

# Discrete-time deterministic models

Modern Techniques in Modelling

LONDON  
SCHOOL of  
HYGIENE  
& TROPICAL  
MEDICINE



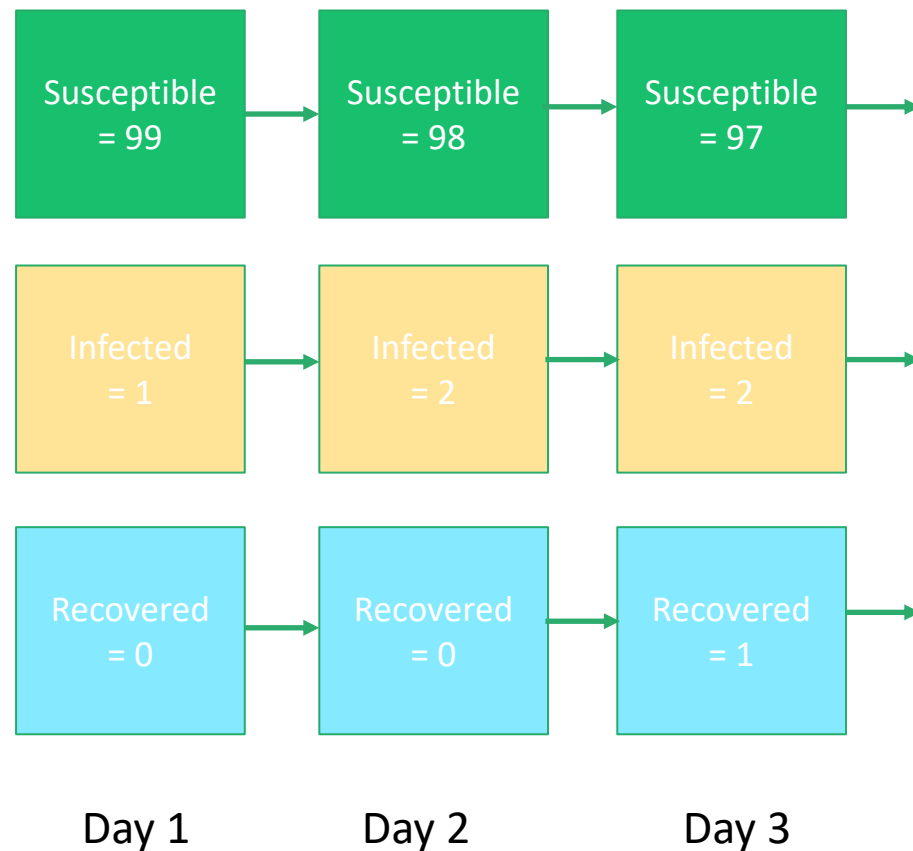
# Introduction

LONDON  
SCHOOL of  
HYGIENE  
& TROPICAL  
MEDICINE



# Discrete-time deterministic models

Often we may want to model an epidemic in terms of a discrete time step, e.g. from one day or week to the next:



This can be modelled by considering the current state of the epidemic,  $y$ , at time  $t$ :

$$\mathbf{y}(t) = S(t) + I(t) + R(t)$$

that updates with each time step by some function of the current state:

$$\mathbf{y}(t + 1) = \mathbf{y}(t) + g(\mathbf{y}(t))$$

This is known generally as a **difference equation**, the solution of which can be approximated simply for discrete time steps using **Euler's method**.

Consider an SIR model for a closed population that follows the following rules:

- Susceptible individuals become **infected** at a rate proportional to the size of the product of *susceptible* and *infectious* populations:

$$S(t + 1) = S(t) - \beta * S(t) * I(t)$$

where  $\beta$  is the per capita infection rate.

- **Recovery** from infection grants life-long immunity

$$R(t + 1) = R(t) + \gamma * I(t)$$

Where  $\gamma$  is the recovery rate, and  $1/\gamma$  is the mean time spent infectious (1/rate = duration).

The only transitions between states are **infection** and **recovery**

$$I(t + 1) = I(t) + \beta * S(t) * I(t) - \gamma * I(t)$$

Which adds those being **infected** and subtracts those who **recover**

We start with:

- 1% of the population are infected,  $I(0) = 0.01$
- no recovered,  $R(0) = 0$ , and
- the remainder of the population is susceptible,  $S(0) = 0.99$ ,
- for  $N(t) = 1$ , the total population.

Our system is therefore:

$$\begin{array}{lll} S(t+1) & = S(t) - \beta * S(t) * I(t), & S(0) = 0.99 \\ I(t+1) & = I(t) + \beta * S(t) * I(t) - \gamma * I(t), & I(0) = 0.01 \\ R(t+1) & = R(t) + \gamma * I(t), & R(0) = 0 \end{array}$$



This system can also be written in vector form as:

$$\begin{bmatrix} S \\ I \\ R \end{bmatrix}_{t+1} = \begin{bmatrix} S \\ I \\ R \end{bmatrix}_t + \begin{bmatrix} -\beta SI \\ \beta SI - \gamma I \\ \gamma I \end{bmatrix}_t$$

Here the subscript indicates the time value and the term on the right is our *update vector*

In R, we can write a function to return our update vector with three elements:

```
update_sir <- function(t, y, parms){  
  S <- y[1]  
  I <- y[2]  
  R <- y[3]  
  
  beta <- parms['beta']  
  gamma <- parms['gamma']  
  
  out <- c(- beta*S*I,  
           + beta*S*I - gamma*I,  
           + gamma*I)  
  
  return(out)  
}
```

# Discrete-time deterministic SIR

```
parms_sir <- c(beta = 1.3,  
               gamma = 0.23)  
  
time_sir <- seq(0, 20, by = 1)  
  
y_sir      <- matrix(data = NA,  
                     nrow = length(time_sir),  
                     ncol = 3)  
  
# initial values at t=0  
y_sir[1, ] <- c(0.99, 0.01, 0)  
  
for (i in 1:(nrow(y_sir) - 1)){  
  
  y_sir[i + 1,] <- y_sir[i,] +  
    update_sir(t      = time_sir[i + 1],  
              y       = y_sir[i, ],  
              parms   = parms_sir)  
}
```

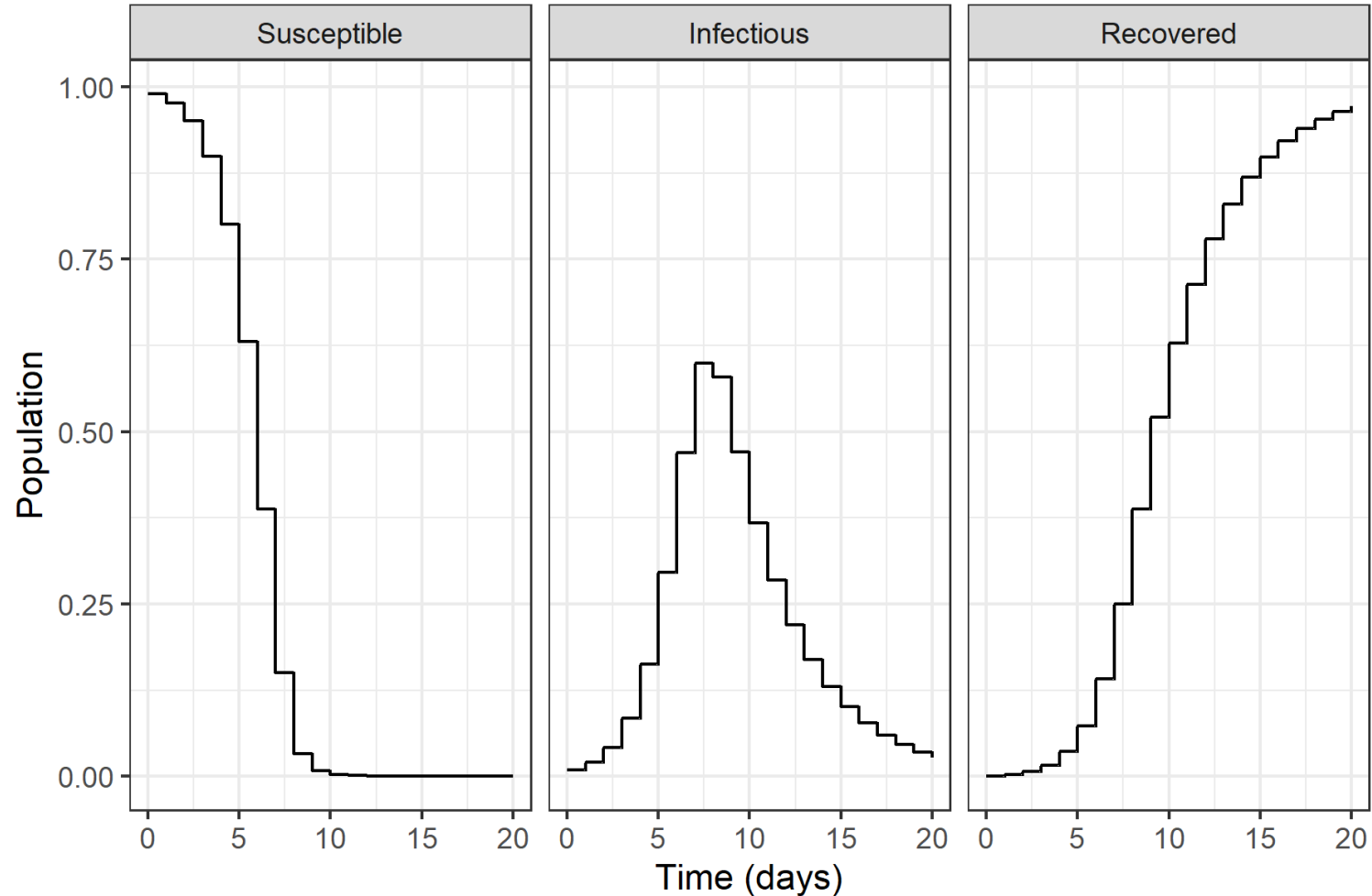
# Discrete-time deterministic SIR

```
y_sir_df <- as.data.frame(y_sir)
names(y_sir_df) <- c('Susceptible',
                     'Infectious',
                     'Recovered')
y_sir_df <- cbind(time = time_sir,
                  y_sir_df)
```

```
head(y_sir_df)
```

##	time	Susceptible	Infectious	Recovered
## 1	0	0.9900000	0.0100000	0.0000000
## 2	1	0.9771300	0.0205700	0.0023000
## 3	2	0.9510006	0.0419683	0.0070311
## 4	3	0.8991151	0.0842011	0.0166838
## 5	4	0.8006967	0.1632533	0.0360501
## 6	5	0.6307654	0.2956363	0.0735983

# Discrete-time deterministic SIR



Our more general form of the update is

$$\mathbf{y}(t + 1) = f(t, \mathbf{y}(t), \boldsymbol{\theta}(t))$$

...but parameters may or may not change with time  $t$ .

# Summary



- Difference equations are defined for discrete time steps, and can be solved approximately by iterating over that time step (Euler's method).
- Discrete time models can be specified either as:

$\mathbf{y}(t + \Delta t) = f(\mathbf{y}(t), t, \boldsymbol{\theta}, \Delta t)$  to transform current state

$\mathbf{y}(t + \Delta t) = \mathbf{y}(t) + \Delta t g(\mathbf{y}(t), t, \boldsymbol{\theta})$  to update current state

- $f(\cdot), g(\cdot)$  can be any function that captures the dynamics of the physical system we're interested in.



# Looking forward

- Later today you'll start to learn about extending to continuous time models with **differential equations**.
- Later in the week you'll learn about update functions that model infection and recovery as probabilistic events.



# Practical