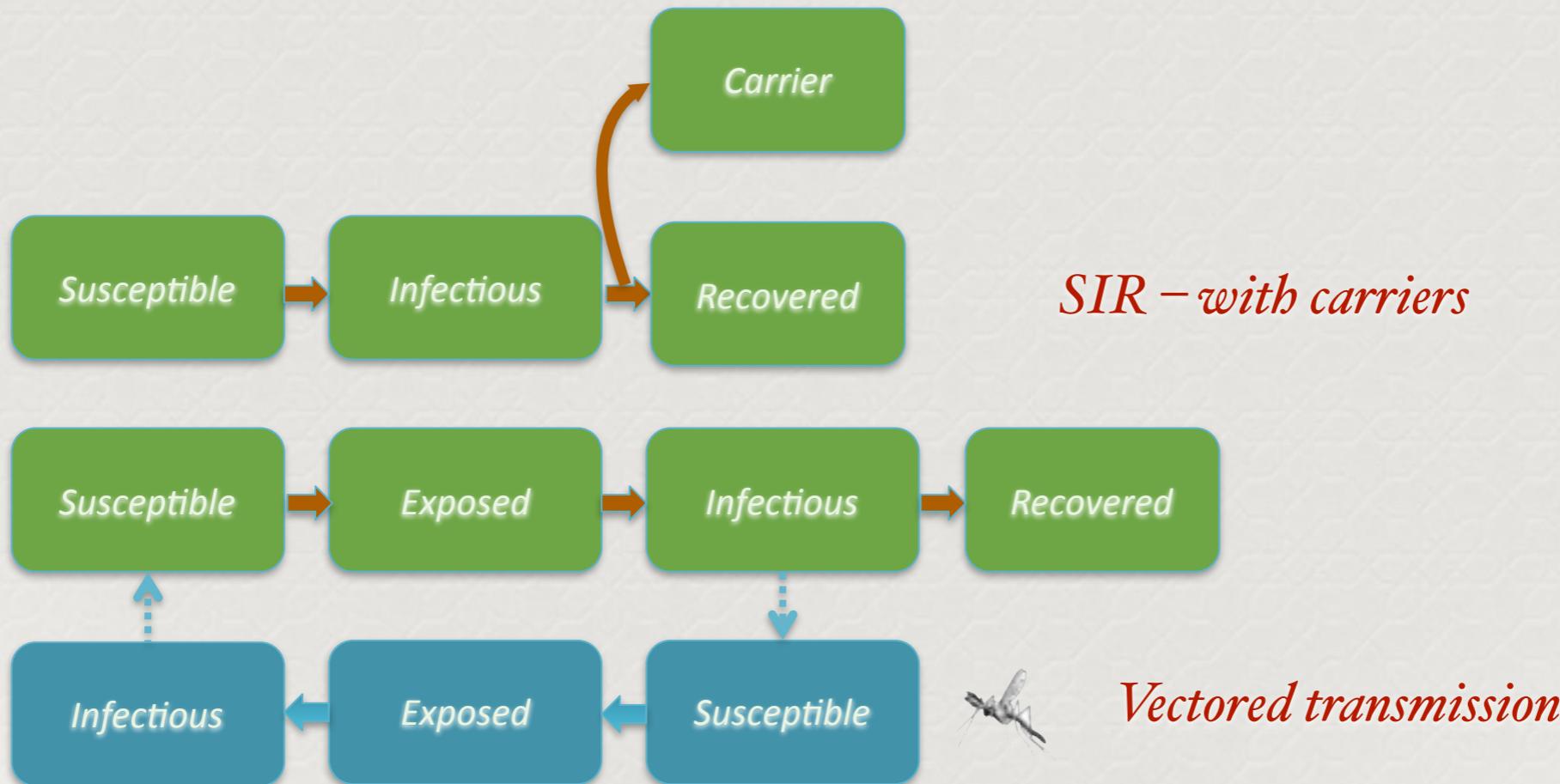


SIS – no immunity elicited

The simplest models

2. What model structure?

-- Determined by pathogen biology



The simplest models

- ◆ **What model structure?**
- ◆ Depends on what do we know about the pathogen (eg, influenza)
 - ◆ It's directly transmitted (aerosol)
 - ◆ An acute infection
 - ◆ Lifelong immunity (to that strain)



The simplest models



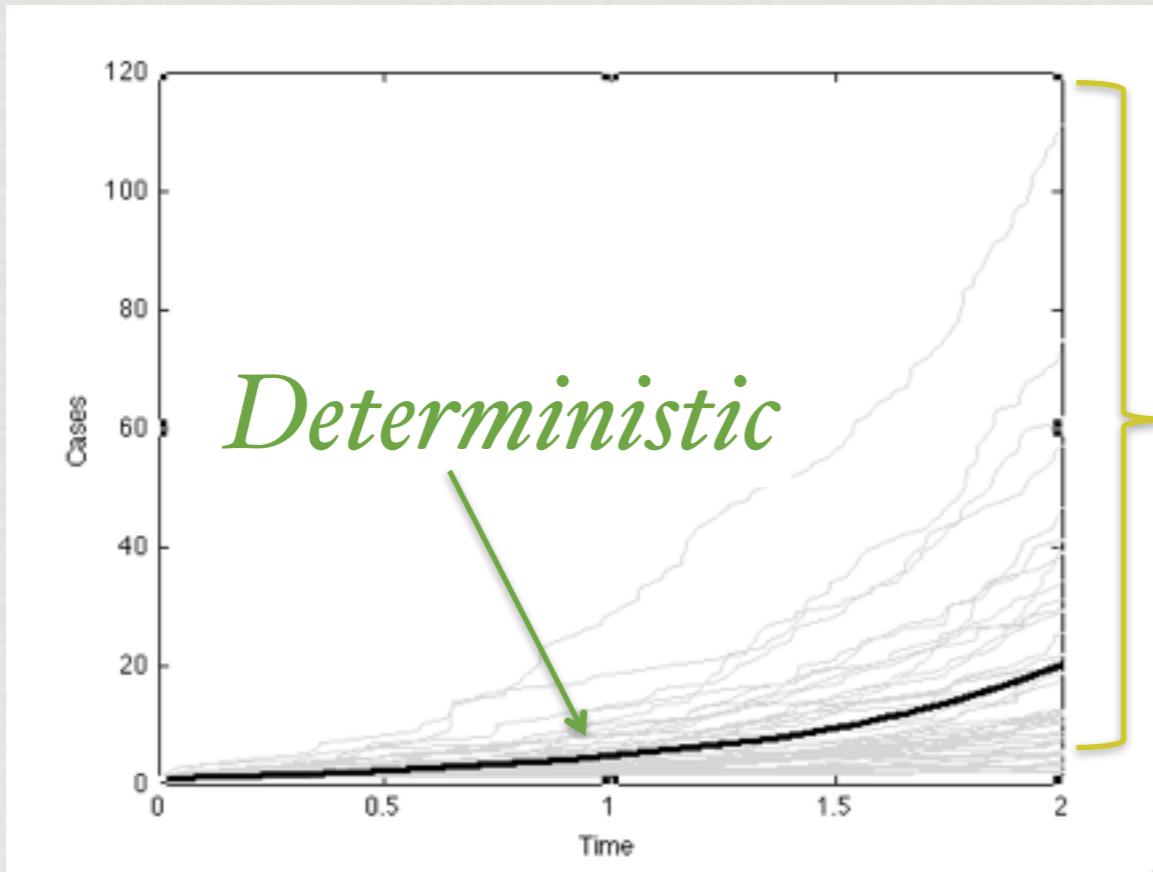
- Flow between classes/compartments determined by details of *host population structure* and *pathogen biology*

- Host population size
- Contact rates
- Pathogen infectivity

These are our “parameters”

The simplest models

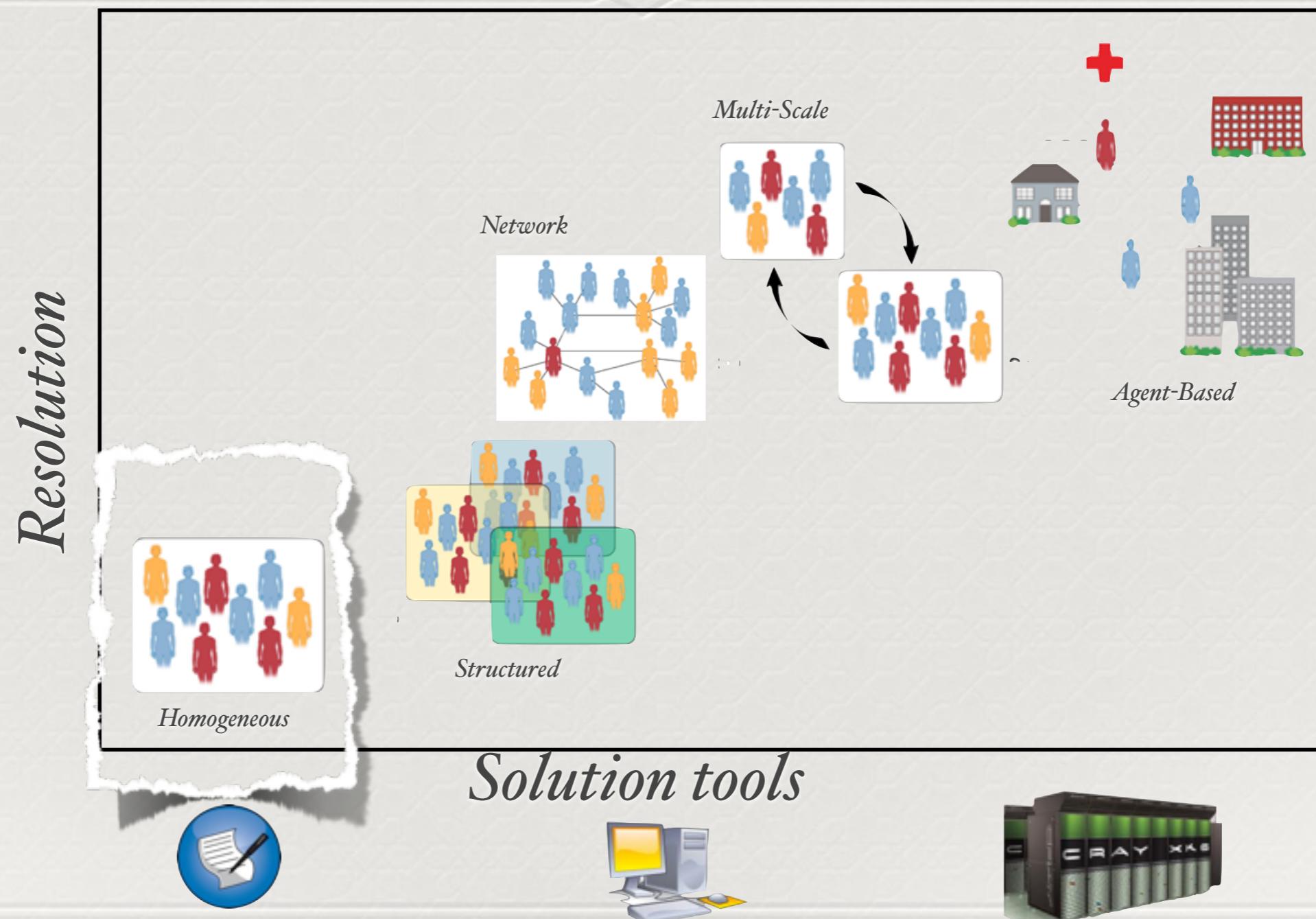
3. Deterministic or stochastic?



*50 independent
stochastic
realizations*

On average, stochastic simulations identical to deterministic predictions, though individual realizations may be quite different

Realism Vs Transparency



The simplest models

- ◆ We've settled on a deterministic SIR model – now what?
- ◆ How do we write down some equations to describe spread of 'flu in this population?
- ◆ Assign each system variable a unique Roman letter, eg:
 - ◆ Susceptible, S (proportion) or X (number)
 - ◆ Infectious, I (proportion) or Y (number)
 - ◆ Recovered/Immune, R (proportion) or Z (number)
- ◆ Assign parameters a unique (typically Greek) letter, eg:
 - ◆ Contact rate, κ
 - ◆ Pathogen infectivity, ν

Very important!

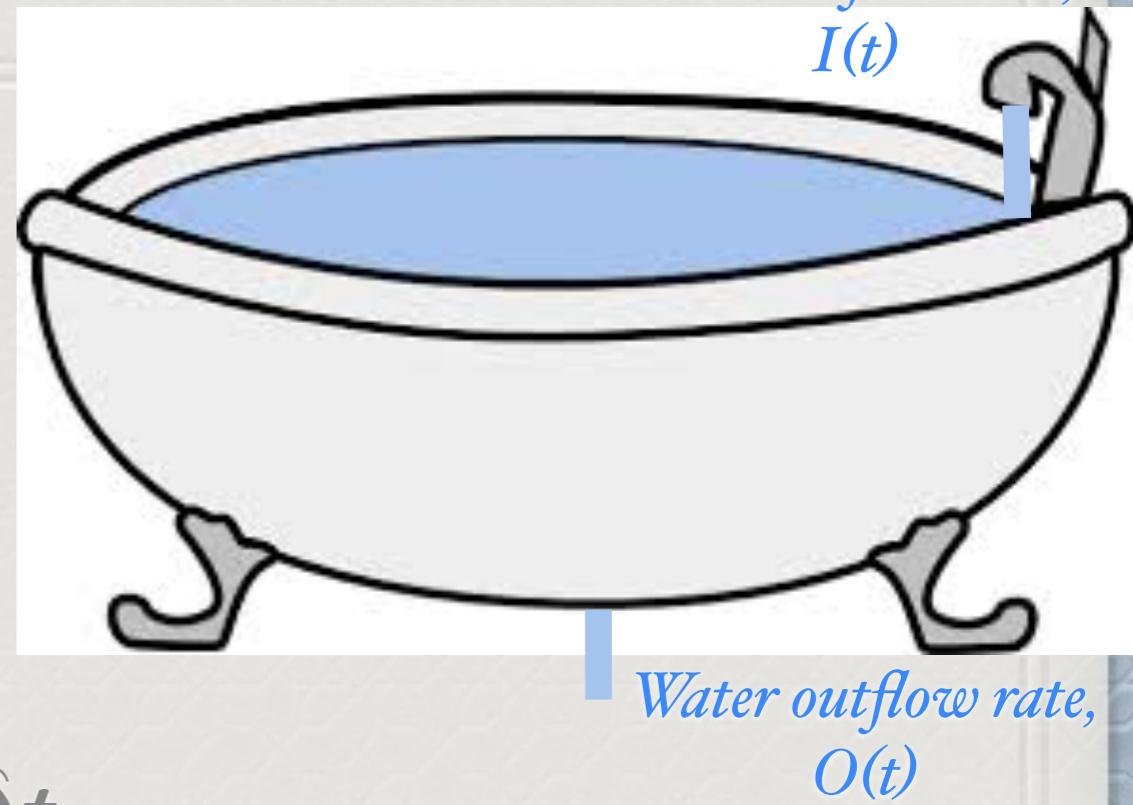
- ◆ NOTHING SPECIAL ABOUT MY CHOICE OF NOTATION
 - USE OF PARTICULAR LETTERS HIGHLY IDIOSYNCRATIC
- ◆ OTHER AUTHORS MAY USE DIFFERENT LETTERS TO DENOTE SAME VARIABLES OR PARAMETERS.
- ◆ YOU CANNOT AUTOMATICALLY ASSUME THAT β IN TWO DIFFERENT PAPERS MEANS THE SAME THING!

3. Model equations

Bath tub example

- ◆ Let $W(t)$ be amount of water in bathtub (ml)
- ◆ Need a dynamic equation that tells us how $W(t)$ will change through time
- * Consider a small time interval, δt
- * Then,

$$W(t + \delta t) = W(t) + \text{Inflow rate} \times \text{elapsed time} - \text{Outflow rate} \times \text{elapsed time}$$



Bath tub example

$$W(t + \delta t) = W(t) + I \times \delta t - O \times \delta t$$

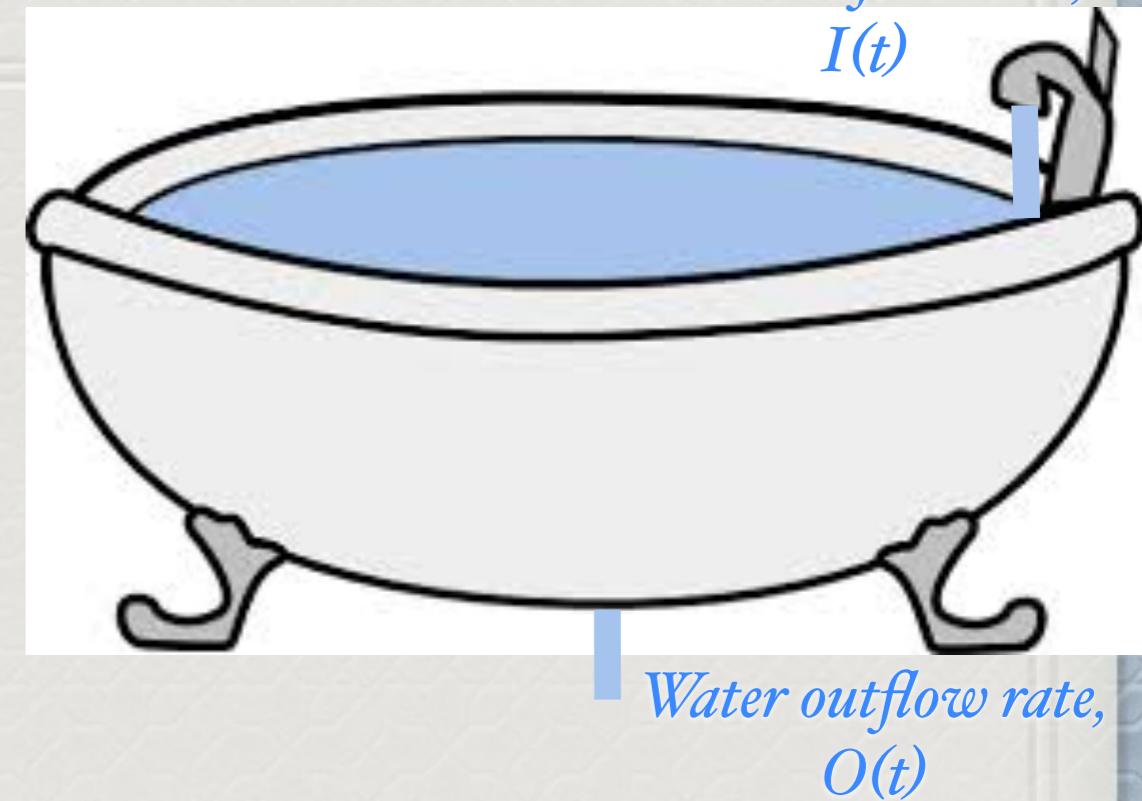
* Rearrange

$$\frac{W(t + \delta t) - W(t)}{\delta t} = I - O$$

* Left hand side is a difference quotient for derivative of W with respect to time

* Let $\delta t \rightarrow 0$

$$\frac{dW}{dt} = I - O$$



Many bathtubs =
compartment models

Model equations

- If we knew X_t and Y_t , could we predict $X_{t+\delta t}$ and $Y_{t+\delta t}$, where δt is some (very short) time later?

$$X_{t+\delta t} = X_t - (\nu \kappa \delta t) X_t Y_t / N$$

$$Y_{t+\delta t} = Y_t + (\nu \kappa \delta t) X_t Y_t / N - (\gamma \delta t) Y_t$$

- And

$$Z_{t+\delta t} = Z_t + (\gamma \delta t) Y_t$$

ν is probability of transmission given contact

κ is contact rate

Basic questions?

$$\beta = \nu K$$

$$X_{t+\delta t} = X_t - (\beta \delta t) X_t Y_t / N$$

$$Y_{t+\delta t} = Y_t + (\beta \delta t) X_t Y_t / N - (\gamma \delta t) Y_t$$

$$Z_{t+\delta t} = Z_t + (\gamma \delta t) Y_t$$

- Average infectious period given by $1/\gamma$ [why?]

Mean life time calculation

Consider recovery of a single infectious individual $I(t) = e^{-\gamma t}$

$$1 = \int_0^\infty ce^{-\gamma t} dt = \frac{c}{\gamma}$$

Hence, probability density function is $\gamma e^{-\gamma t}$

$$\tau = \int_0^\infty t \gamma e^{-\gamma t} dt = \frac{1}{\gamma}$$

For a random variable x , with probability density function $f(x)$, the mean is given by $\int_0^\infty xf(x)dx$

An ODE model

- Consider the equation describing Susceptible dynamics

$$X_{t+\delta t} = X_t - (\beta \delta t) X_t Y_t / N$$

- Re-write as

$$X_{t+\delta t} - X_t = - (\beta \delta t) X_t Y_t / N$$

$$(X_{t+\delta t} - X_t) / \delta t = \beta X_t Y_t / N$$

By fundamental theorem of calculus, as $\delta t \rightarrow 0$,

$$dX/dt = - \beta X Y / N$$

An ODE SIR model

$$\frac{dX}{dt} = -\beta X \frac{Y}{N}$$

$$\frac{dY}{dt} = \beta X \frac{Y}{N} - \gamma Y$$

$$\frac{dZ}{dt} = \gamma Y$$

- By definition, $X+Y+Z = N$
- These equations describe rates of change in state variables
- Parameters β, γ represent instantaneous rates

An ODE SIR model

In my lectures (as in K&R 2008),
variables X , Y & Z refer to the
numbers of individuals in each class.
Variables S , I , & R refer to the
proportions of the population in
each class

- These equations describe rates of change in state variables
- Parameters β , γ represent instantaneous rates

An ODE SIR model

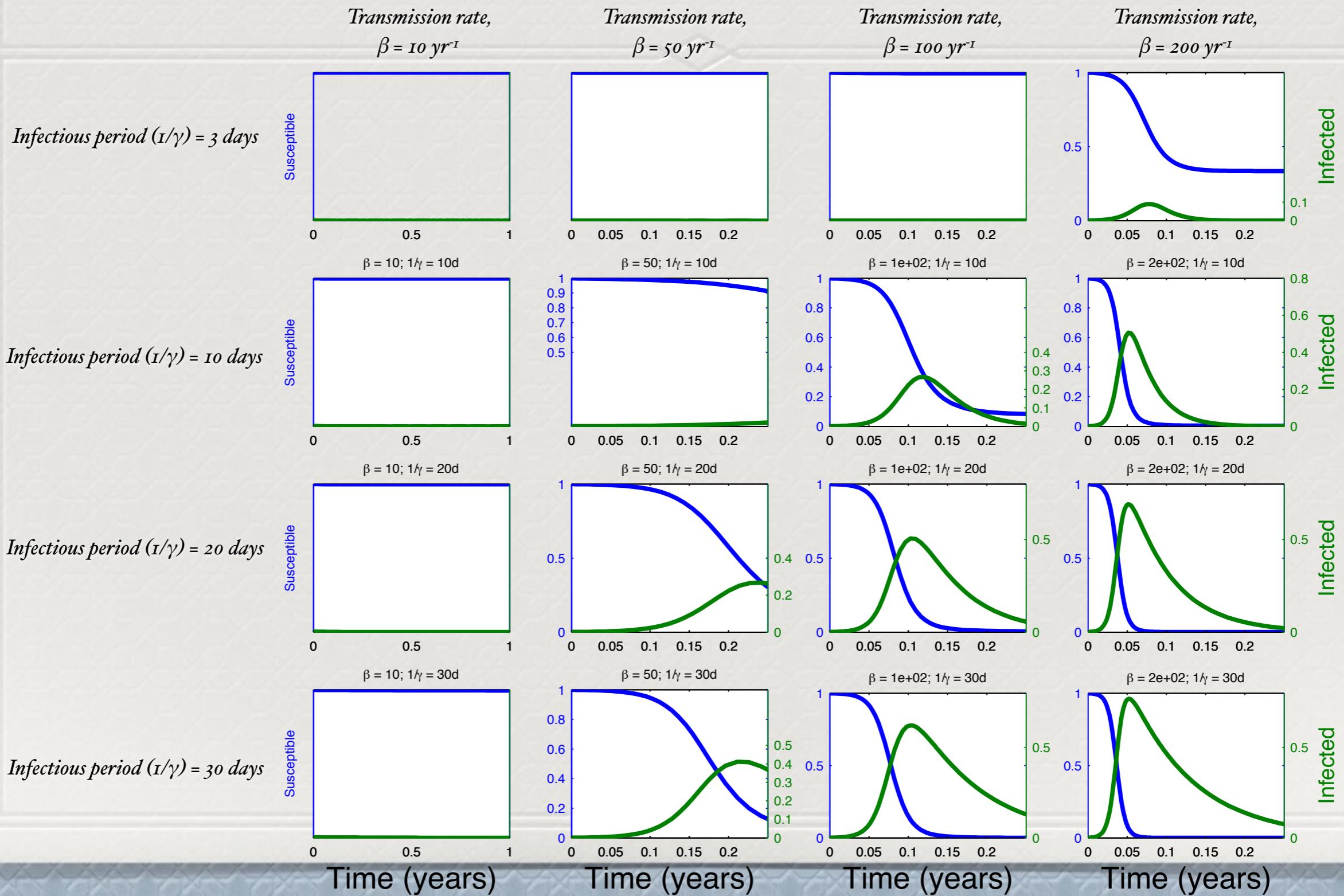
$$\frac{dX}{dt} = -\beta X \frac{Y}{N}$$

$$\frac{dY}{dt} = \beta X \frac{Y}{N} - \gamma Y$$

$$\frac{dZ}{dt} = \gamma Y$$

- Important to notice: transmission rate is assumed to depend on frequency of infecteds in population (Y/N). Hence, this is **frequency-dependent transmission**

Simulating epidemics



Model dynamics

- ◆ *As parameters are varied, model predicts different outcomes*
- ◆ *Can we anticipate trajectories without resorting to numerical integration?*
- ◆ *Question: under what conditions will an infectious disease invade a system?*

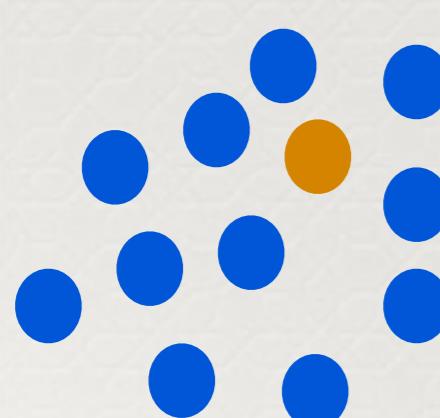
The Invasion Threshold

- ◆ When can an infectious disease invade a population?
- ◆ Initial conditions: $X(0) = N$, $Y(0) = 1$, $Z(0) = 0$
- ◆ Invasion only if $dY/dt > 0$
- ◆ ie, $\beta XY/N - \gamma Y > 0 \Rightarrow Y(\beta X/N - \gamma) > 0$
 - ◆ If and only if $X/N > \gamma/\beta$
 - ◆ Since $X=N$, requires $1 > \gamma/\beta$
 - ◆ Or $\beta/\gamma > 1$

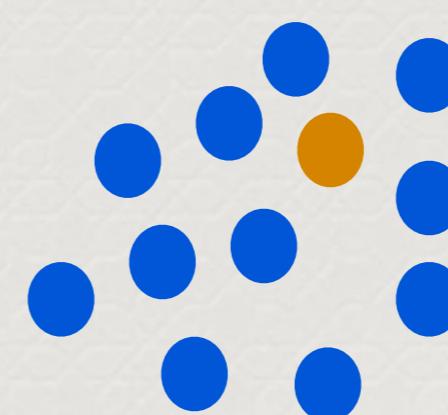
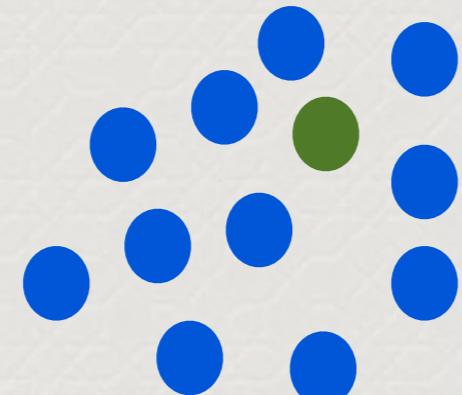
Kermack & McKendrick (1927)

Basic Reproductive Ratio, R_o

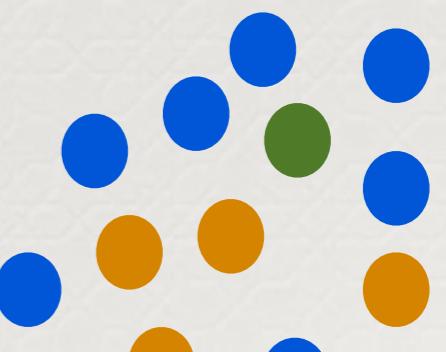
- Ratio β/γ gives number of cases before infected individual recovers
- Universally referred to as R_o or **Basic Reproductive Ratio**
- Definition: Number of secondary cases generated by a typical infected in an entirely susceptible population



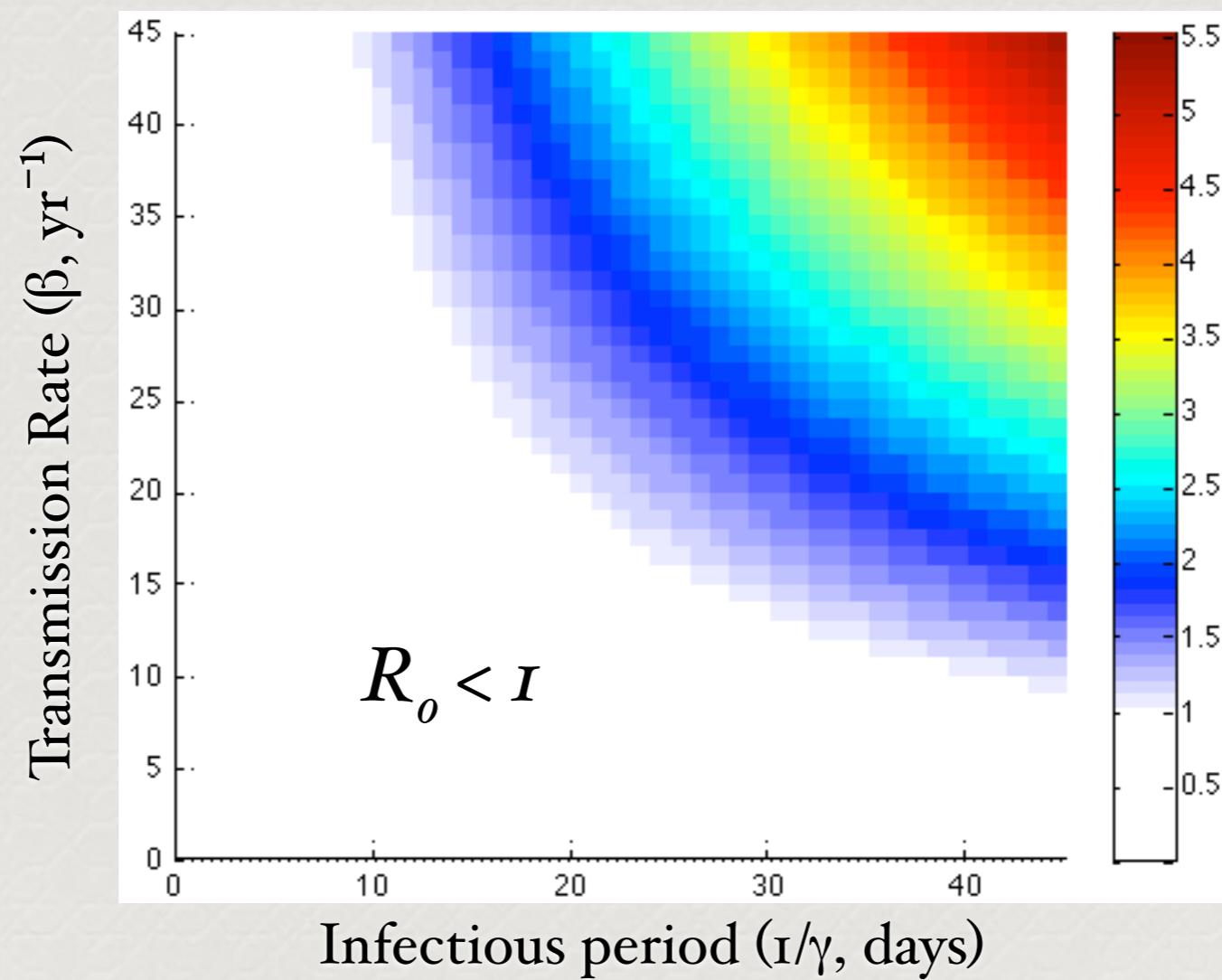
$R_o < 1$
No invasion



$R_o = 4$
Successful invasion



R_o and Model parameters



Estimates of R_0

R_0



Hepatitis C



Seasonal Influenza



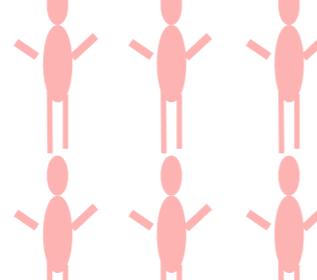
1918 Influenza



Ebola



SARS



Phocine Distemper



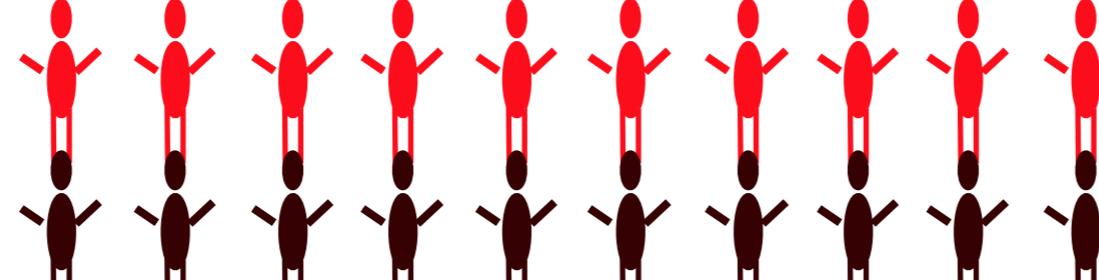
HIV (MSM)



HIV (FSW)



Mumps



Pertussis



The death of an epidemic

- ◆ In SIR equations, let's divide equation for dX/dt by dZ/dt :

$$\begin{aligned} dX/dZ &= -(\beta X Y/N)/(\gamma Y) \\ &= -R_o X/N \end{aligned}$$

- ◆ Integrate with respect to Z
 - ◆ $X(t) = X(0) e^{-Z(t) R_o/N}$

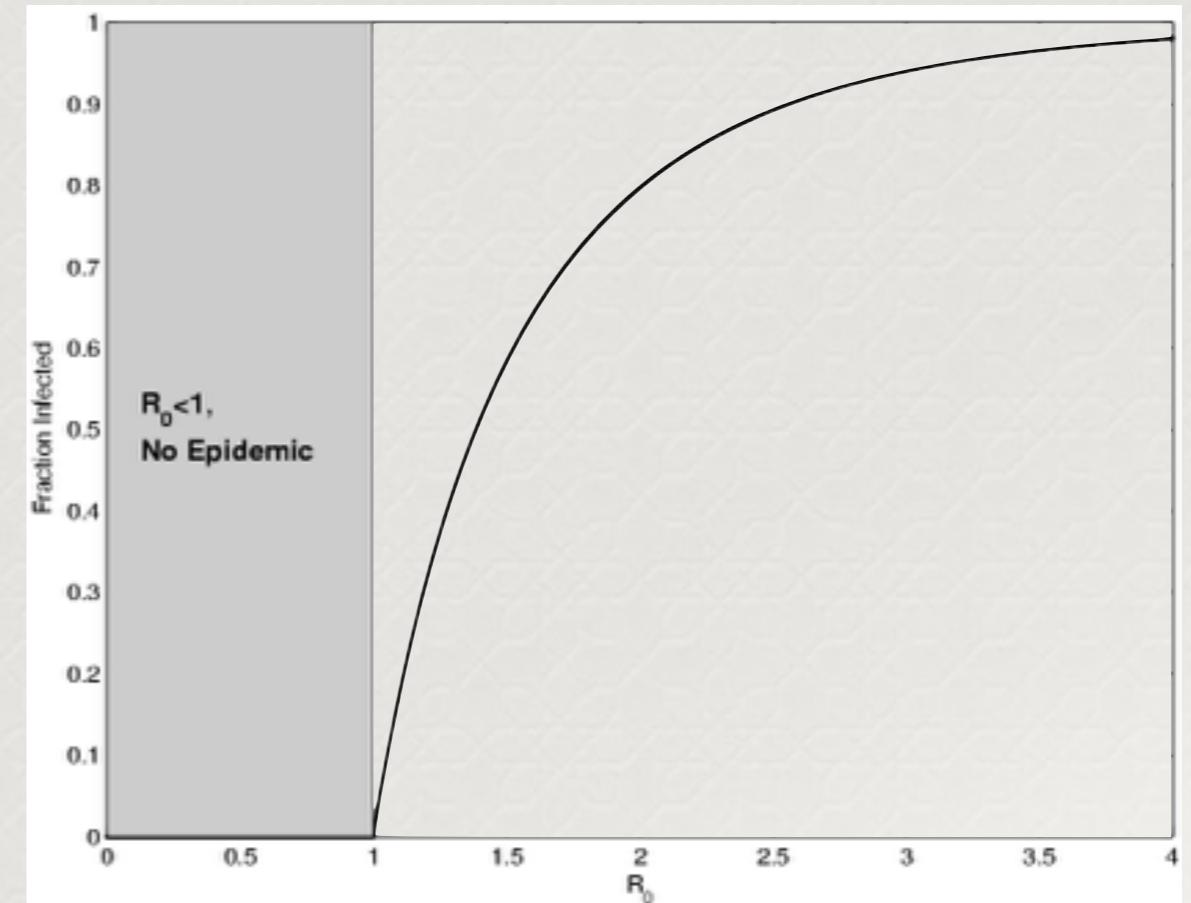
- ◆ When epidemic is over, by definition, we have $X(\infty)$, $Y(\infty) (=0)$, and $Z(\infty)$

- ◆ $X(\infty) = N - Z(\infty) = X(0) e^{-Z(\infty) R_o/N}$

The death of an epidemic

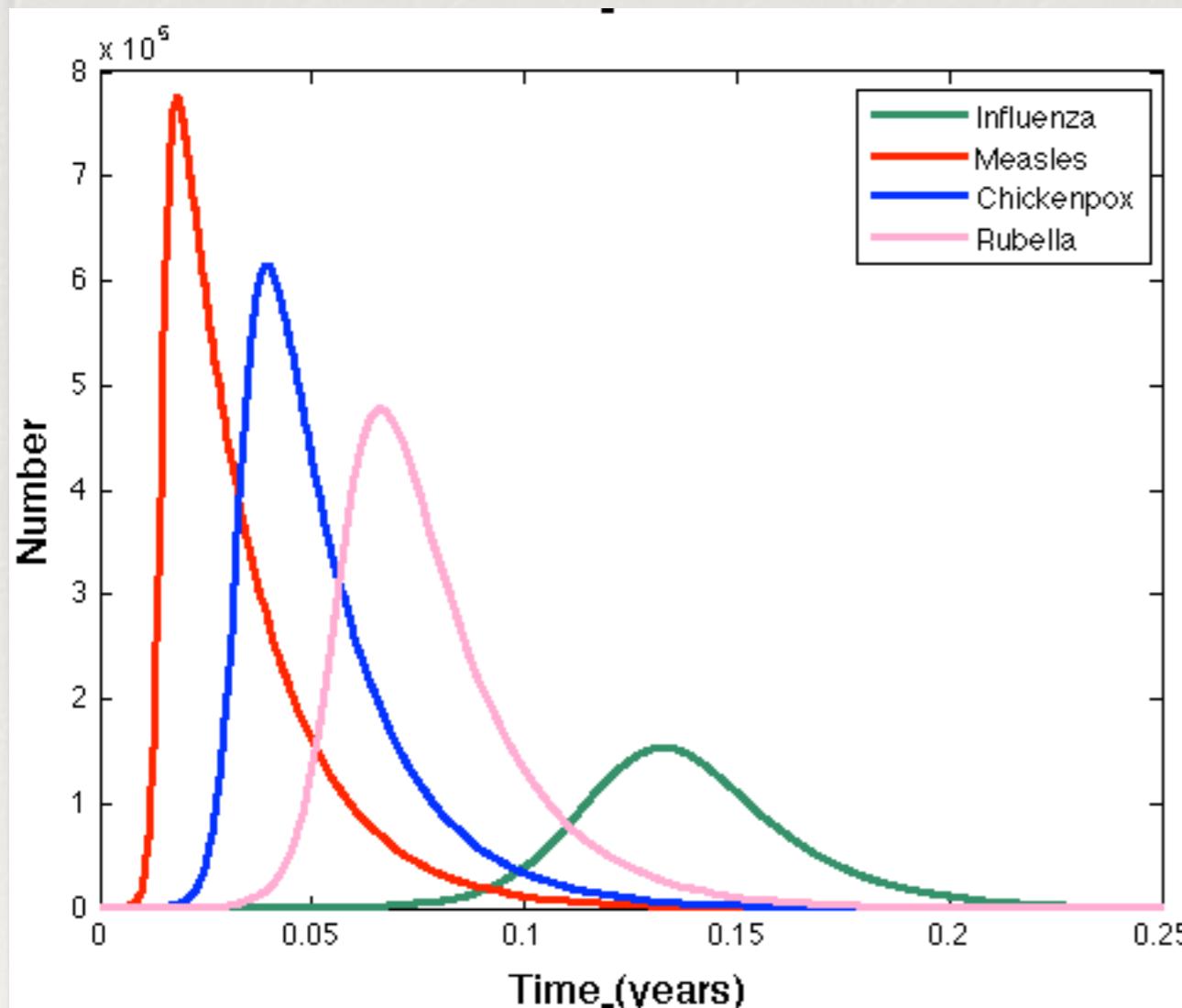
- So, $N - Z(\infty) - X(0) e^{-Z(\infty) R_0/N} = 0$
- Solve this numerically ('transcendental' equation)

Epidemic dies out because there are too few infectives, not because of too few susceptibles



Kermack & McKendrick (1927)

Simple Epidemics



	β	γ	R_0
"Measles"	886 yr^{-1}	0.019 yr	17
"Influenza"	180 yr^{-1}	0.011 yr	2
"Chickenpox"	315 yr^{-1}	0.022 yr	7
"Rubella"	200 yr^{-1}	0.025 yr	5

Frequency- or Density-Dependent Transmission?

- Assumed contact rate, κ , constant: ‘mixing’ is independent of population size: **frequency-dependent transmission**. Reasonable?
- If we assume contact rate to be κN (increases with ‘crowding’), then transmission rate is
 - $dX/dt = -\beta XY$
- Called **density-dependent transmission**



Does it Matter?

- ◆ Again, pathogen invasion if $dY/dt > 0$
- ◆ If initially everyone susceptible ($X=N$),
$$\beta NY - \gamma Y > 0 \Rightarrow Y(\beta N - \gamma) > 0$$
- ◆ In this case, we define $R_0 = \beta N / \gamma$, so need $R_0 > 1$
- ◆ Hence, for any particular β and γ , there's now a threshold population density required for invasion

Incorporating virulence

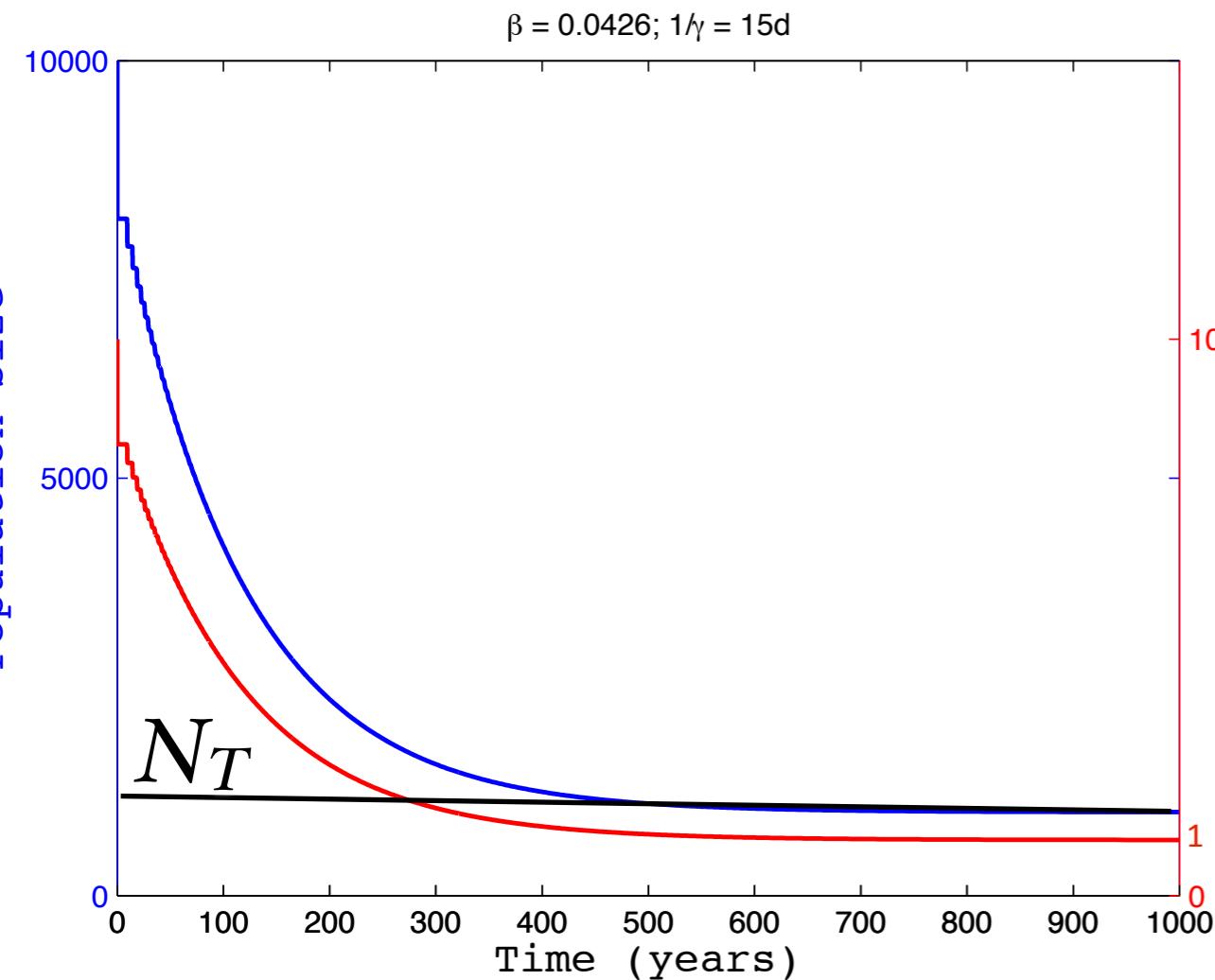
- ◆ *Assume infectious individuals die at rate α*

$$\frac{dY}{dt} = \dots - \gamma Y - \alpha Y$$

Transmission & R_0

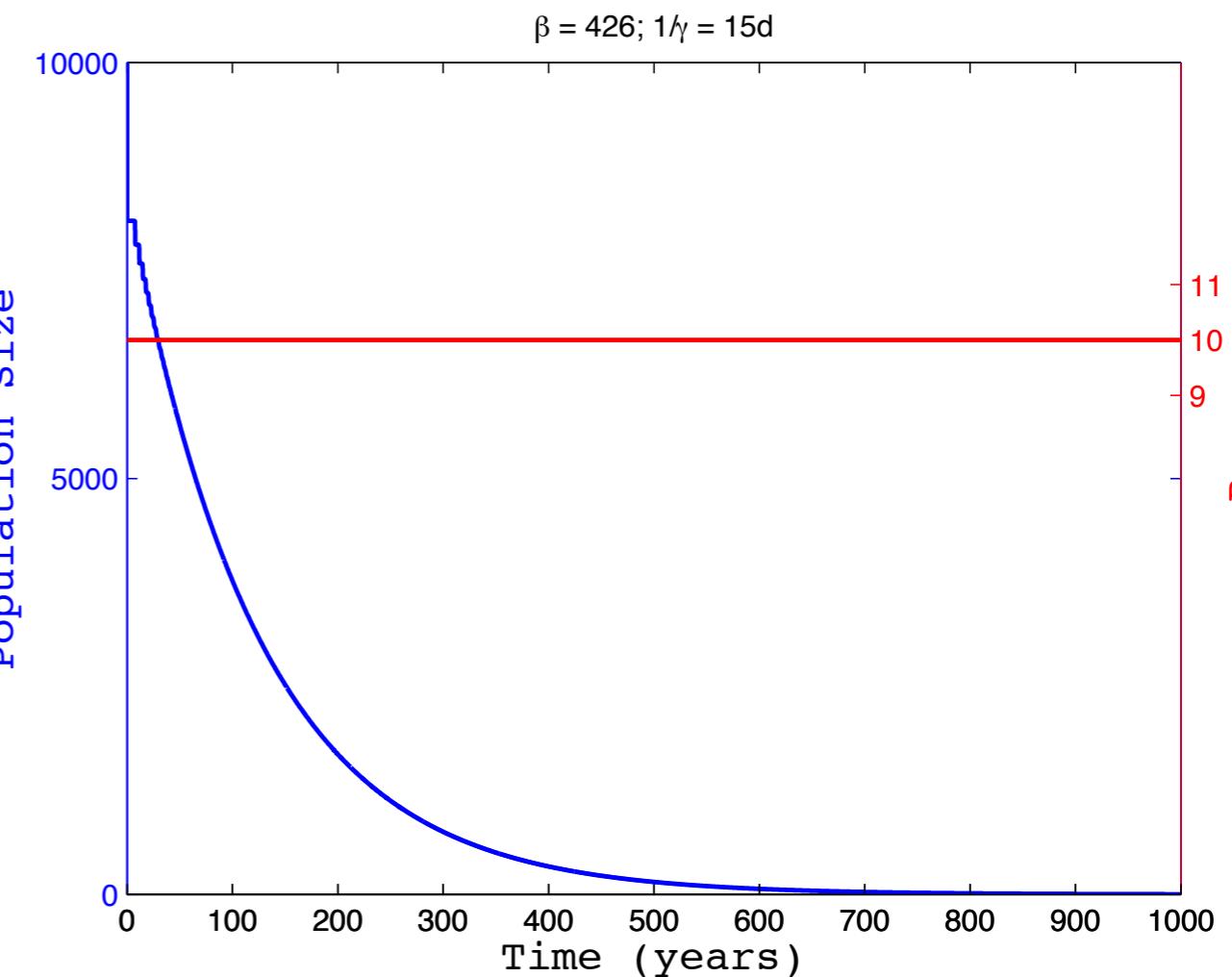
Density Dependent

$\beta=0.0426, \gamma=24, \alpha=18, \mu=0.02$
 $N_T = 1000$



Frequency Dependent

$\beta=426, \gamma=24, \alpha=18, \mu=0.02$
No invasion threshold



FD transmission \rightarrow pathogen can wipe out host

What should we do?

- ◆ If population size doesn't change, FD & DD equivalent ($\beta_{FD} = N \times \beta_{DD}$)
- ◆ Otherwise:
 - ◆ Frequency-dependence generally more appropriate in large populations with heterogeneous mixing, STDs, vector-borne pathogens
 - ◆ Density-dependence representative of wildlife & livestock diseases (especially with smaller population sizes)