#### Modelling problem



We are interested in modelling the 1918 influenza pandemic, specifically what its impact may have been on Greater London at the time. It is estimated that the average person infected will have been infectious from 2 days after exposure, until 6 days after exposure. The basic reproduction number,  $R_0$ , was estimated at around 2.4. Individuals who recover have long-lasting immunity.

What is an appropriate model structure?

**SEIR** 

Draw the model structure?

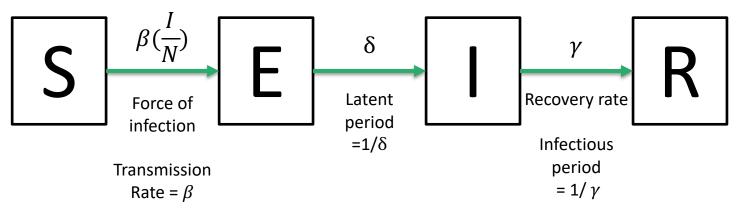
What are the key parameters (by name/symbol)?

What are the key parameters (by value)?

**Initial conditions?** 

$$\delta$$
 = 0.5 days<sup>-1</sup>  
 $\gamma$  = 0.25 days<sup>-1</sup>  
 $R_0 = \beta(1/\gamma)$   
 $\beta = R_0 \gamma = 2.4 \times 0.25 \text{ days}^{-1}$   
 $\beta = 0.6 \text{ days}^{-1}$   
 $N = 7.3 \text{ million}$ 

0



# Metapopulations with ODEs

Modern Techniques in Modelling







### Previously: ODE models in R



- We defined a set of ODEs
- Then used deSolve to simulate the model output

$$\frac{\mathrm{d}S}{\mathrm{d}t} = -\beta SI/N$$

$$\frac{\mathrm{d}I}{\mathrm{d}t} = \beta SI/N - \gamma I$$

$$\frac{\mathrm{d}R}{\mathrm{d}t} = \gamma I$$

#### Example code



```
SIR model <- function(times, state, parms){</pre>
  ## Define variables
  S <- state["S"]
  I <- state["I"]</pre>
  R <- state["R"]</pre>
  N \leftarrow S + I + R
  # Extract parameters
  beta <- parms["beta"]</pre>
  gamma <- parms["gamma"]</pre>
  # Define differential equations
  dS \leftarrow - (beta * S * I) / N
  dI \leftarrow (beta * S * I) / N - gamma * I
  dR <- gamma * I
  res <- list(c(dS, dI, dR))
  return(res)
```

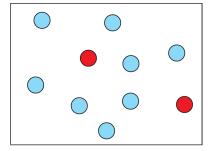
## Metapopulations

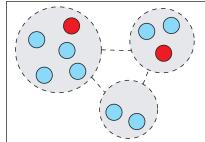


#### Non-random mixing



- So far we've assumed everyone mixes together randomly
- In reality, people may group together in different locations/settings/group s
- If we're interested in heterogeneity, we need to model multiple populations ('metapopulations')





Random mixing vs metapopulations

#### Metapopulation model



- Let's consider two linked populations.
- This means expanding our SIR model:

$$\frac{dS_1}{dt} = -\beta S_1 I_1 / N_1$$

$$\frac{dI_1}{dt} = \beta S_1 I_1 / N_1 - \gamma I_1$$

$$\frac{dR_1}{dt} = \gamma I_1$$

$$\frac{dS_2}{dt} = -\beta S_2 I_2 / N_2$$

$$\frac{dI_2}{dt} = \beta S_2 I_2 / N_2 - \gamma I_2$$

$$\frac{dR_2}{dt} = \gamma I_2$$

#### Connecting the populations



- Susceptibles in population 1 can have contact with infectives in population 1 and population 2.
- But may contact population 2 at a different (lower?) rate:

$$\frac{dS_1}{dt} = -S_1 \begin{bmatrix} \widehat{\beta I_1/N_1} + \widehat{\alpha \beta I_2/N_2} \end{bmatrix}$$

#### Connecting the populations



— And same logic for infectious compartment and population 2:

$$\frac{dS_1}{dt} = -S_1[\beta I_1/N_1 + \alpha \beta I_2/N_2] 
\frac{dI_1}{dt} = S_1[\beta I_1/N_1 + \alpha \beta I_2/N_2] - \gamma I_1 
\frac{dS_2}{dt} = -S_2[\beta I_2/N_2 + \alpha \beta I_1/N_1] 
\frac{dI_2}{dt} = S_2[\beta I_2/N_2 + \alpha \beta I_1/N_1] - \gamma I_2$$

#### **Practical**



- Open the practical 05\_ODEMetapopulation/01\_ODE\_Metapop.R and add your code to the existing R script
- Objective : implement SIR metapopulation model with two populations
- Answer questions A-D
- Question E is optional
- Book by Keeling & Rohani (Princeton, 2007) has more details about metapopulations.