

Introduction to dynamical modeling

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2016 Summer Course on Mathematical Modeling and Analysis
of Infectious Diseases

National Taiwan University

Goals

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 - ▶ introduce the idea of dynamical modeling

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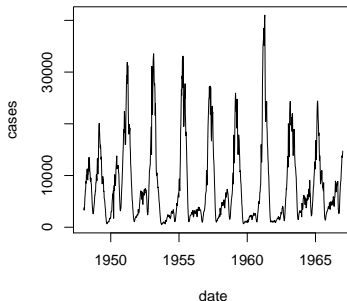
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Dynamic modeling connects scales



Measles reports from England and Wales

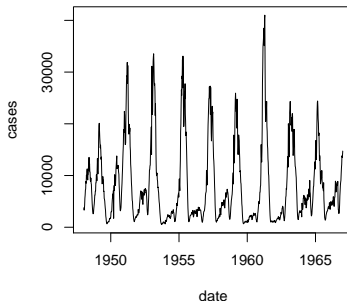


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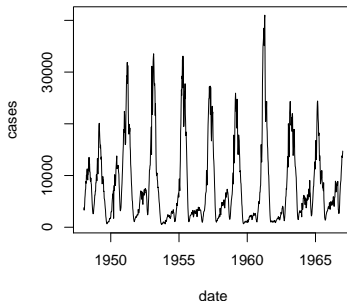


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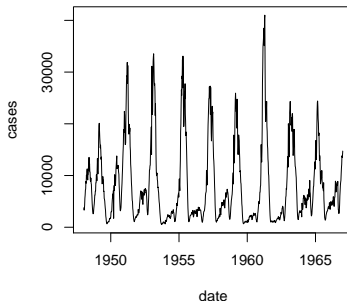


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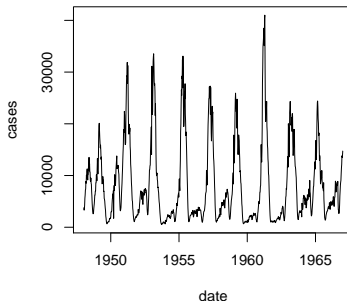


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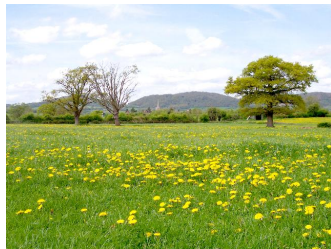
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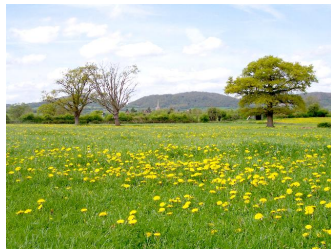
Example: Dandelions

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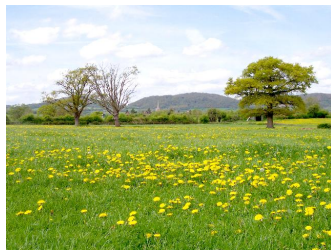
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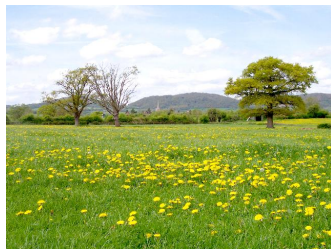
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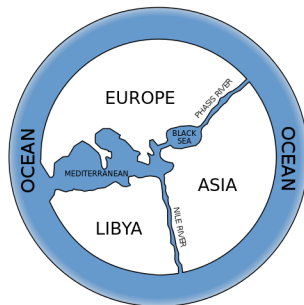
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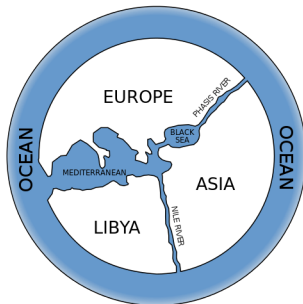
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- ▶ A dynamic model is based on a model world



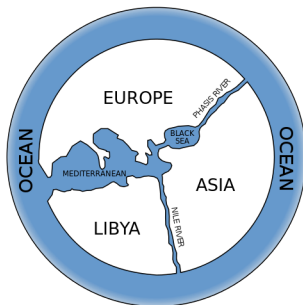
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- ▶ The model world has *enough* assumptions to allow us to calculate dynamics



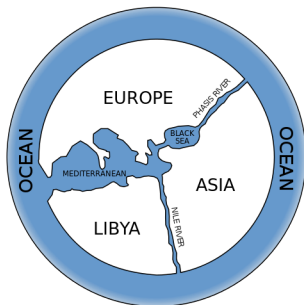
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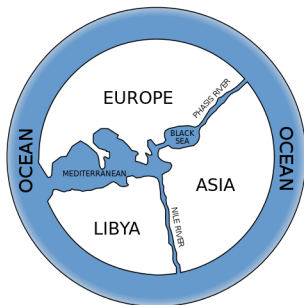
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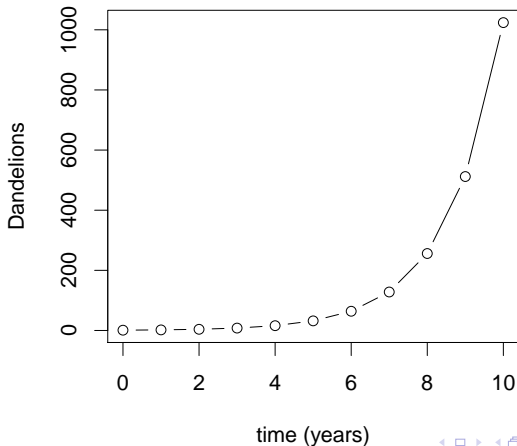
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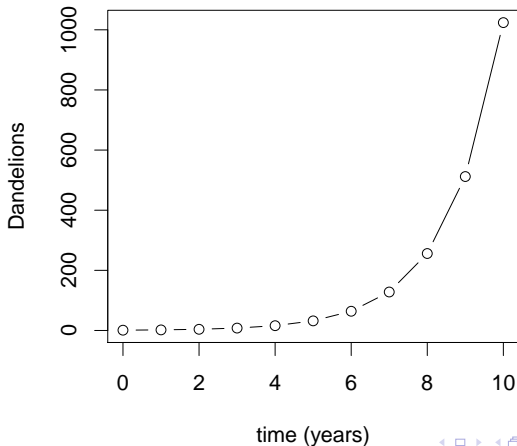
Model result

- If each individual is reproducing independently at each time step, the population changes *exponentially*



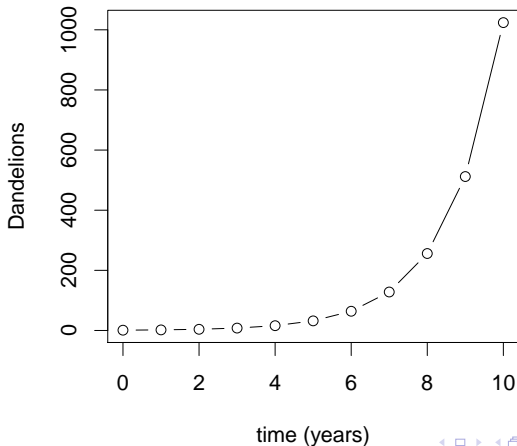
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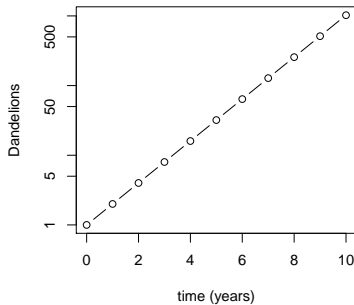
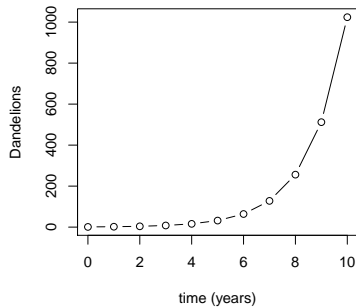
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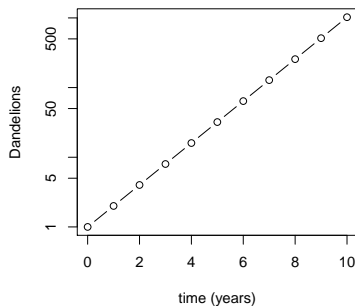
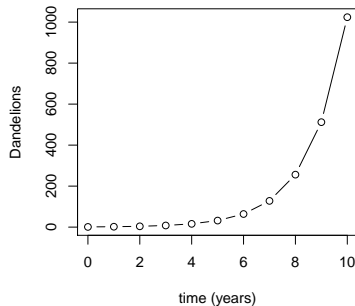
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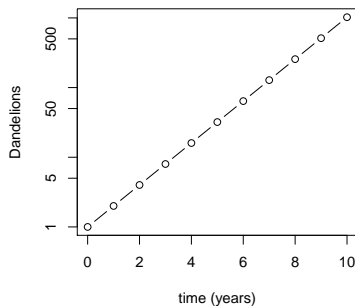
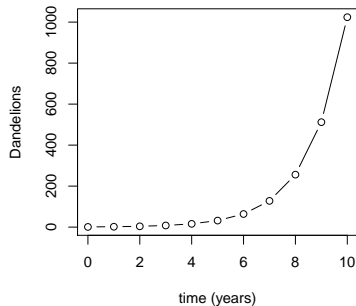
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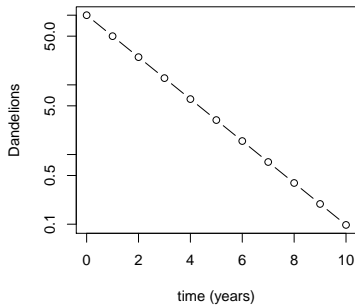
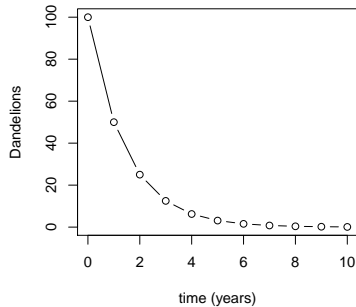
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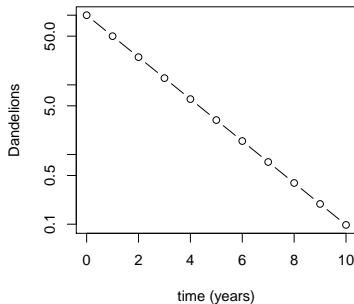
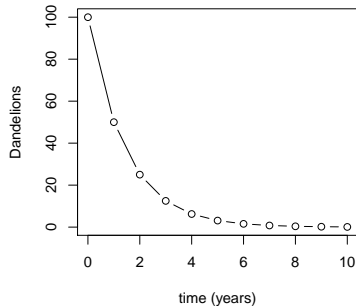
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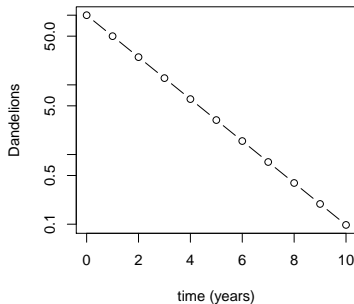
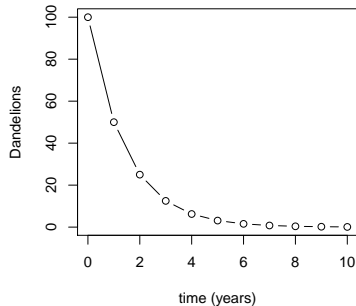
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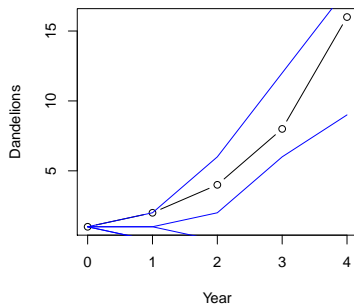
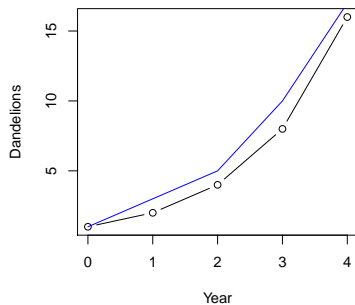
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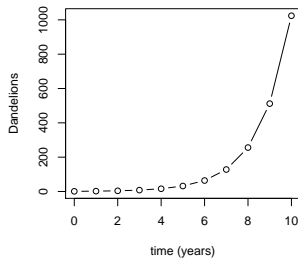
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Stochastic model



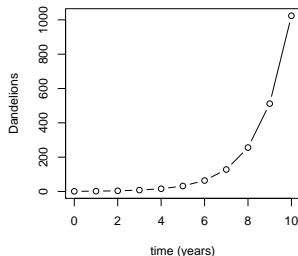
Time steps

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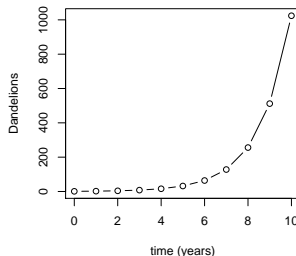
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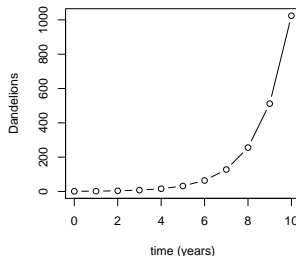
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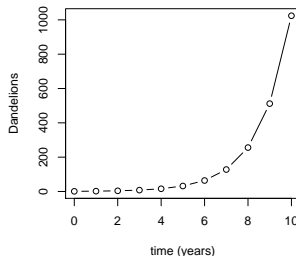
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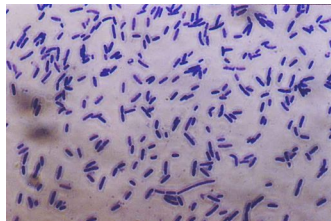
Bacteria

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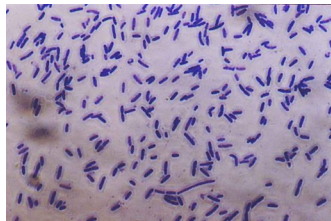
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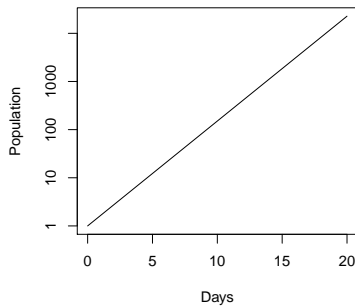
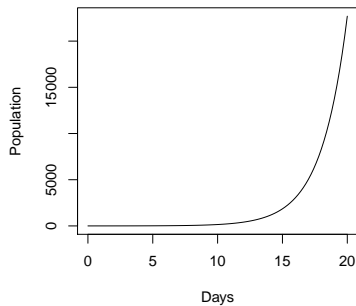
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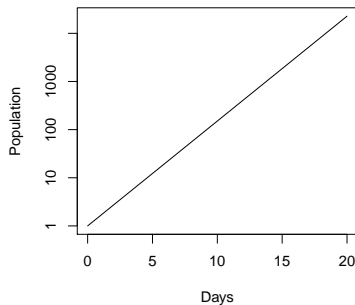
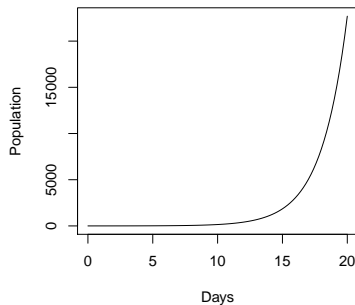
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Model result



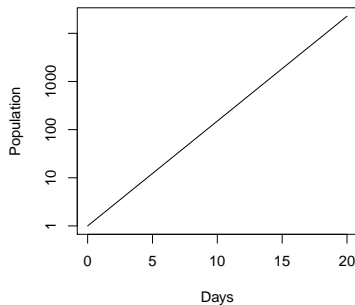
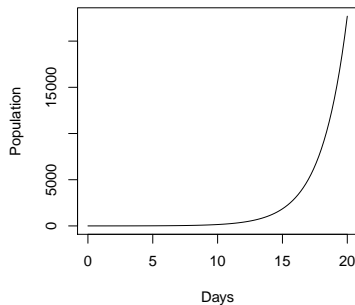
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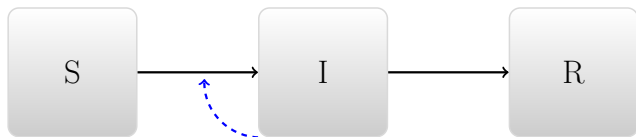
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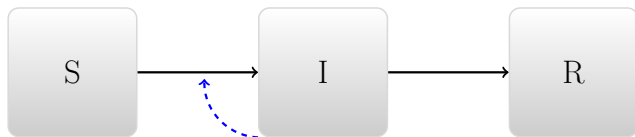
Simple models of disease spread

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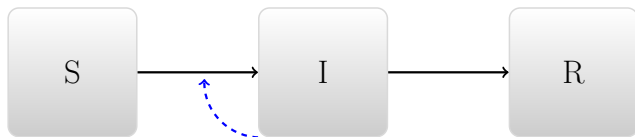
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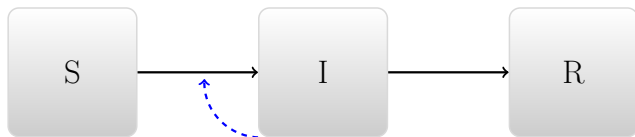
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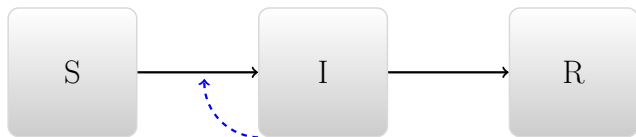
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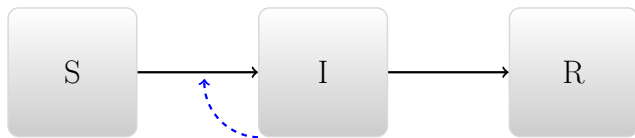
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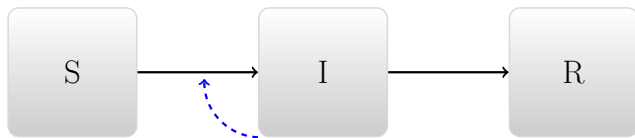
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What determines transition rates?



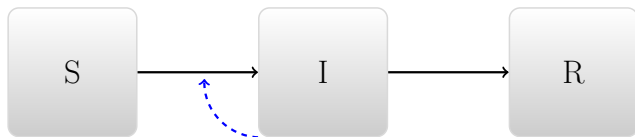
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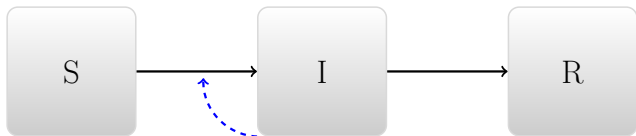
- ▶ People get better independently
- ▶ People get infected by infectious people

What determines transition rates?

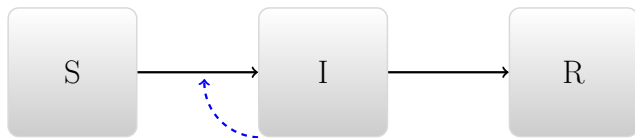


- ▶ People get better independently
- ▶ People get infected by infectious people

Conceptual modeling

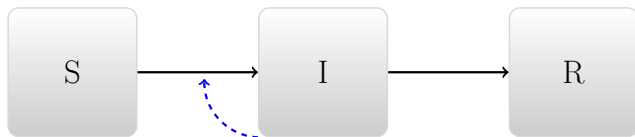


Conceptual modeling



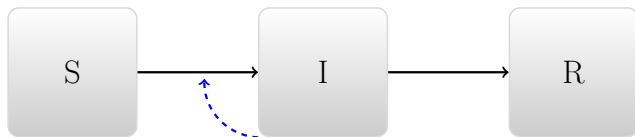
- What is the final result?

Conceptual modeling



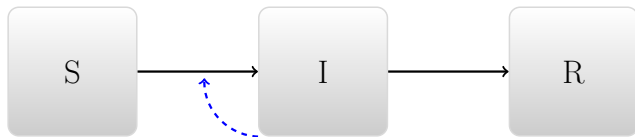
- ▶ What is the final result?
- ▶ When does disease increase, decrease?

Conceptual modeling



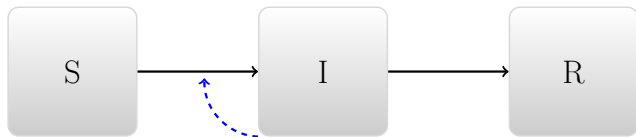
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Dynamic implementation



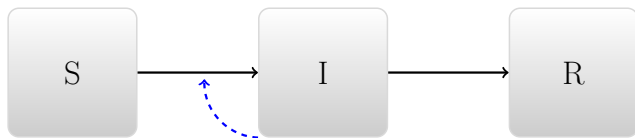
- Requires assumptions about time distributions

Dynamic implementation



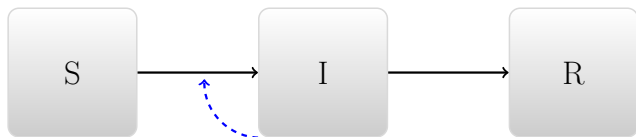
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Dynamic implementation



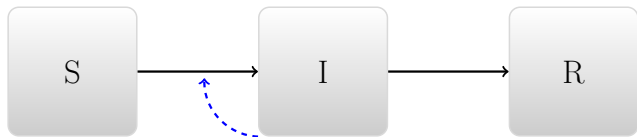
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Dynamic implementation



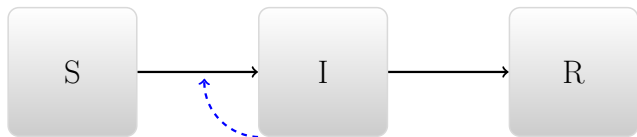
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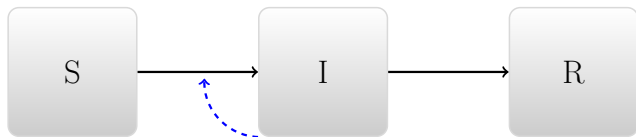
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Recovery



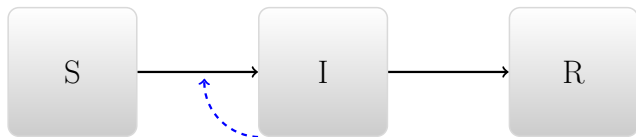
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Recovery



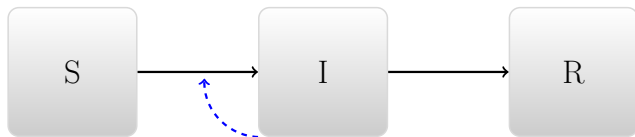
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Recovery



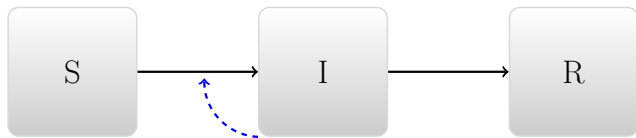
- ▶ Infectious people recover at *per capita* rate γ
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Recovery



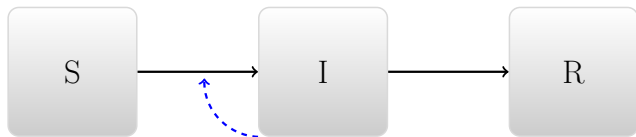
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Transmission



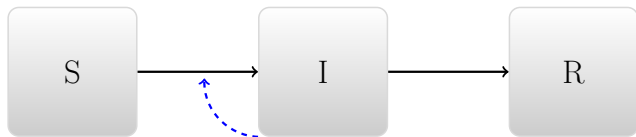
- Susceptible people get infected by:

Transmission



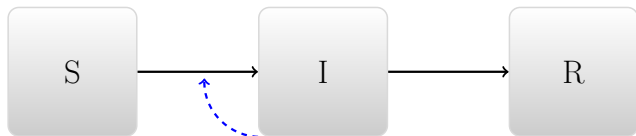
- ▶ Susceptible people get infected by:
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Transmission



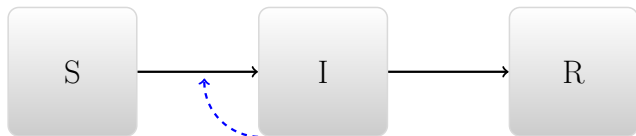
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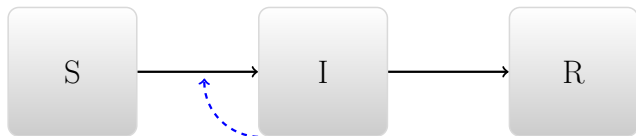
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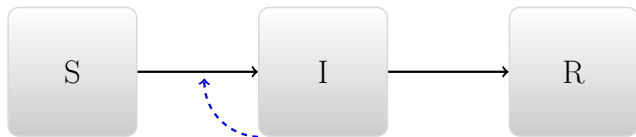
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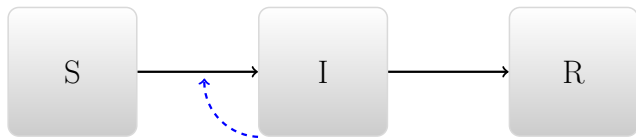
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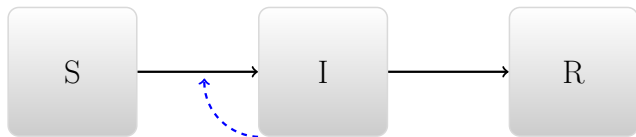
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Another perspective on transmission



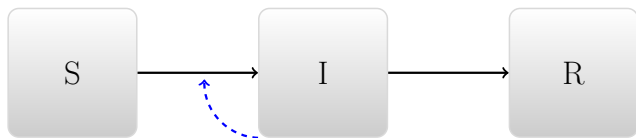
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Another perspective on transmission



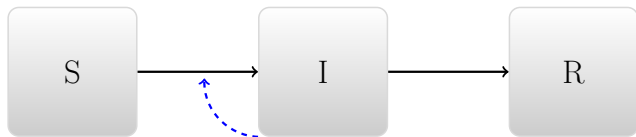
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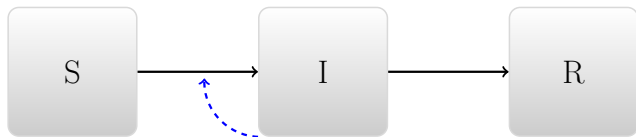
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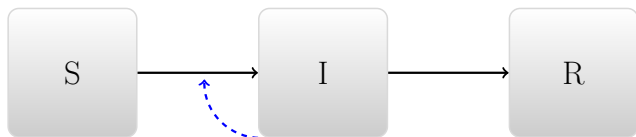
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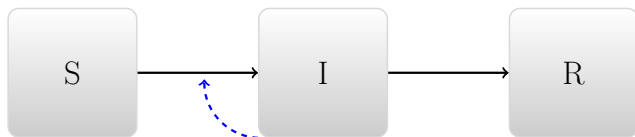
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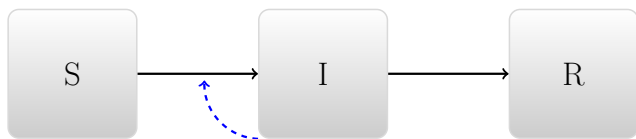
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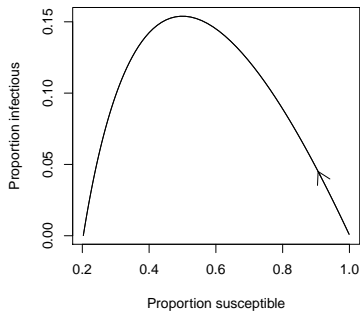
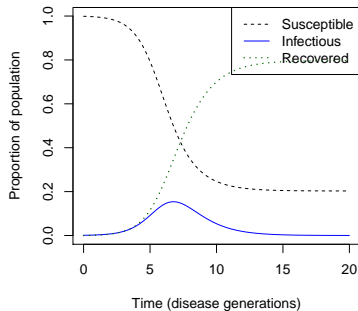
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ODE implementation



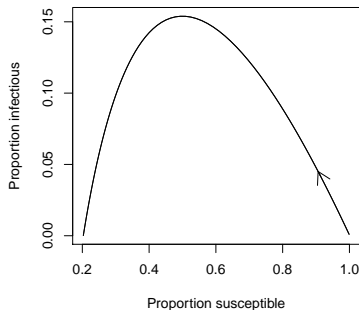
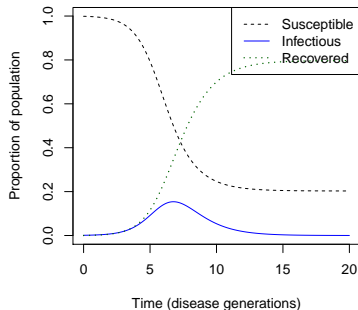
$$\begin{aligned}\frac{dS}{dt} &= -\beta \frac{SI}{N} \\ \frac{dI}{dt} &= \beta \frac{SI}{N} - \gamma I \\ \frac{dR}{dt} &= \gamma I\end{aligned}$$

ODE implementation



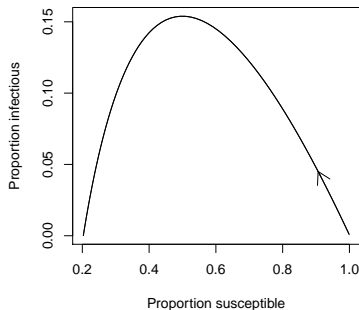
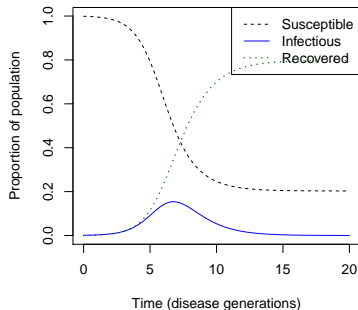
► Not everyone will get infected

ODE implementation



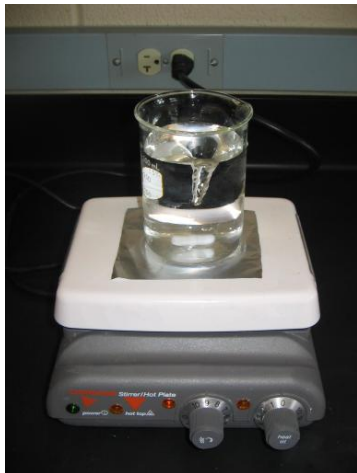
- ▶ Not everyone will get infected
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ODE implementation



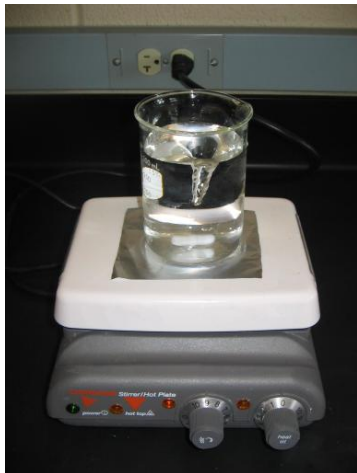
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ODE assumptions



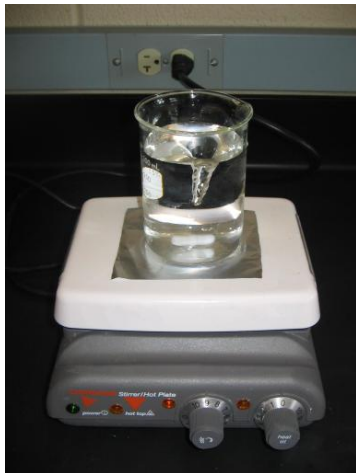
- Lots and lots of people

ODE assumptions



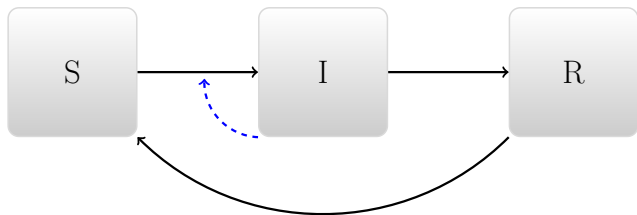
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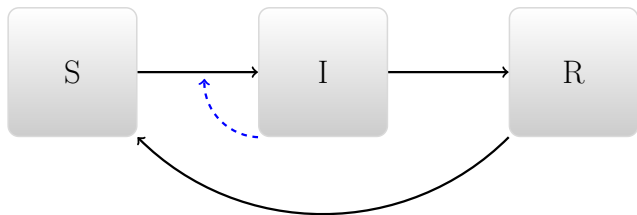
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Closing the circle



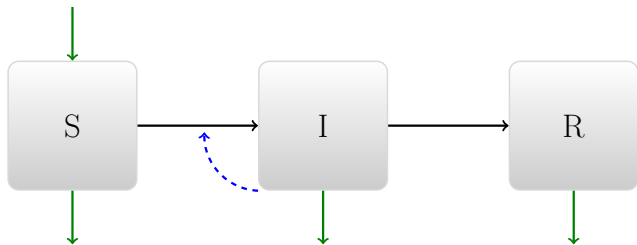
- Loss of immunity

Closing the circle



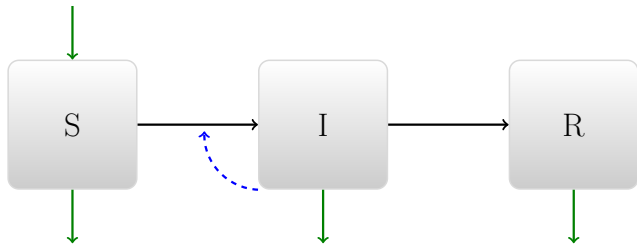
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Closing the circle



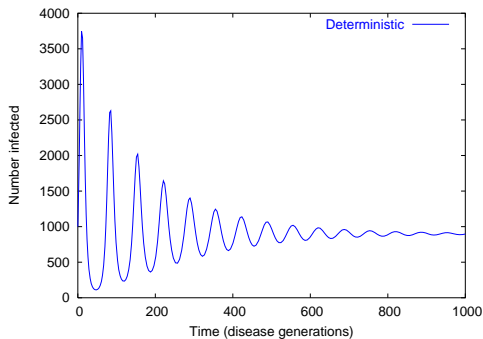
- Births and deaths

Closing the circle



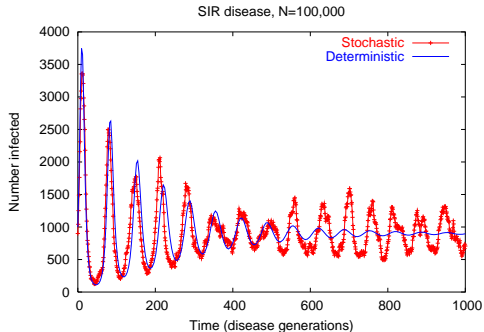
- Births and deaths

Tendency to oscillate



Modeling individuals as individuals usually requires a *stochastic* model

With individuality



Even in the simplest form, this can cause large random oscillations even in large populations

Types

- ▶ **Discrete** vs. **Continuous** time steps

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