## Stochastic Models - Goals

- Conceptual sense of stochastic dynamics

Play very mundanely with implementations

- Consider regimes of applicability / convenience


## Stochastic Models - Outline

- Broad Considerations
- What is a probabilistic process?
- What is a rate?
- Basic Mathematics of rates and flows
- Different implementations of stochasticity


## Real World and Model World

- Real World
- Lots of unknowns
- high complexity
- Model World
- Everything is known
- We control complexity
- Model
- Mathematical representation of model world Rules
- Scenario
- Mathematical representation of model world Events


## Why Stochasticity?

- The Real World is stochastic
- 'risks' rather than clear 'causes'
- Data would be stochastic even if the world were not
- Model world scenarios may not cluster close to average
- Population / critical sub-populations not large enough?
- 'rare' events?
- Can be simpler than complicated flow models
- Who contacts whom?
- Complex variability of biology, environment


## Choosing a Model

- Does it take in what we know and put out what we're curious about?
- Does it process the inputs according to our understanding
- Which aspects of 'real world' have been left out / included?
- How do we evaluate the 'severity' of the 'imperfections'?


## Implementing a Model

- Faithfulness of implementation
- Runtime
- Ease of (initial) deployment
- Ease of variation / maintenance


## The Most Important Diagram



## The Basic Dynamical Question

What (tiny) changes occur in the time step ( $d t$ )?


## Basic Approaches

- Formal Solution
- Discretisation - naive flow in time steps
- Simulating events.

Can vary the approach:

- for each individual, vs for population as a whole
- As happening (or not) in time steps, vs generating times to next event.


## The Most Simple Model(?)



## The Most Simple Model(?)



## The Most Simple Model(?)



$$
\frac{d S}{d t}=-r S
$$

- How many people are susceptible at time $=t$ ?
- What is $\mathrm{S}(\mathrm{t})=$ ?

There are a number of approaches...

## The Big Magic Trick

- While the (tiny) changes happen in ( $d t$ ),
- The rates of change (the rules) stay the same.
- We just express the rules (in algebra).
- This leads to (differential) equations.
- We solve these using mature tools (calculus, $R$, etc).
- The solutions are functions of time, $\mathrm{N}(\mathrm{t})$.



## 'Exact' Solution - Calculus


$d S$
$\frac{d S}{d t}=-r S(t) \quad$ Differential equation

This one is not that difficult to solve

## 'Exact' Solution - Calculus


$d S$
Solving for $\mathrm{S}(\mathrm{t})$

$$
S(t)=S_{0} e^{-r t}
$$ yields:

Analytic / Closed Form

## Finite Time Steps



$$
S(t+\Delta t)=S(t)-r S(t) \Delta t
$$

Discrete / Numerical

## 'Simulation'

In each time step, each individual "leaves" $S$ with probability $=r$


A population of 10 susceptible Individuals at $T_{\text {。 }}$

$$
T_{0}
$$

## 'Simulation'

In each time step $\Delta$, each individual "leaves" S with probability $=\delta$


$$
T_{0}
$$

- Suppose $\delta(=r \Delta)=0.05$

U ~ Uniform on the interval $[0,1]$

## Random Events in a Time Step

In each time step $\Delta$, each individual "leaves" $S$ with probability $=\delta$

$T_{0}$

| Individual $_{\mathrm{i}}$ | $\mathrm{U}_{\mathrm{i}}$ |
| :---: | :--- |
| 1 | 0.871 |
| 2 | 0.600 |
| 3 | 0.290 |
| 4 | 0.335 |
| 5 | 0.421 |
| 6 | 0.180 |
| 7 | 0.033 |
| 8 | 0.663 |
| 9 | 0.338 |
| 10 | 0.246 |

## Completing the Time Step

In each time step $\Delta$, each individual "leaves" $S$ with probability $=\delta$

$T_{0}$

$$
\begin{array}{cccc}
I_{1} & I_{2} & I_{3} & \\
I_{10} & \text { S } & I_{5} & l_{4} \\
I_{9} & I_{8} & & \\
I_{6}
\end{array}
$$

$T_{1}$

## Abstracting Individuals

Using indicator variables to denote if an individual is still in " $S$ "

$T_{0}$

$T_{1}$

## The Most Simple Model(?)

Approach \#3
Repeat for the desired number of time steps to determine $\mathrm{S}(\mathrm{t})$

$T_{1} \quad T_{t}$


Stochastic/Simulation

## Discretisation 'Error'

- In each time step $\Delta$, each individual "leaves" S with probability $=\delta$
- Can we really have $\delta=r \Delta$ when $\Delta$ is not small?
- 'Discretisation error'
- How about $\delta=1-\exp (-r \Delta)$
- Taylor expansion: $f(x)=f(o)+f^{\prime}(0) x^{1}+f^{\prime \prime}(0) x^{2} / 2!+\ldots$
- So then $\exp (x)=1-x+\ldots$
- And so $\mathrm{r} \Delta \approx 1-\exp (-r \Delta)$ ?


## Events versus Time Steps

- In this simple model, we can discretise 'correctly'
- This still only evaluates the system at discrete time points
- Why not calculate exact times at which population members leave?

$$
S(t)=S_{0} e^{-r t} \rightarrow \quad P(t)=e^{-r t}
$$

## Scheduling Events

Transforming a uniform random number into a 'waiting time'


## Scheduling Scheduling

- Should we schedule each exit up front?
- Or should we schedule the next exit
- then the next, ...

