

(Hidden) assumptions of simple compartmental ODE models

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MMED 2019

Goals

- Review the main uses of applied epidemiological modelling
- Introduce our conceptual framework for applied modelling
- Review commonly overlooked assumptions that are inherent in the structure of simple compartmental ODE models
- Discuss when these assumptions might be problematic, and when they may be desirable
- Begin to explore some alternative model structures that relax the assumptions

Applied Epi. Modelling

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graph TD; A[Applied Epi. Modelling] --> B[For application to the real world]; A --> C[Related to the distribution and determinants of health-related states and events]; A --> D[The use of simplification to represent the key components of something you're trying to understand more clearly];
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For application to the real world

Related to the distribution and determinants of health-related states and events

The use of simplification to represent the key components of something you're trying to understand more clearly

Applied Epi. Modelling



Applied Epi. Modelling

INSIGHT

- Improving understanding of the dynamics of health and disease
- Translation of results into decision-making and communication tools

Applied Epi. Modelling

INSIGHT

ESTIMATION

- Improving measurement and interpretation of key health indicators at the population and individual levels

Applied Epi. Modelling

INSIGHT

ESTIMATION

PREDICTION

- Projection and forecasting of expected future trends

Applied Epi. Modelling

INSIGHT

ESTIMATION

PREDICTION

PLANNING

- Guiding study design and intervention roll-out
- Informing decisions through analysis and comparison of policy scenarios

Applied Epi. Modelling

INSIGHT

ESTIMATION

PREDICTION

PLANNING

ASSESSMENT

- Evaluating the impact of public health interventions
 - Assessing risk of future public health events

Applied Epi. Modelling

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- Translation of results into decision-making and communication tools

ESTIMATION

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PREDICTION

- Projection and forecasting of expected future trends

PLANNING

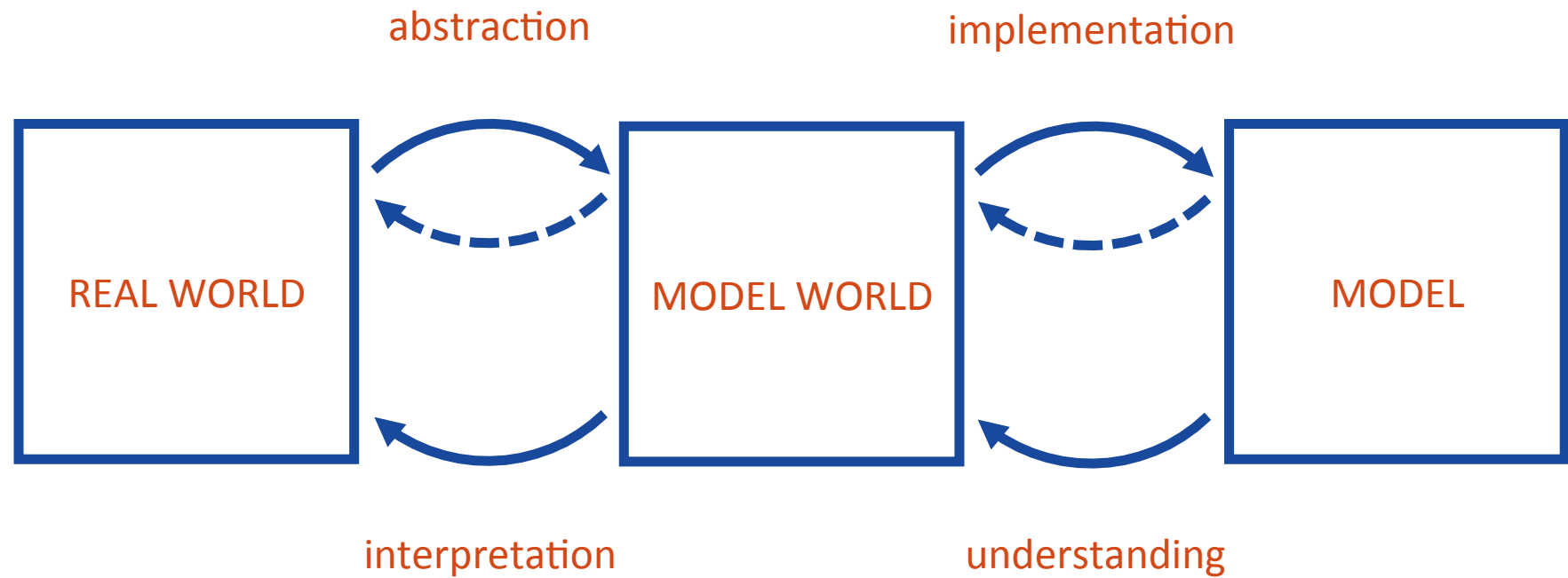
- Guiding study design and intervention roll-out
- Informing decisions through analysis and comparison of policy scenarios

ASSESSMENT

- Evaluating the impact of public health interventions
 - Assessing risk of future public health events

Goals

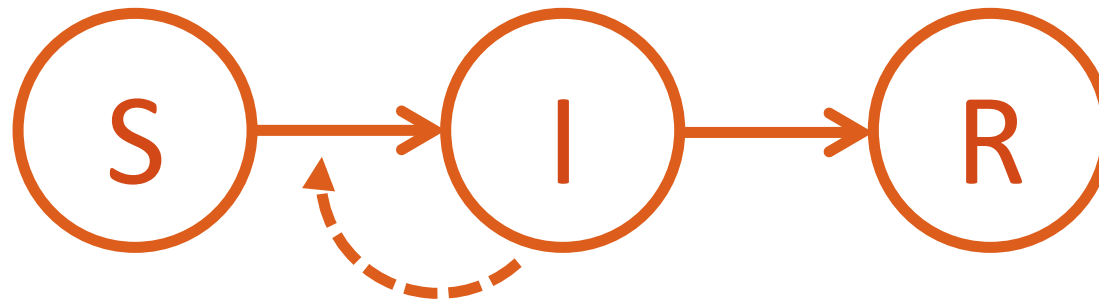
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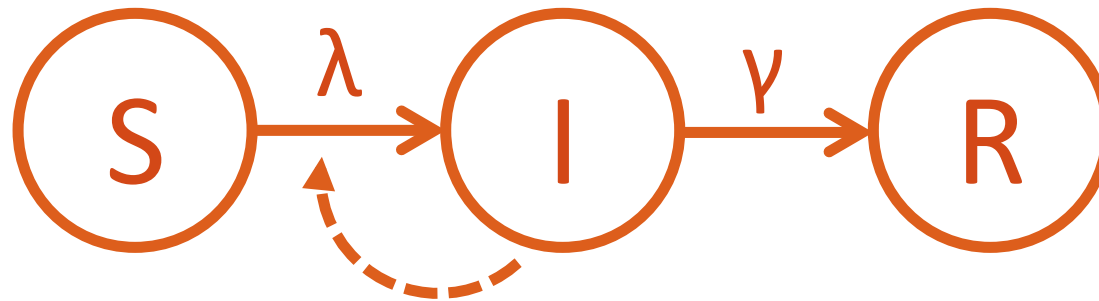
Model Worlds

- A **model world** is an abstraction of the world that is simple and fully specified, which we construct to help us understand particular aspects of the real world
- A **mathematical model** is formal description of the assumptions that define a model world
 - We know exactly what assumptions we've made, and we can follow those assumptions to their logical conclusions to address research questions

The SIR Model World



SIR: ODE Model



$$\lambda = \frac{\beta I}{N}$$

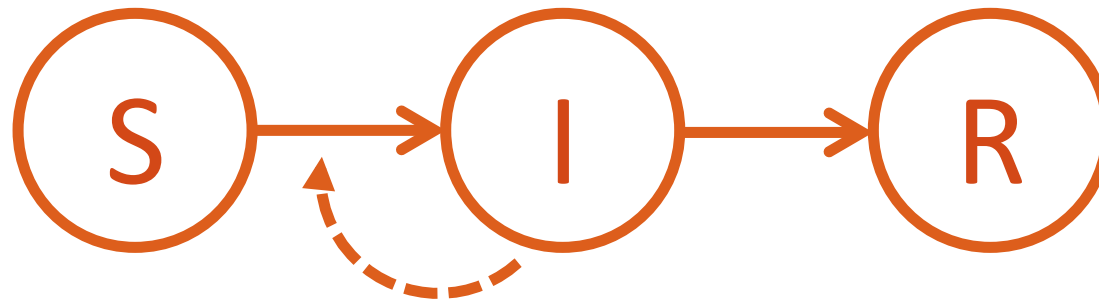
$$N = S + I + R$$

$$\frac{dS}{dt} = -\frac{\beta SI}{N}$$

$$\frac{dI}{dt} = \frac{\beta SI}{N} - \gamma I$$

$$\frac{dR}{dt} = \gamma I$$

SIR: Reed-Frost Model



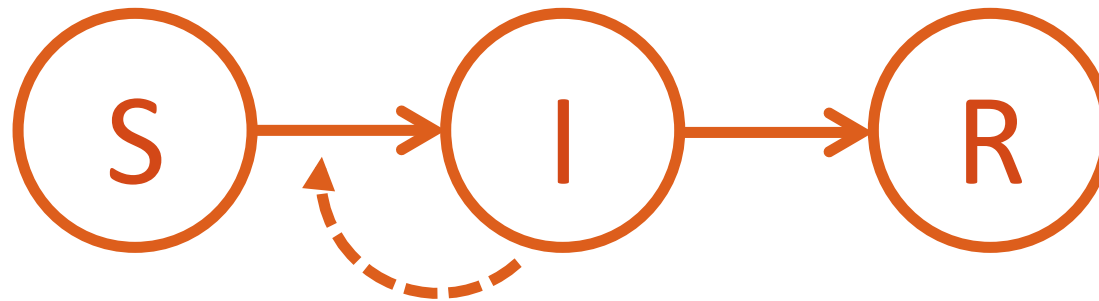
$$S_{t+1} = S_t - I_{t+1}$$

$$N = S + I + R$$

$$I_{t+1} = S_t(1 - q^{I_t})$$

$$R_{t+1} = R_t + I_t$$

SIR: Stochastic Reed-Frost

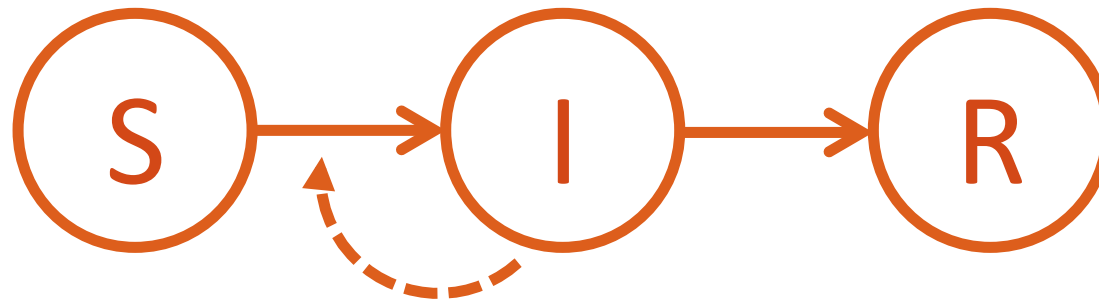


$$S_{t+1} = S_t - I_{t+1}$$

$$\mathbb{P}(I_{t+1} = x) = \binom{S_t}{x} (1 - q^{I_t})^x (q^{I_t})^{S_t - x}$$

$$R_{t+1} = R_t + I_t$$

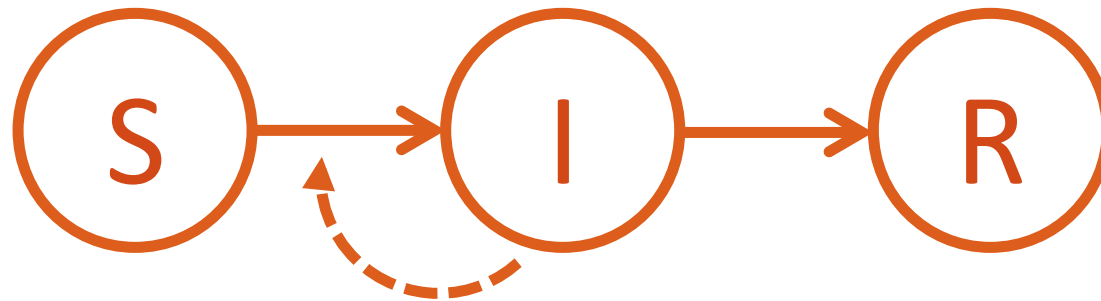
SIR: Chain Binomial Model



$$\begin{aligned} S_{t+\Delta t} &= S_t - X \\ I_{t+\Delta t} &= I_t + X - Y \\ R_{t+\Delta t} &= R_t + Y \end{aligned}$$

$$\begin{aligned} \mathbb{P}(X = x) &= \binom{S_t}{x} p^x (1 - p)^{S_t - x} \\ \mathbb{P}(Y = y) &= \binom{I_t}{y} r^y (1 - r)^{I_t - y} \end{aligned}$$

The SIR Model Family



A **mathematical model is formal description of the assumptions that define a **model world****

Taxonomy of compartmental models

CONTINUOUS TREATMENT OF INDIVIDUALS

(averages, proportions, or population densities)

DISCRETE TREATMENT OF INDIVIDUALS

DETERMINISTIC

CONTINUOUS TIME

- Ordinary differential equations
- Partial differential equations

DISCRETE TIME

- Difference equations
(eg, Reed-Frost type models)

STOCHASTIC

CONTINUOUS TIME

- Stochastic differential equations

DISCRETE TIME

- Stochastic difference equations

CONTINUOUS TIME

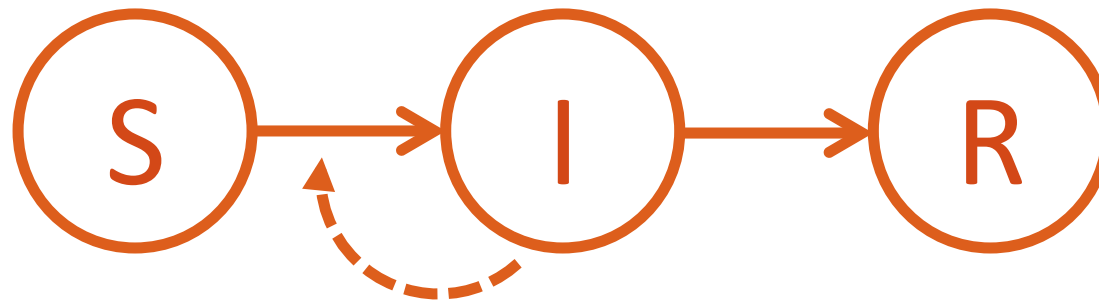
- Gillespie algorithm

DISCRETE TIME

- Chain binomial type models
(eg, Stochastic Reed-Frost models)

Taxonomy of compartmental models

What is a **compartmental** model?



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Taxonomy of compartmental models

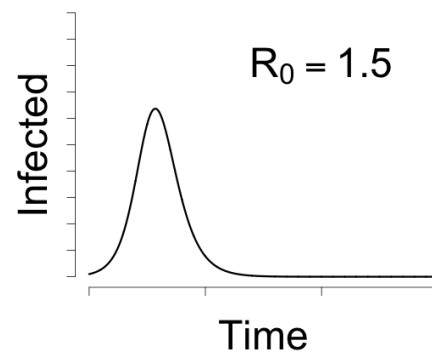
CONTINUOUS TREATMENT OF INDIVIDUALS

DISCRETE TREATMENT OF INDIVIDUALS

(averages, proportions, or population densities)

DETERMINISTIC

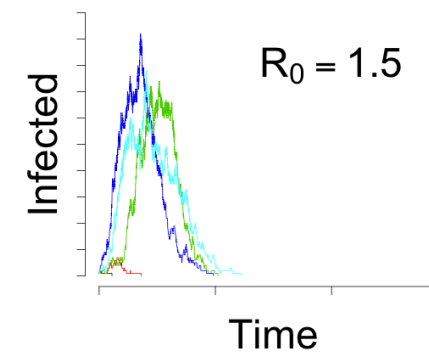
CONTINUOUS TIME



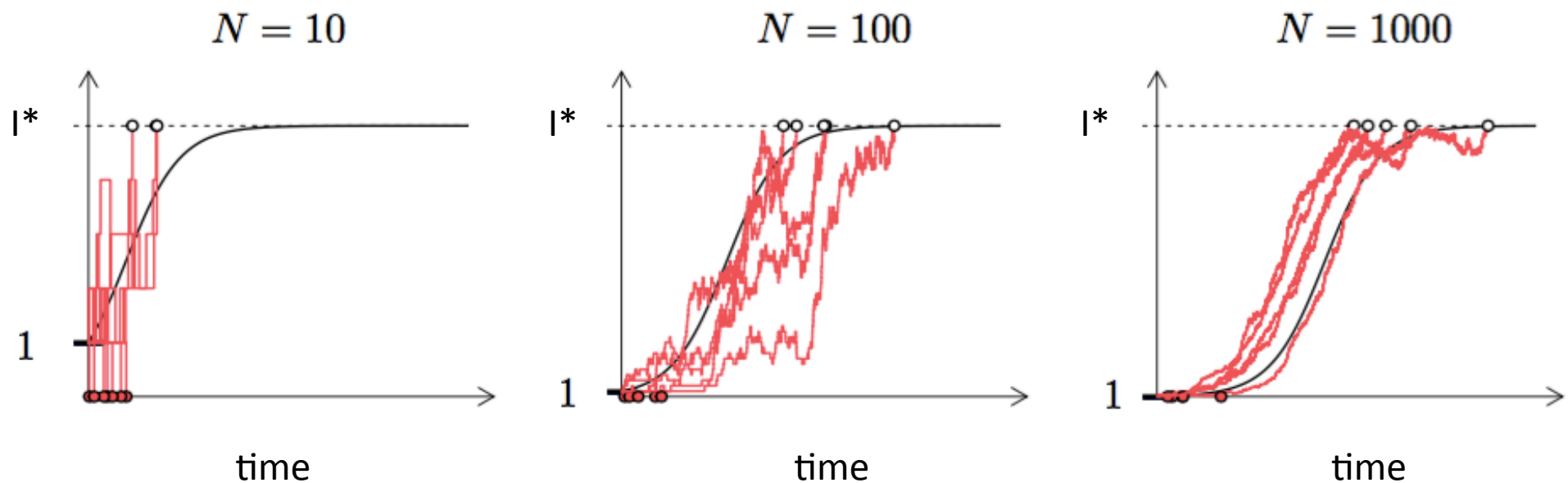
large population size

STOCHASTIC

CONTINUOUS TIME



Demographic stochasticity



Taxonomy of compartmental models

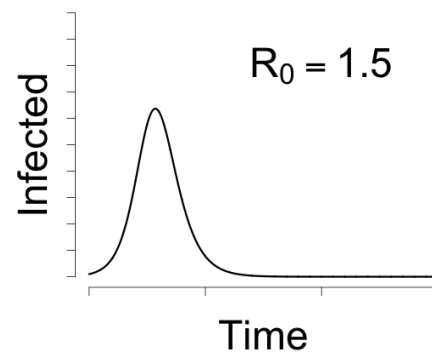
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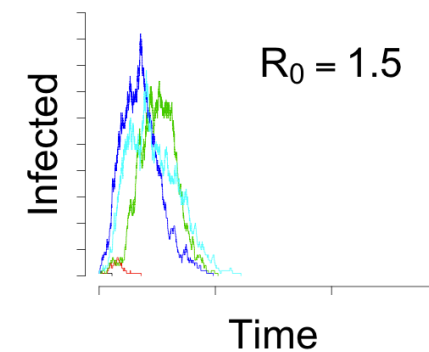
CONTINUOUS TIME



large population size

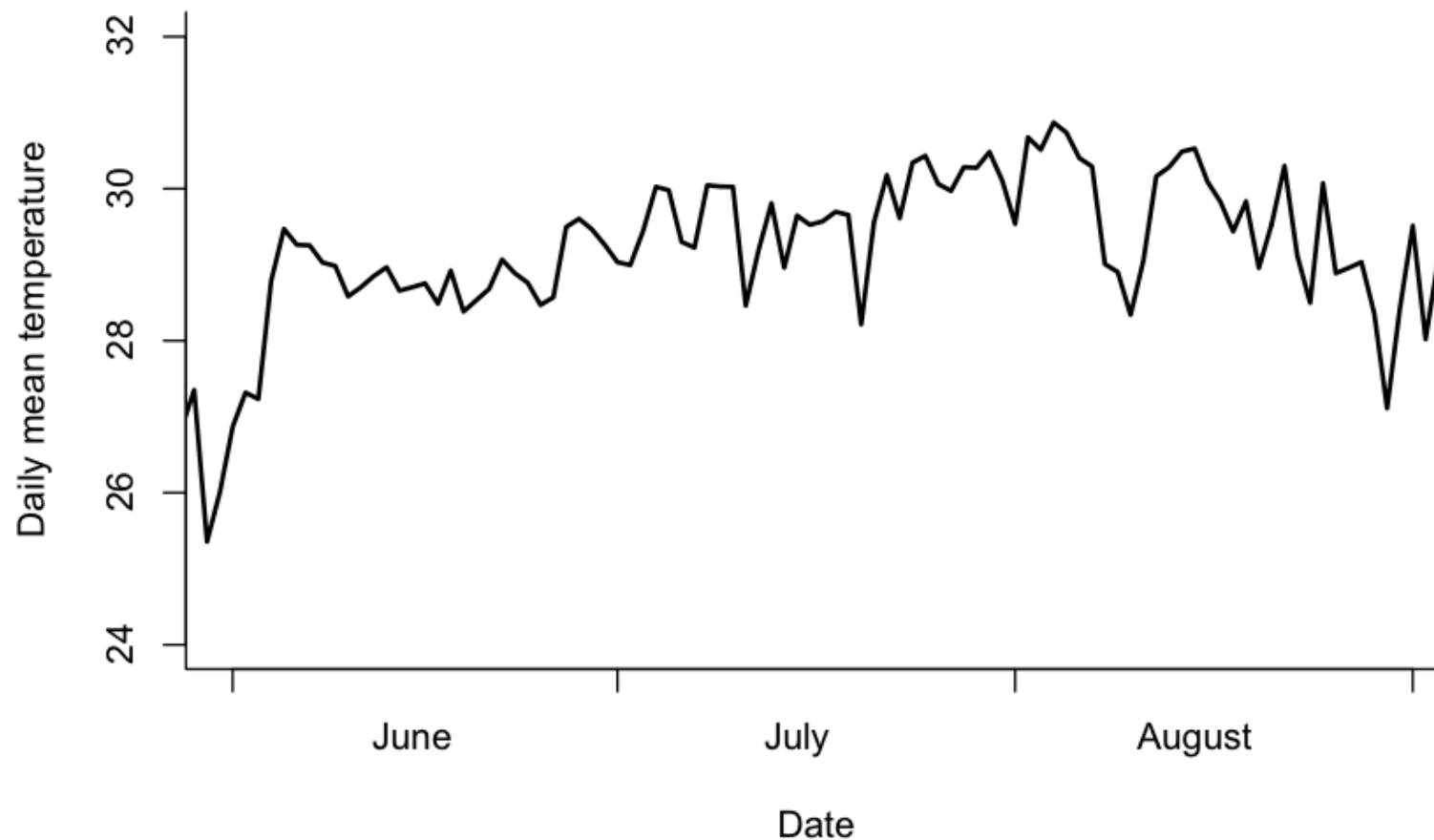
STOCHASTIC

CONTINUOUS TIME



Environmental stochasticity

Key West, 1993

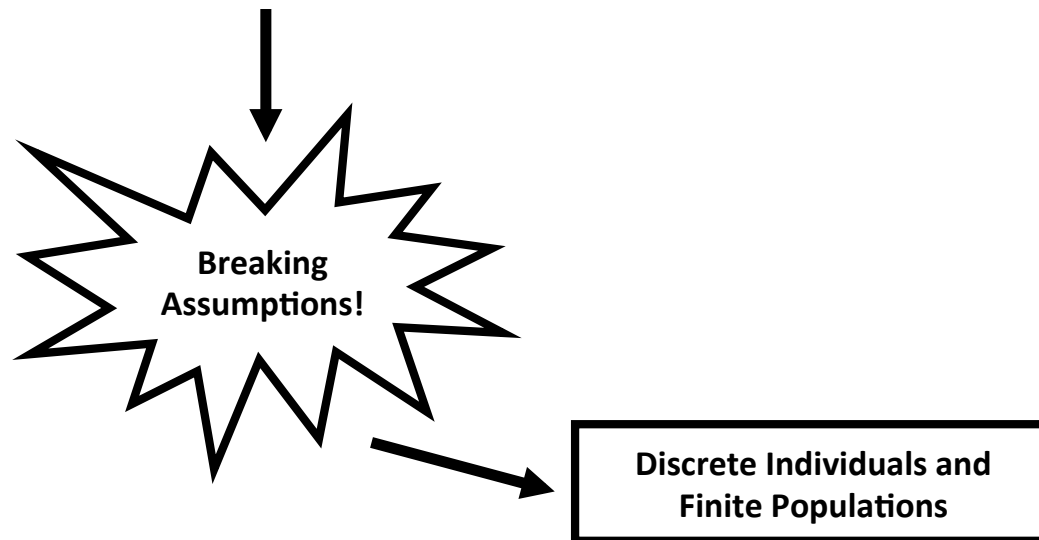


Compartmental ODE models assume

- Large population size
- Deterministic progression
 - for a given set of initial conditions and parameter values, a deterministic model always gives the same outcome
- Quickly assess what model outcomes are possible, and when

Compartmental ODE models assume

- Large population size
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 - for a given set of initial conditions and parameter values, a deterministic model always gives the same outcome



Taxonomy of compartmental models

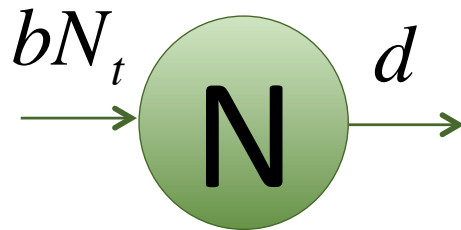
CONTINUOUS TREATMENT OF INDIVIDUALS
(averages, proportions, or population densities)

DISCRETE TREATMENT OF INDIVIDUALS

DETERMINISTIC

CONTINUOUS TIME

- Ordinary differential equations



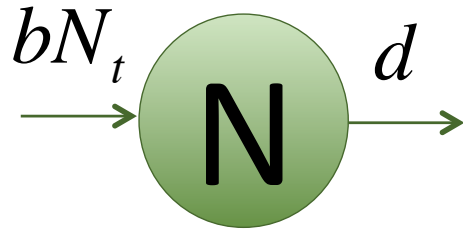
STOCHASTIC

Simple ODE models assume

- Time proceeds in a continuous manner
- Parameter values remain constant

CONTINUOUS TIME

- Ordinary differential equations



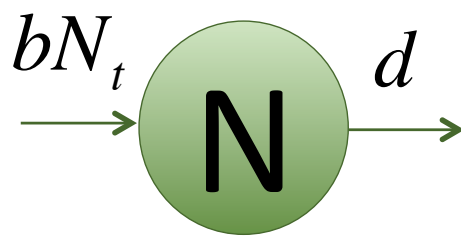
$$\frac{dN_t}{dt} = bN_t - dN_t$$

Simple ODE models assume

- Time proceeds in a continuous manner
- Parameter values remain constant

CONTINUOUS TIME

- Ordinary differential equations



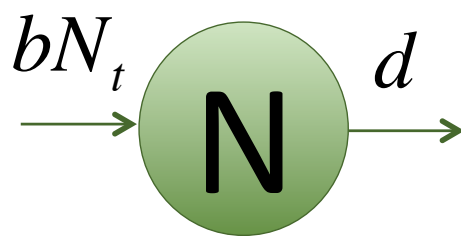
$$r = b - d$$

$$\frac{dN_t}{dt} = rN_t$$

$$N_{t+\Delta t} = N_t e^{r\Delta t}$$

CONTINUOUS TIME

- Ordinary differential equations



$$r = b - d$$

$$\frac{dN}{dt} = rN$$

DISCRETE TIME

- Difference equations

$$\lambda_{\Delta t} = e^{r\Delta t}$$

$$N_{t+\Delta t} = N_t e^{r\Delta t}$$

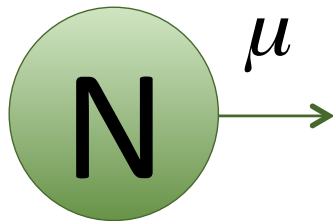
$$N_{t+\Delta t} = \lambda_{\Delta t} N_t$$

Simple compartmental ODE models assume

- Homogeneity within compartments
- Large population size
- Deterministic progression
- Time proceeds in a continuous manner
- Parameter values remain constant
- Memory-less processes

Simple ODE models assume

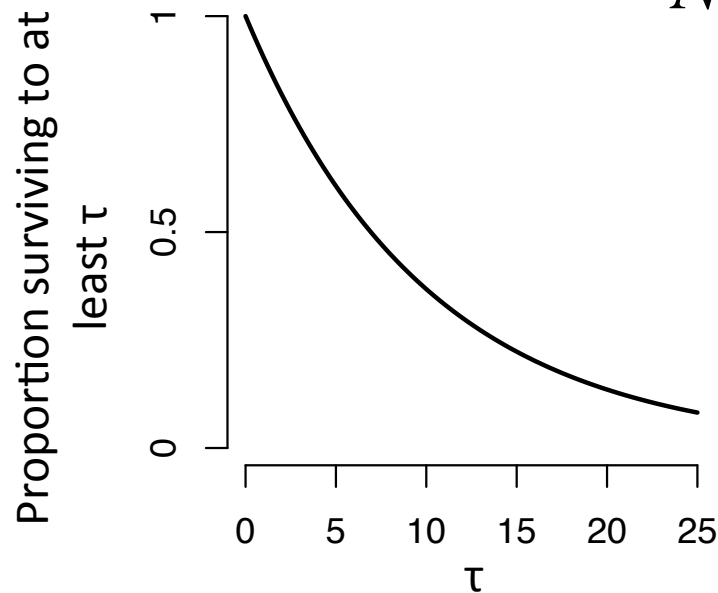
- Memory-less processes



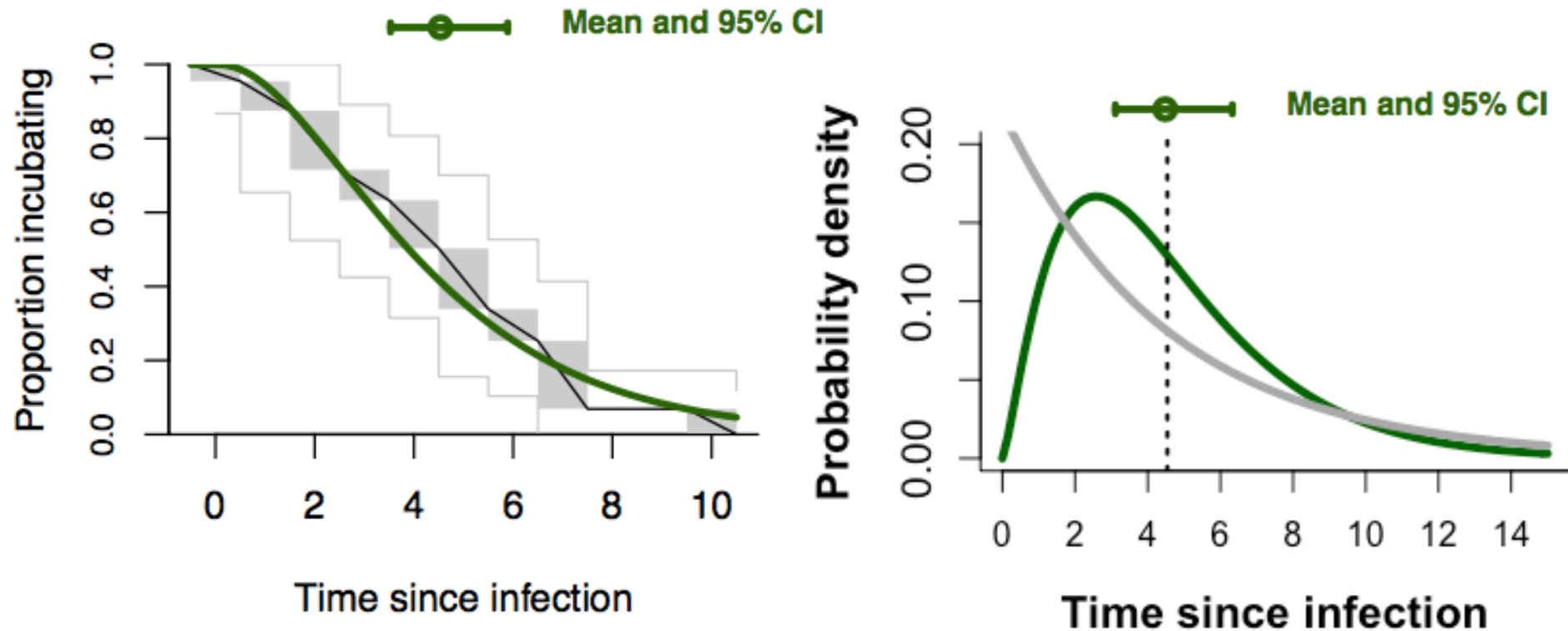
$$\frac{N_\tau}{N_0} = e^{-\mu\tau}$$

$$\frac{dN_t}{dt} = -\mu N_t$$

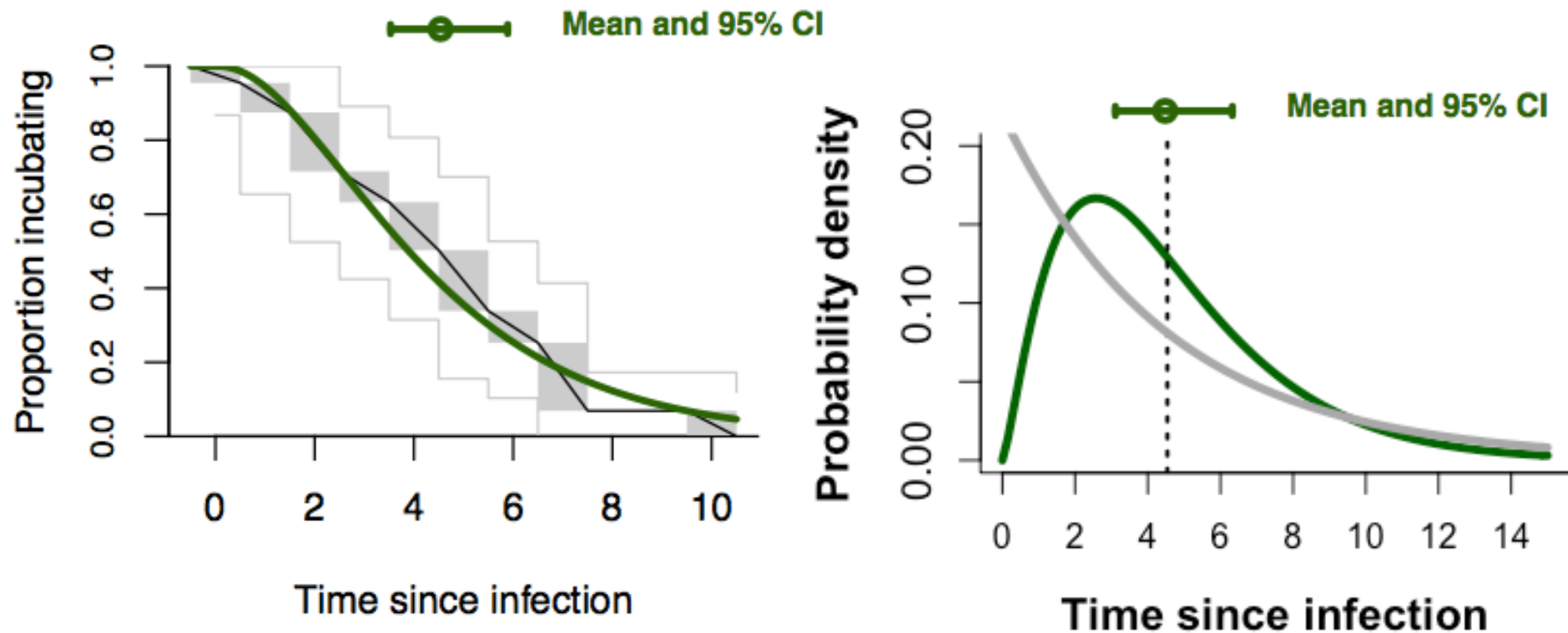
$$N_\tau = N_0 e^{-\mu\tau}$$



Realistic waiting times

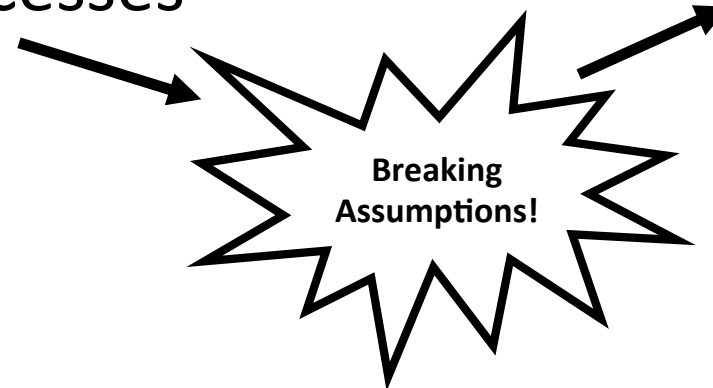


Realistic waiting times



Simple compartmental ODE models assume

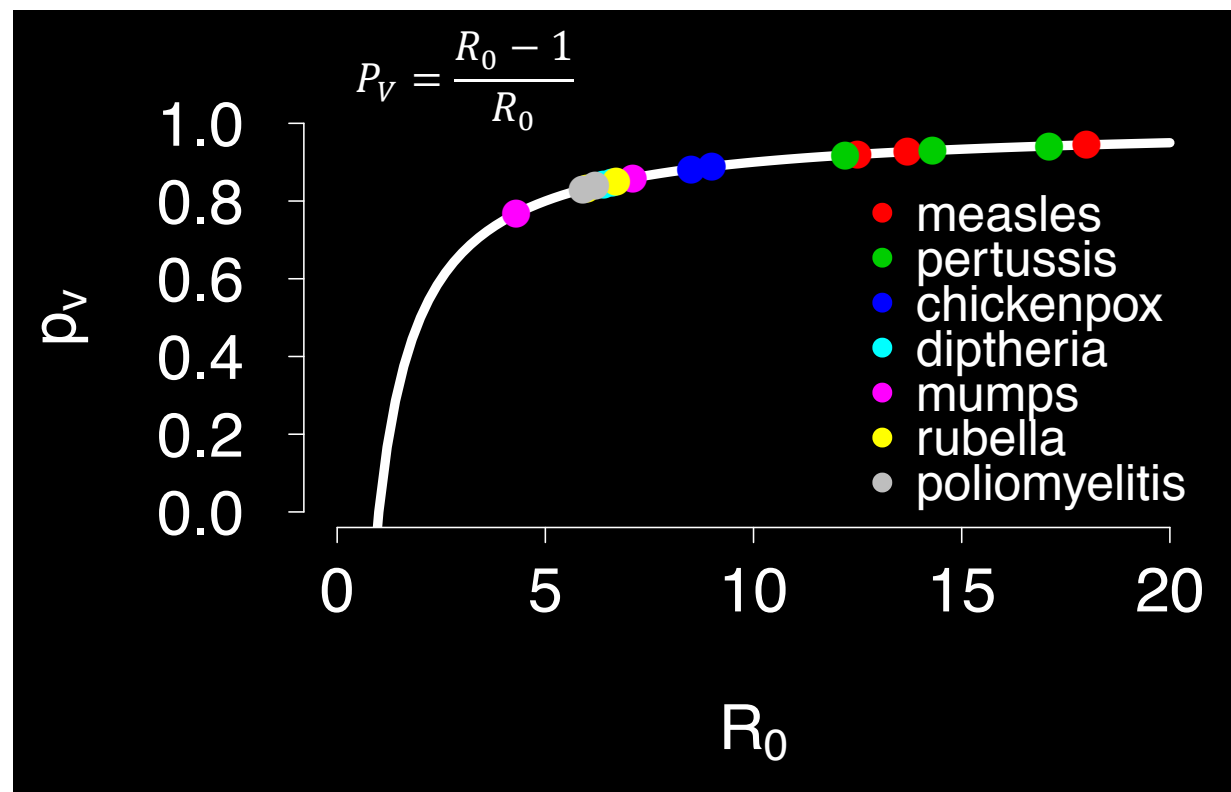
- Homogeneity within compartments
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Non-exponential waiting times

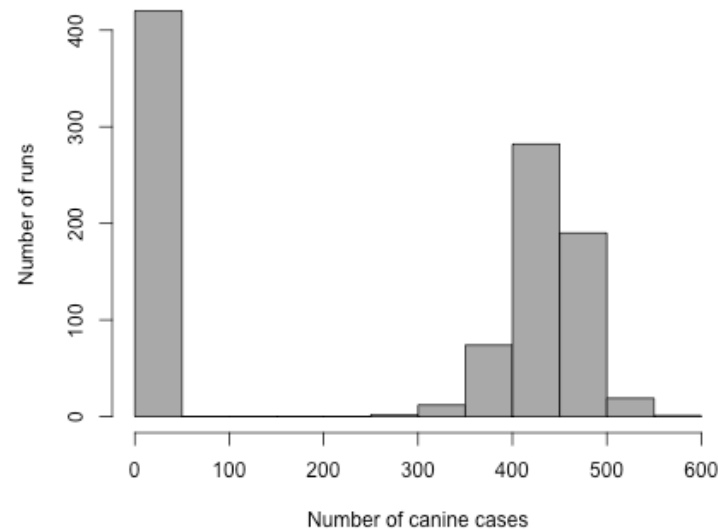
Summary

- Simple ODE models are important tools for building understanding



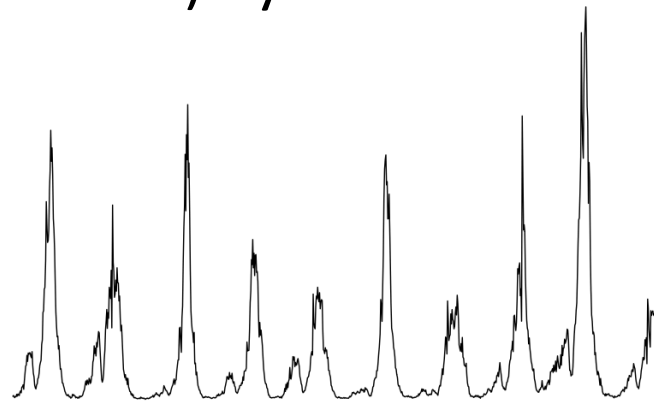
Summary

- Simple ODE models are important tools for building understanding
- It's important to recognize the assumptions built into these models
 - When populations are small, average behaviors can be misleading



Summary

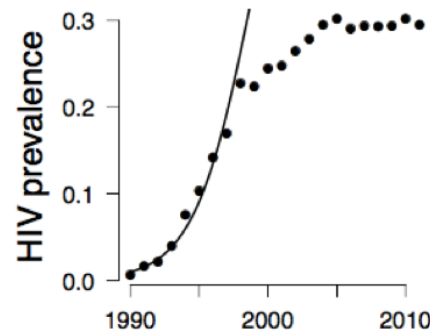
- Simple ODE models are important tools for building understanding
- It's important to recognize the assumptions built into these models
 - When populations are small, average behaviors can be misleading
 - When rates vary, simple ODEs can fail to reproduce important (observed) dynamics



Data from Earn et al. 2000 *Science*

Summary

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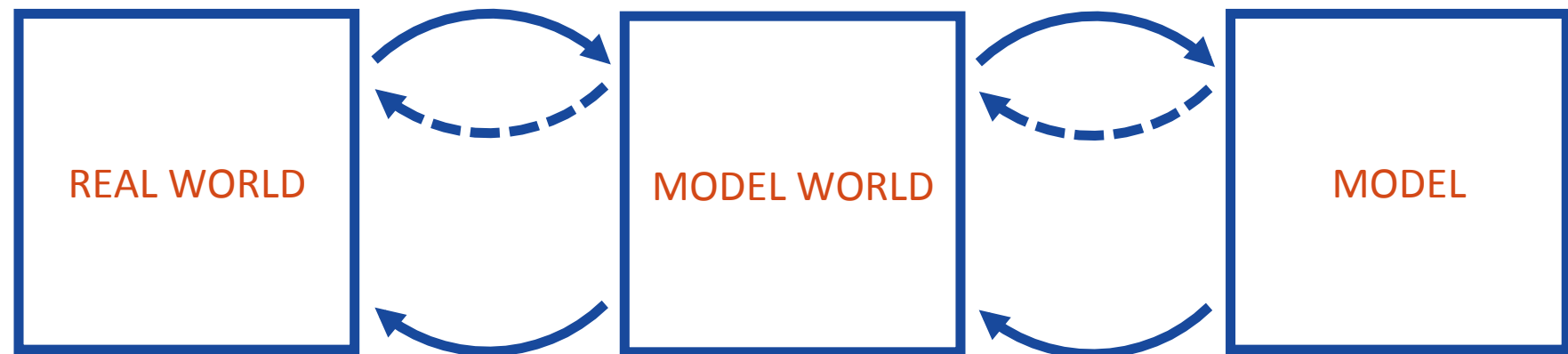
Dushoff lecture on heterogeneity (Wed)

Summary

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- It's important to recognize the assumptions built into these models
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Summary

- The applied epidemiological modelling process requires
 - abstraction
 - specification and implementation
 - gaining an understanding of the dynamics
 - interpretation





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(Hidden) assumptions of simple compartmental ODE models. DOI: [10.6084/m9.figshare.5044606](https://doi.org/10.6084/m9.figshare.5044606)

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Clinic on the Meaningful Modeling of Epidemiological Data

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