

BIO 512

Digital Epidemiology

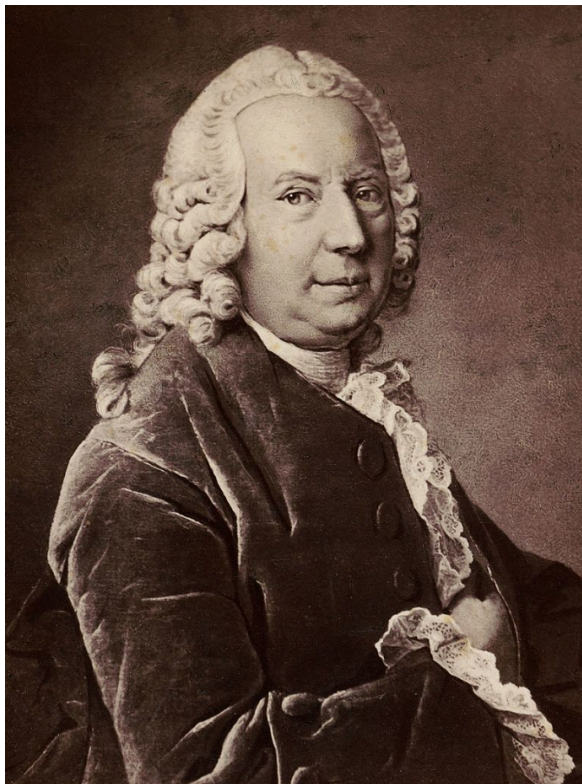
Modeling Infectious Diseases

Learning Objectives

- Develop basic infectious disease models and understand their dynamics
- S(E)IR(S) models
- closed vs. open models
- seasonality
- stochastic vs. deterministic models

Modeling Infectious Diseases

History



- Daniel Bernoulli, 1766: first mathematical model (specific to smallpox variolation)
- 1917/18: Ross & Hudson, first general epidemiological models
- 1927: Kermack & McKendrick, first SIR model

Modeling Infectious Diseases

A Basic SIR Model



The diagram illustrates the three compartments of a Basic SIR model. It consists of three rounded rectangular boxes arranged horizontally. The first box on the left is blue and labeled 'Susceptible'. The middle box is red and labeled 'Infected'. The third box on the right is yellow and labeled 'Recovered'. All boxes have a thin black border.

Susceptible

Infected

Recovered

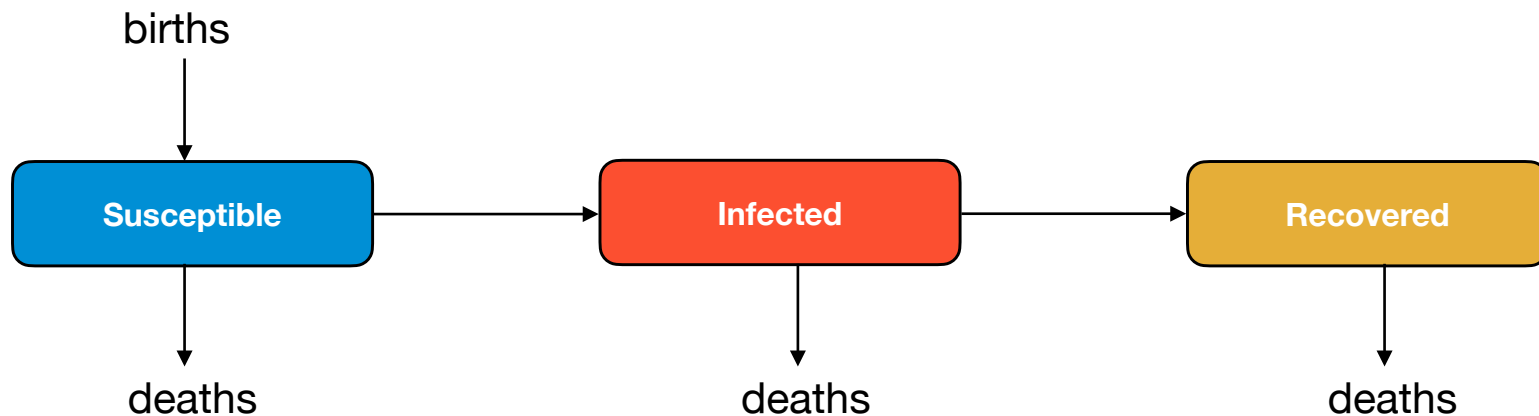
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A Basic SIR Model



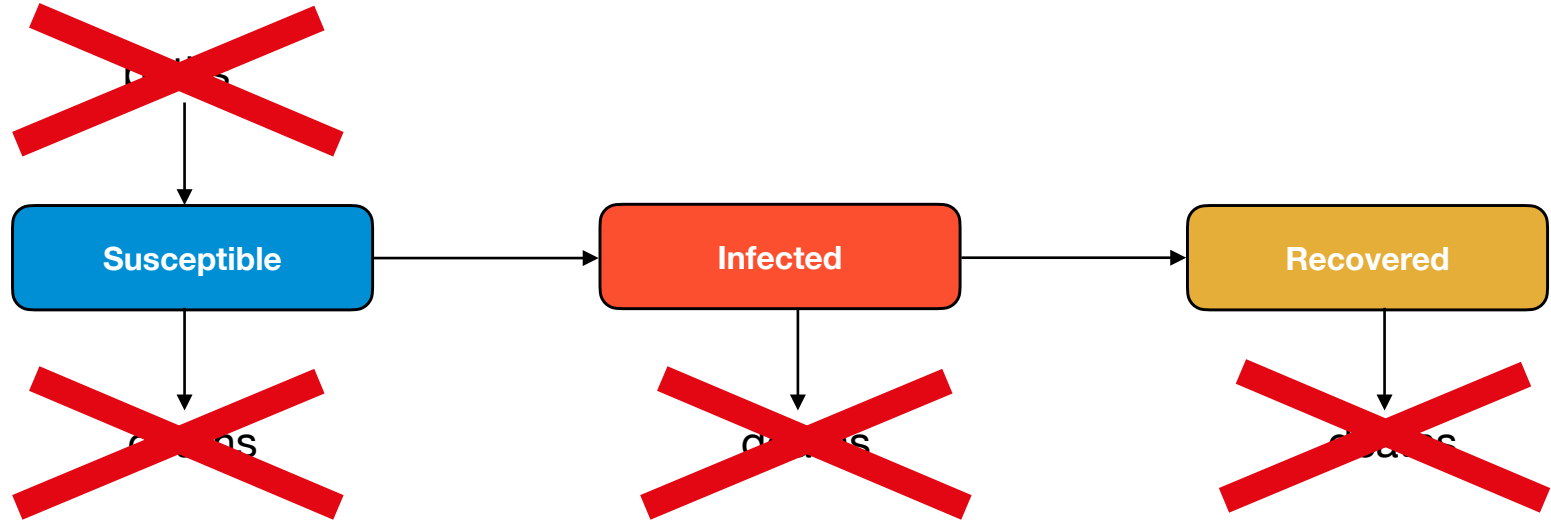
Modeling Infectious Diseases

A Basic SIR Model



Modeling Infectious Diseases

A Basic SIR Model



Modeling Infectious Diseases

A Basic SIR Model



β : per capita contact and disease transmission rate per unit time

γ : per capita recovery rate per unit time

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A Basic SIR Model

$$\frac{dS}{dt} = -\beta I \frac{S}{N}$$

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

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

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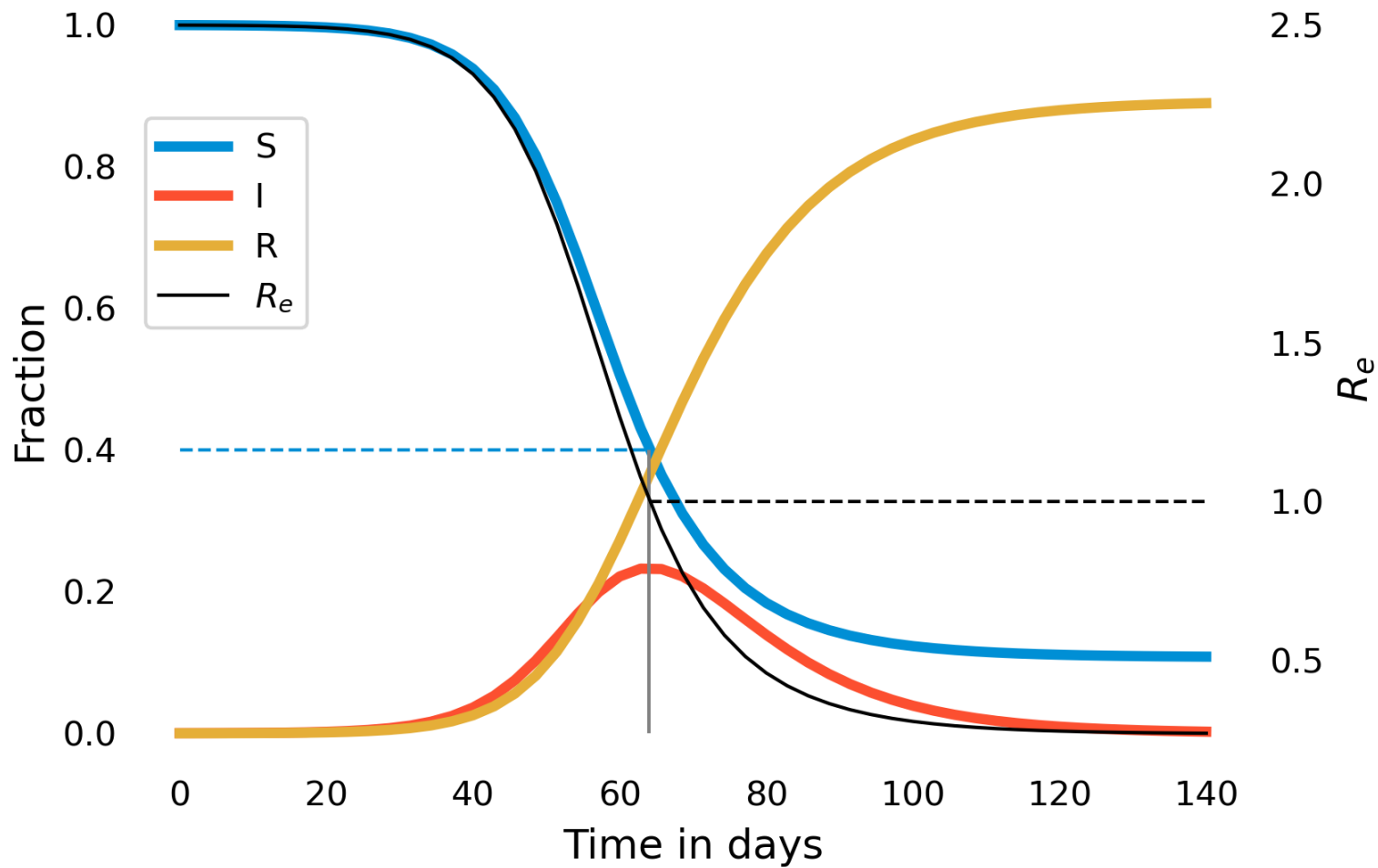
A Basic SIR Model

- First key insight: $R_0 = \beta / \gamma$

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A Basic SIR Model

- Second key insight: outbreak is not over when $R_e \leq 1$

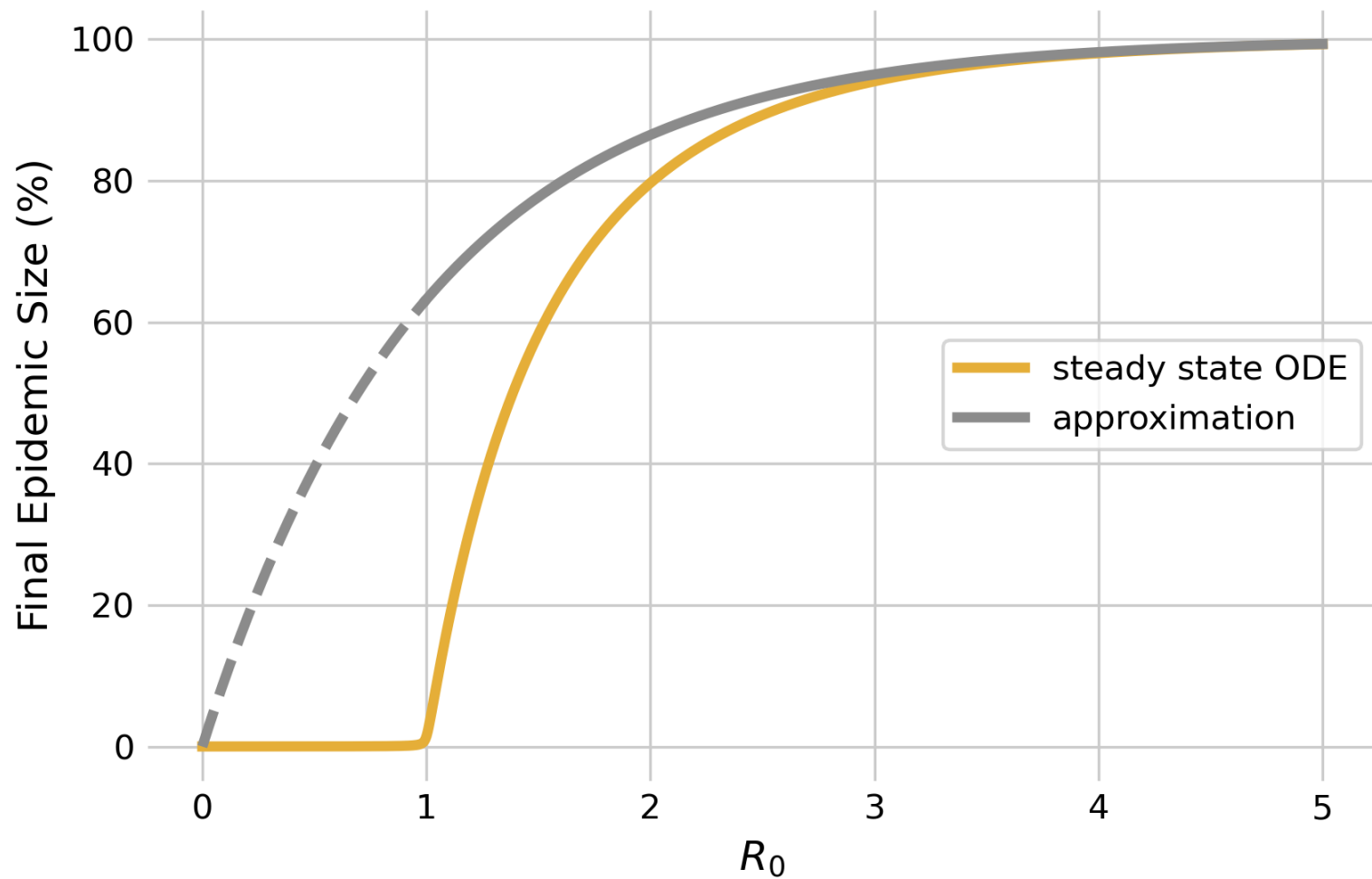


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A Basic SIR Model

- Third key insight: not everybody gets infected

$$S(\infty) = 1 - e^{-R_0}$$



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Mitigation

- “Flatten the curve”

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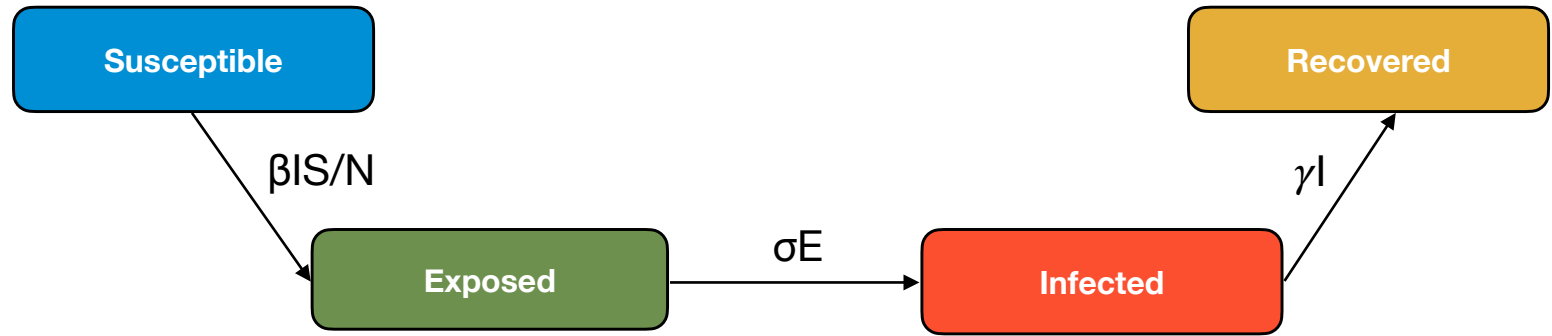
Mitigation

- Vaccination and herd immunity

$$c > 1 - \left(\frac{1}{R_0}\right)$$

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SEIR Model



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SEIR Model

$$\frac{dS}{dt} = -\beta I \frac{S}{N}$$

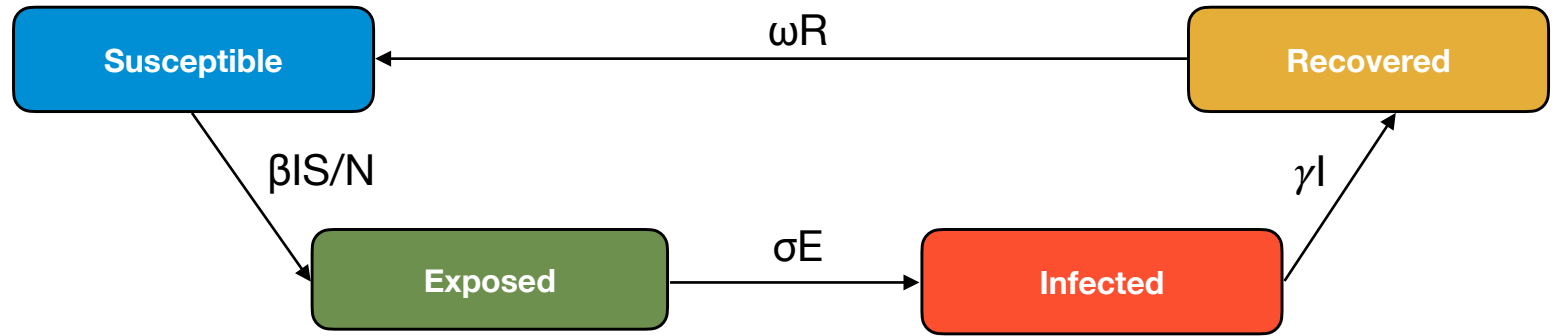
$$\frac{dE}{dt} = \beta I \frac{S}{N} - \sigma E$$

$$\frac{dI}{dt} = \sigma E - \gamma I$$

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

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SEIRS Model



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SEIRS Model

$$\frac{dS}{dt} = \omega R - \beta I \frac{S}{N}$$

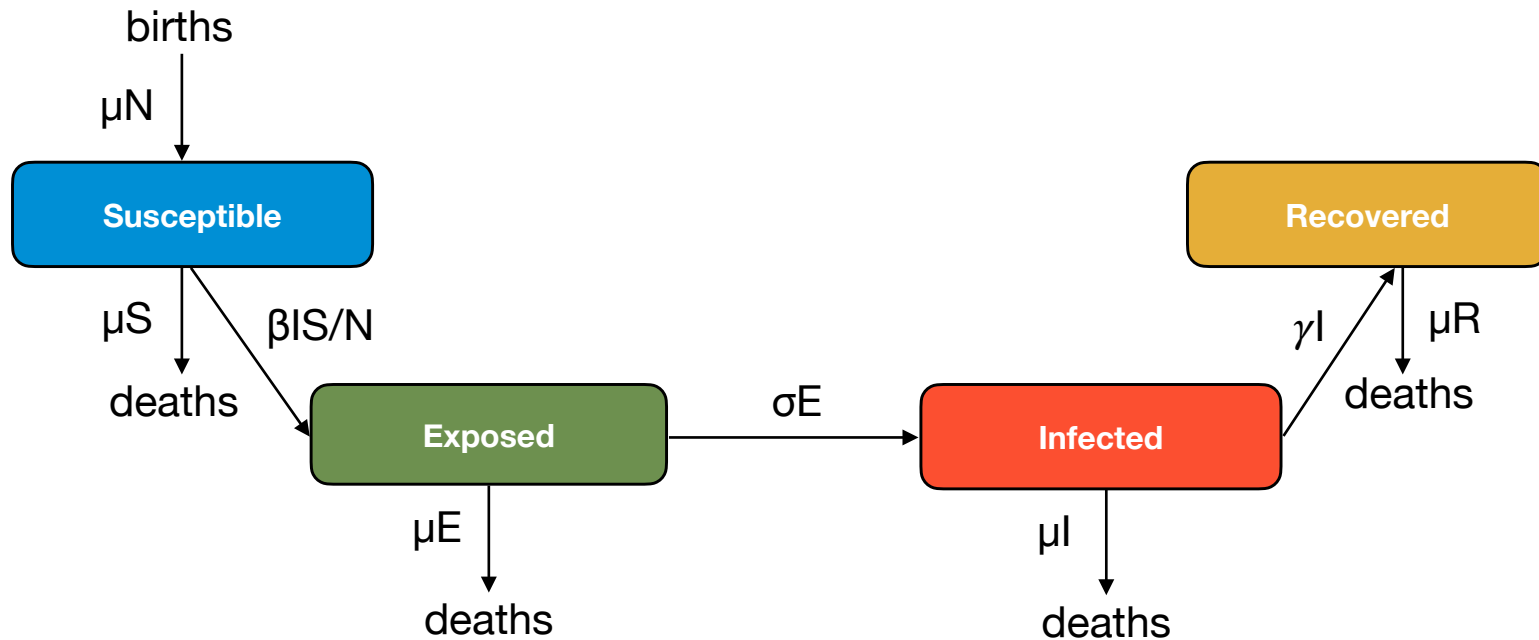
$$\frac{dE}{dt} = \beta I \frac{S}{N} - \sigma E$$

$$\frac{dI}{dt} = \sigma E - \gamma I$$

$$\frac{dR}{dt} = \gamma I - \omega R$$

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An Open Epidemic



Modeling Infectious Diseases

An Open Epidemic

$$\frac{dS}{dt} = \mu N - \beta I \frac{S}{N} - \mu S$$

$$\frac{dE}{dt} = \beta I \frac{S}{N} - \sigma E - \mu E$$

$$\frac{dI}{dt} = \sigma E - \gamma I - \mu I$$

$$\frac{dR}{dt} = \gamma I - \mu R$$

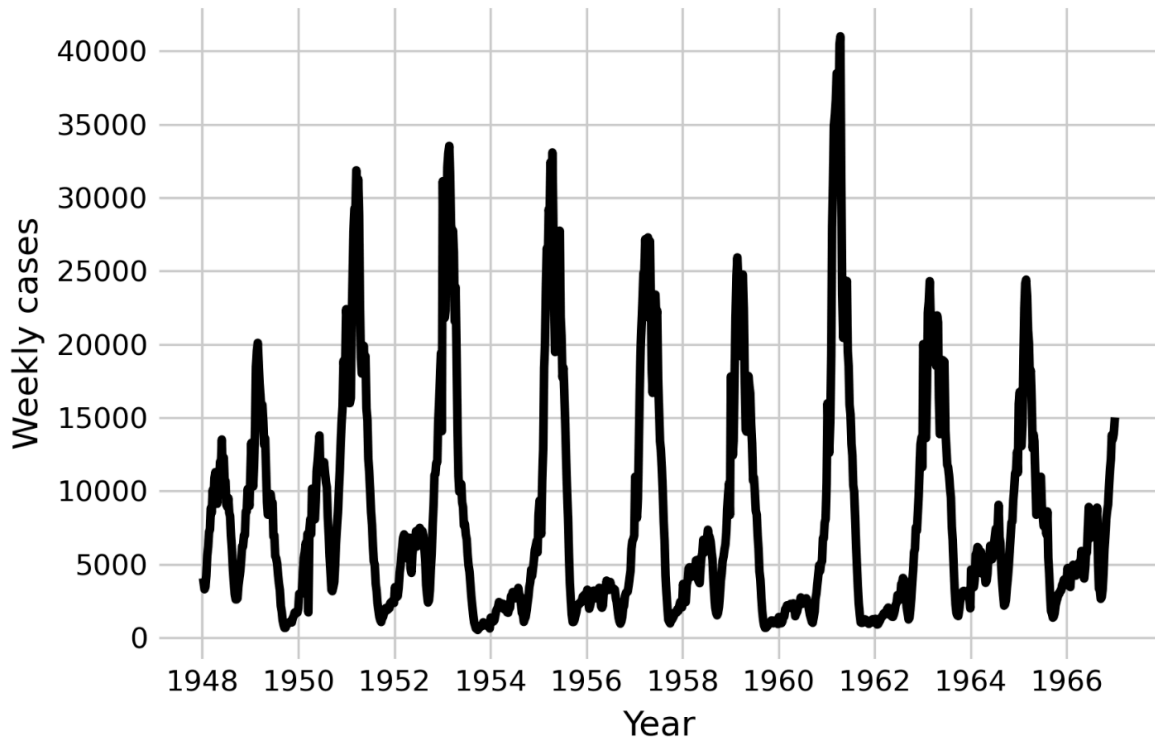
EPFL Modeling Infectious Diseases

An Open Epidemic With Seasonality

$$\beta \rightarrow \beta(1 + \beta_S \cos(2\pi t))$$

Modeling Infectious Diseases

An Open Epidemic With Seasonality



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Stochastic Models

- So far, all models were deterministic - run them again and they will produce exactly the same, predetermined result.
- Stochastic models integrate randomness - some decision are made by “rolling a dice” -> no two simulations will be identical.

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Gillespie Algorithm

- Stochastic simulation method:
 1. generates random times for each event
 2. updates states based on state transition probabilities

Modeling Infectious Diseases

Gillespie Algorithm

$$\frac{dS}{dt} = \mu N - \beta I \frac{S}{N} - \mu S$$

$$\frac{dE}{dt} = \beta I \frac{S}{N} - \sigma E - \mu E$$

$$\frac{dI}{dt} = \sigma E - \gamma I - \mu I$$

$$\frac{dR}{dt} = \gamma I - \mu R$$

Transition	Rate
$S \rightarrow S + 1$	μN
$S \rightarrow S - 1$	$\beta I \frac{S}{N} + \mu S$
$E \rightarrow E + 1$	$\beta I \frac{S}{N}$
$E \rightarrow E - 1$	$\sigma E + \mu E$
$I \rightarrow I + 1$	σE
$I \rightarrow I - 1$	$\gamma I + \mu I$
$R \rightarrow R + 1$	γI
$R \rightarrow R - 1$	μR

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Gillespie Algorithm

- Standard version:
 1. Calculate time to next event
 2. Choose the next transition, according to probabilities
- “tau leaping”:
 1. Fix regular time step τ
 2. Calculate all transitions that happen in that timestep

