## 1 Introduction

#### Goals

- Introduce the idea of individual-based models
- Discuss the need for stochasticity
- Work through some pedagogical examples
- Talk about how to investigate stochastic models

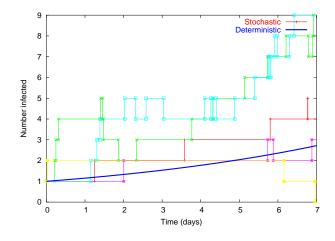
### Modelling individual events

- Differential equations model continuous processes
- Disease spreads in the real world through discrete events
- Discrete events are fundamentally stochastic
  - Even in theory we don't know when the next event will occur, nor even what it will be

# Types of stochasticity

- Demographic stochasticity is caused by the existence of individual people and discrete events
- Environmental stochasticity refers to events that affect more than one person at a time
  - Weather
  - Politics
  - Economics

## Demographic spread

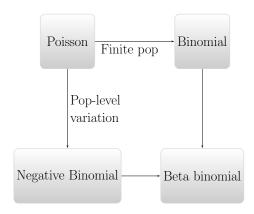


# 2 A discrete-time example

#### Model world

- Fixed  $\mathcal{R}_0$
- Model spread by disease generation
  - Generations don't overlap or we're being sophisticated
- No recovery, birth or death
  - Answer: Maybe epidemic is fast enough that these don't matter

## Practical simulation Distribution diagram



## Probability distributions

- Our practical example was equivalent to using which probability distribution?
  - **Answer:** Binomial
- What are other distributions we could have used, and what would that mean?
  - Poisson preserves the mean, but neglects the population size
    - \* Can be good when population size is unknown or large
    - \* Answer: compared to the mean
  - Poisson → negative binomial binomial → beta binomial allow for additional sources of variance

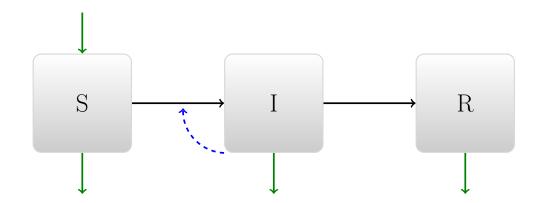
\* Answer: Analogous to environmental stochasticity

## 3 A continuous-time framework

### States and rates

- We describe our system in terms of the *probability rates* of events happening
  - If the rate of event E is  $r_E(t)$ , the probability of the event occurring in the time interval (t, t + dt) is  $r_E(t)dt$
- If we assume that event rates depend on measurable states then waiting times are exponential
  - We'll talk about relaxing this assumption

## States and rates (Demographic)



Event	transition	rate	Effect $(S, I)$
Infection	$S \to I$	$\beta SI/N$	(-1, 1)
Recovery	$I \to R$	$\gamma I$	(0, -1)
Rebirth	$R \to S$	$\mu(N-S-I)$	(1,0)
Rebirth	$I \to S$	$\mu I$	(1, -1)

## Analogy

- The demographic model is an exact analogue of the deterministic one
  - Conceptually
  - In the limit as  $N \to \infty$

# 4 A continuous time example

#### Model world

- Simple SIR
- No births, deaths or loss of immunity
- Homogeneous mixing

#### Practical simulation

Spreadsheet with event-based simulation

# 5 Analyzing stochastic systems

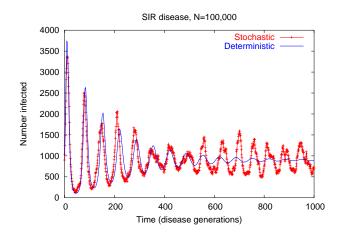
#### Realizations and ensembles

- How do we think about the behavior of a stochastic process?
  - A single example of how the process could go (e.g., from a stochastic simulation) is called a *realization*
  - The universe of possible realizations is called the *ensemble*.
  - The probability distribution that describes what state we expect the population to be in at time t is called the *ensemble distribution*

# Some techniques

- Simulate one or many realizations
- Simulate the ensemble distribution
  - Requires one state variable for each possible state of the system
- Solve the ensemble distribution dynamics exactly!
  - Rarely possible
- Approximations to the ensemble distribution

## Demographic model



## Questions

- What kind of questions do we want to ask with a stochastic model?
  - How does stochasticity affect disease dynamics?
    - \* Spatial distribution
    - \* Establishment
    - \* Persistence
  - How much variance do we expect stochasticity to cause?
  - Under what circumstances can we eliminate or eradicate a disease?

### The fate of infectious disease

- Fizzle
  - Disease fails to "establish"
- Burn-out
  - Disease goes extinct after first epidemic
- Fade-out
  - Disease goes extinct after system approaches quasi-equilibrium

#### Mathematics

- Simple models can approximate:
  - The probability of fizzle, and burn out
  - The average time until fade-out if the system survives fizzle and burn-out

- What about permanence
  - Mathematically impossible! Why?
  - Answer: In practice, corresponds to a very large average time until fade-out

## 6 Conclusions

- Stochasticity is real
  - People are individuals
  - The world has weather, and history
- Even minimal stochasticity can have dramatic effects on models of disease transmission
  - Amplified by acquired immunity
- Stochastic models are hard, and we usually combine techniques to understand them
- Even though it's real, you may not need it for your research question