

# Day 1

## Lecture 2:

# Basic concepts of compartmental models



**Short course on modelling infectious disease dynamics in R**

Ankara, Türkiye, September 2025

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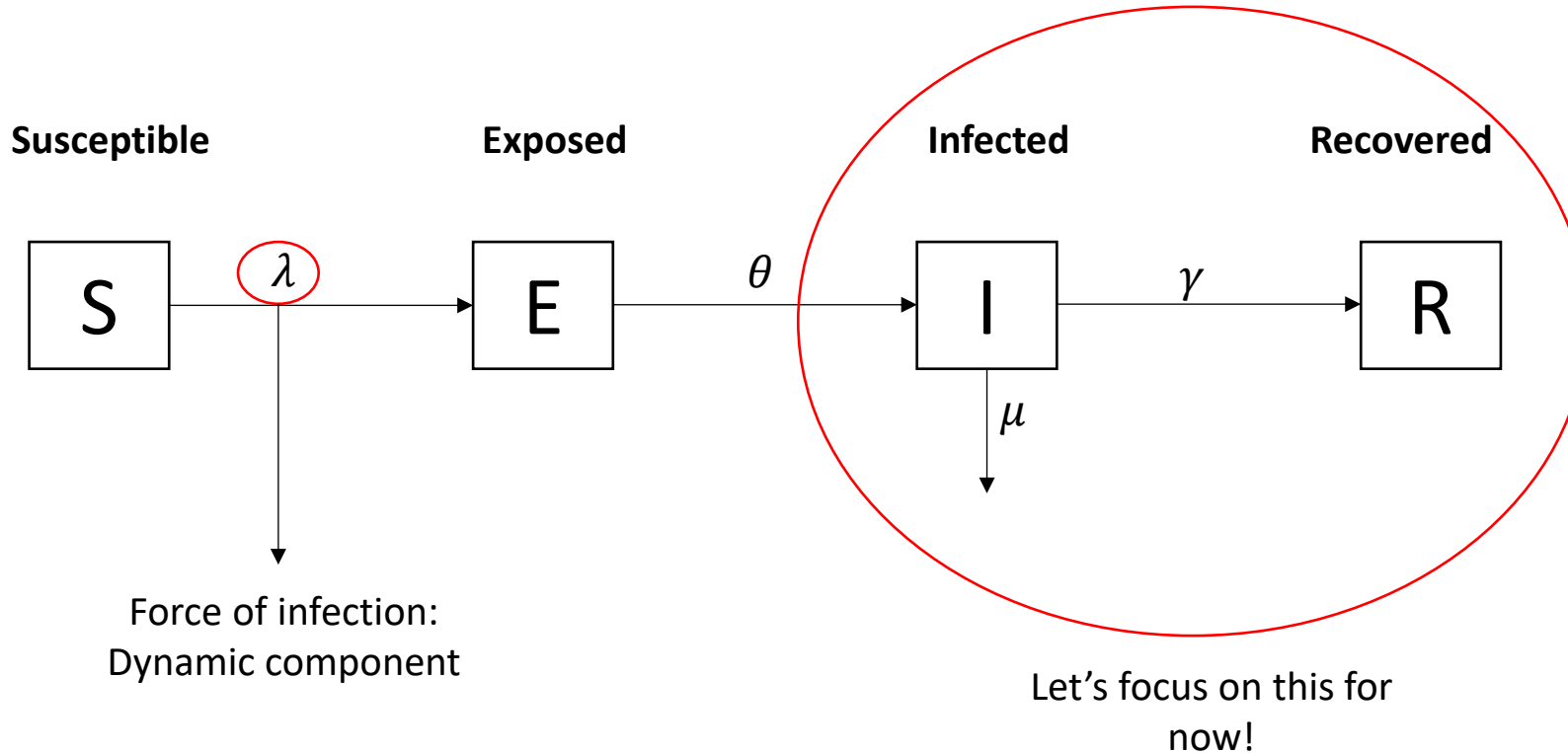
# Aims of the session

- To Understand how disease models are designed
- Review concepts of probabilities, rates and competing hazards
- Understand the assumptions behind compartmental models

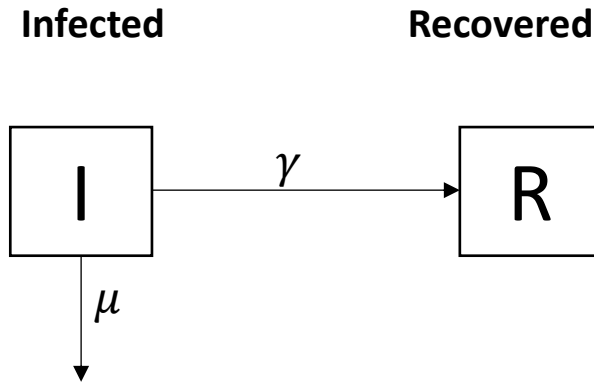
# Checklist for designing a disease model

- 1) Describe the natural course of disease
- 2) Identify the necessary transitions between compartments
- 3) Interpret these transitions and find the relevant data to estimate the parameters
- 4) Review your research question and adapt your model complexity accordingly
- 5) Code your model!

# A simple example

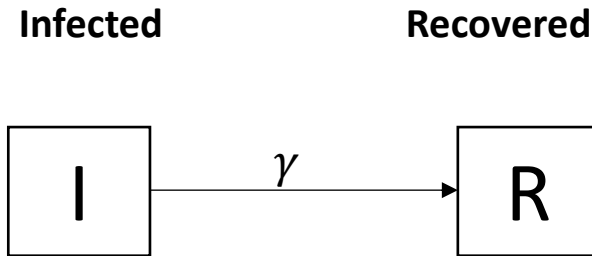


# A cohort model



- How do we specify rates of transitions?
- How multiple rates affect a compartment?

# A cohort model

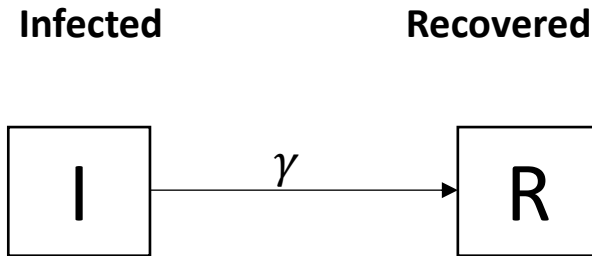


$$\frac{dI}{dt} = -\gamma I(t)$$

$$\frac{dR}{dt} = \gamma I(t)$$

- The rate of flow out of compartment  $I$  is proportional to the number of people on  $I$
- $\gamma$  is the proportionality constant
- We call  $\gamma$  a **constant hazard**
- A hazard is the event rate at a specific time  $t$

# A cohort model

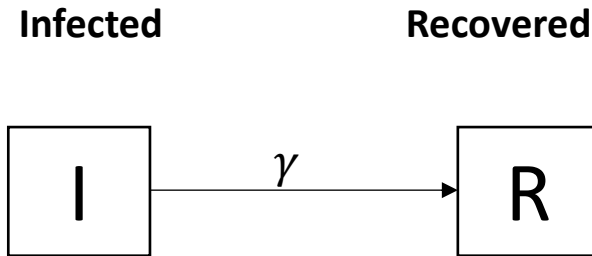


$$\frac{dI}{dt} = -\gamma I(t)$$

$$\frac{dR}{dt} = \gamma I(t)$$

- In our cohort  $\gamma$  is the recovery rate
- The larger  $\gamma$  is, the quicker they recover
- This means  $\gamma$  must be expressed in units of inverse time  $\text{day}^{-1}$
- An average recovery rate of 10 days =  $0.1 \text{ day}^{-1} = 1/10$

# So how do we interpret this transitions?



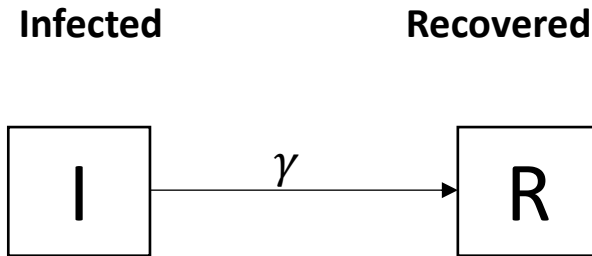
$$\frac{dI}{dt} = -\gamma I(t)$$

$$\frac{dR}{dt} = \gamma I(t)$$

- An initial cohort of 1000 people infected
- A recovery rate of  $0.1 \text{ day}^{-1}$  (10 days)
- When will 50% of infected be recovered?



# So how do we interpret this transitions?



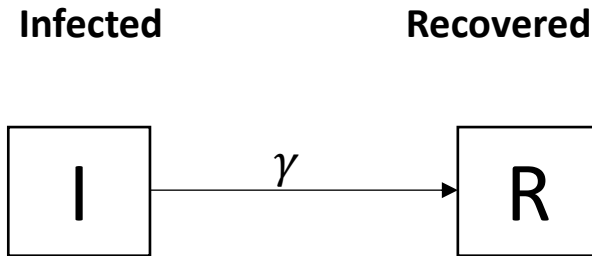
Solution

$$\frac{dI}{dt} = -\gamma I(t) \quad \Longrightarrow \quad I = I_0 e^{-\gamma t}$$

$$\frac{dR}{dt} = \gamma I(t) \quad \Longrightarrow \quad R = I_0 (1 - e^{-\gamma t})$$

- An initial cohort of 1000 people infected
- A recovery rate of  $0.1 \text{ day}^{-1}$  (10 days)
- When will 50% of infected be recovered?

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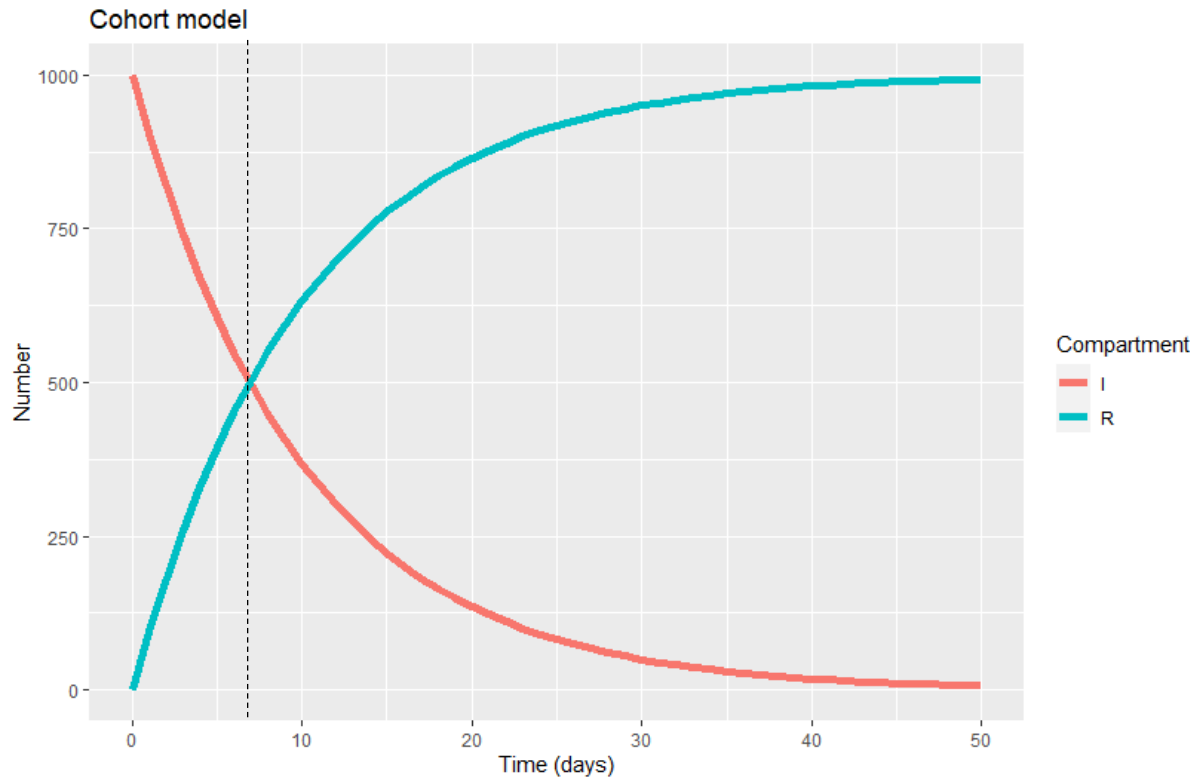
Solution

$$\frac{dI}{dt} = -\gamma I(t) \quad \Longrightarrow \quad I = 1000e^{-0.1(10)} \text{ recovered?}$$

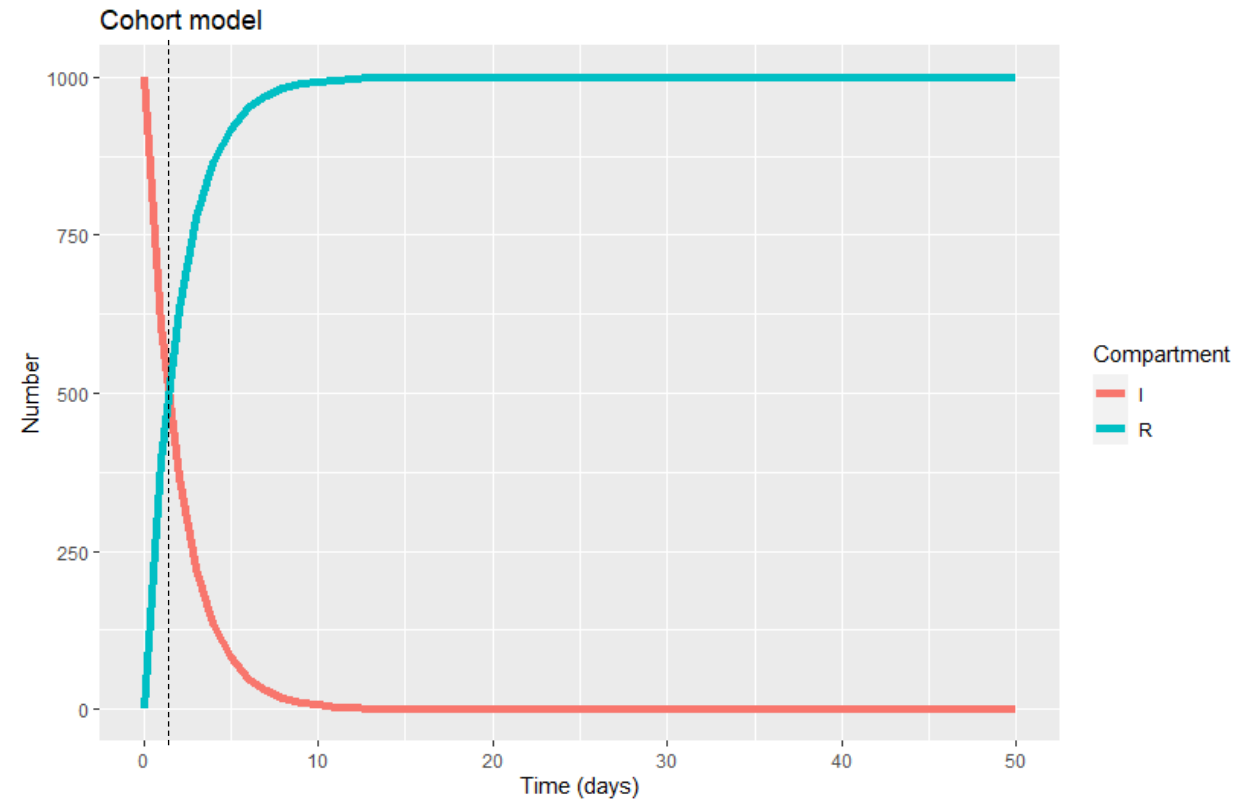
$$\frac{dR}{dt} = \gamma I(t) \quad \Longrightarrow \quad R = 1000(1 - e^{-0.1(10)})$$

# So how do we interpret this transitions?

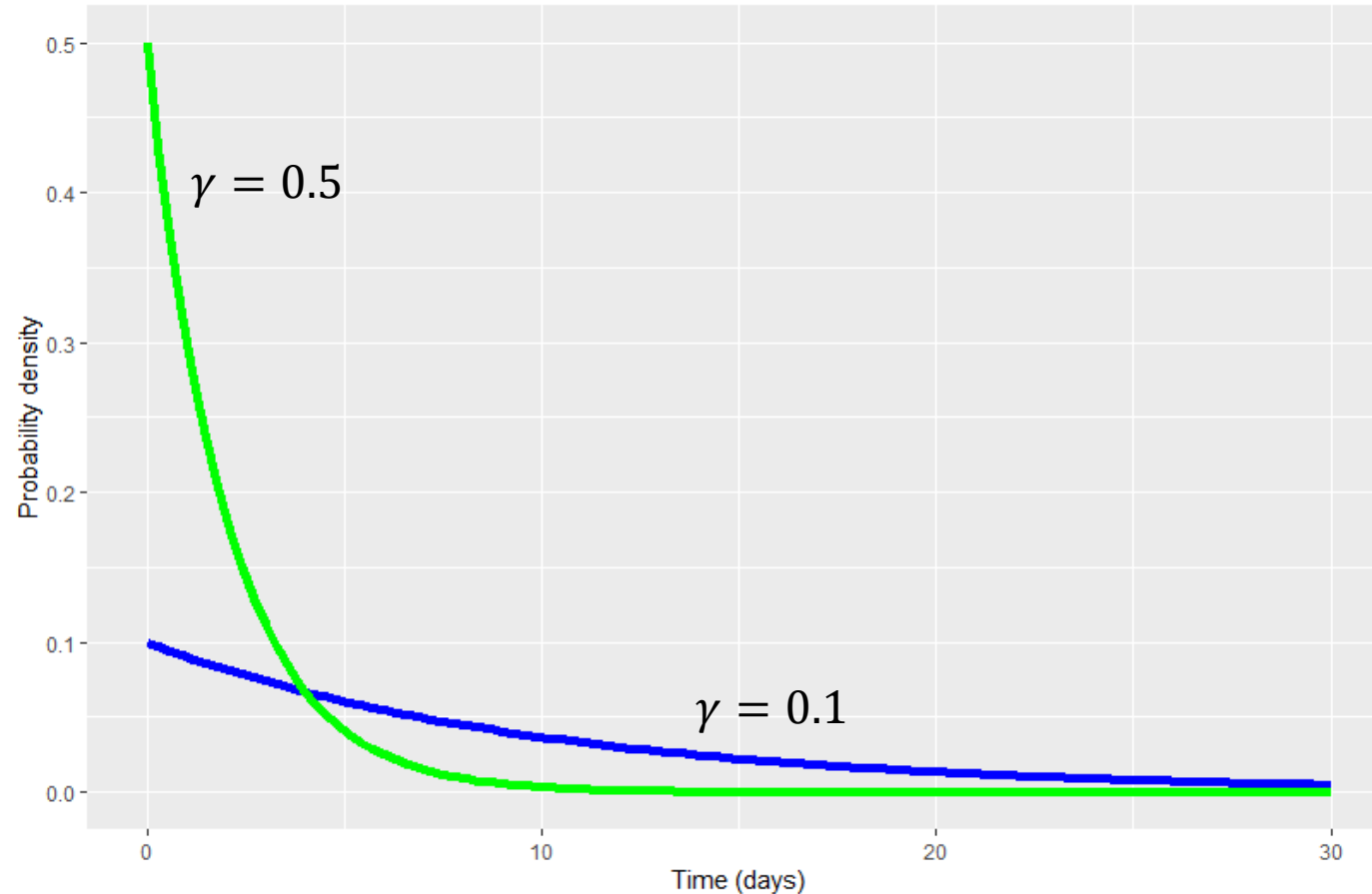
Recovery rate of 10 days



Recovery rate of 2 days. Much quicker !

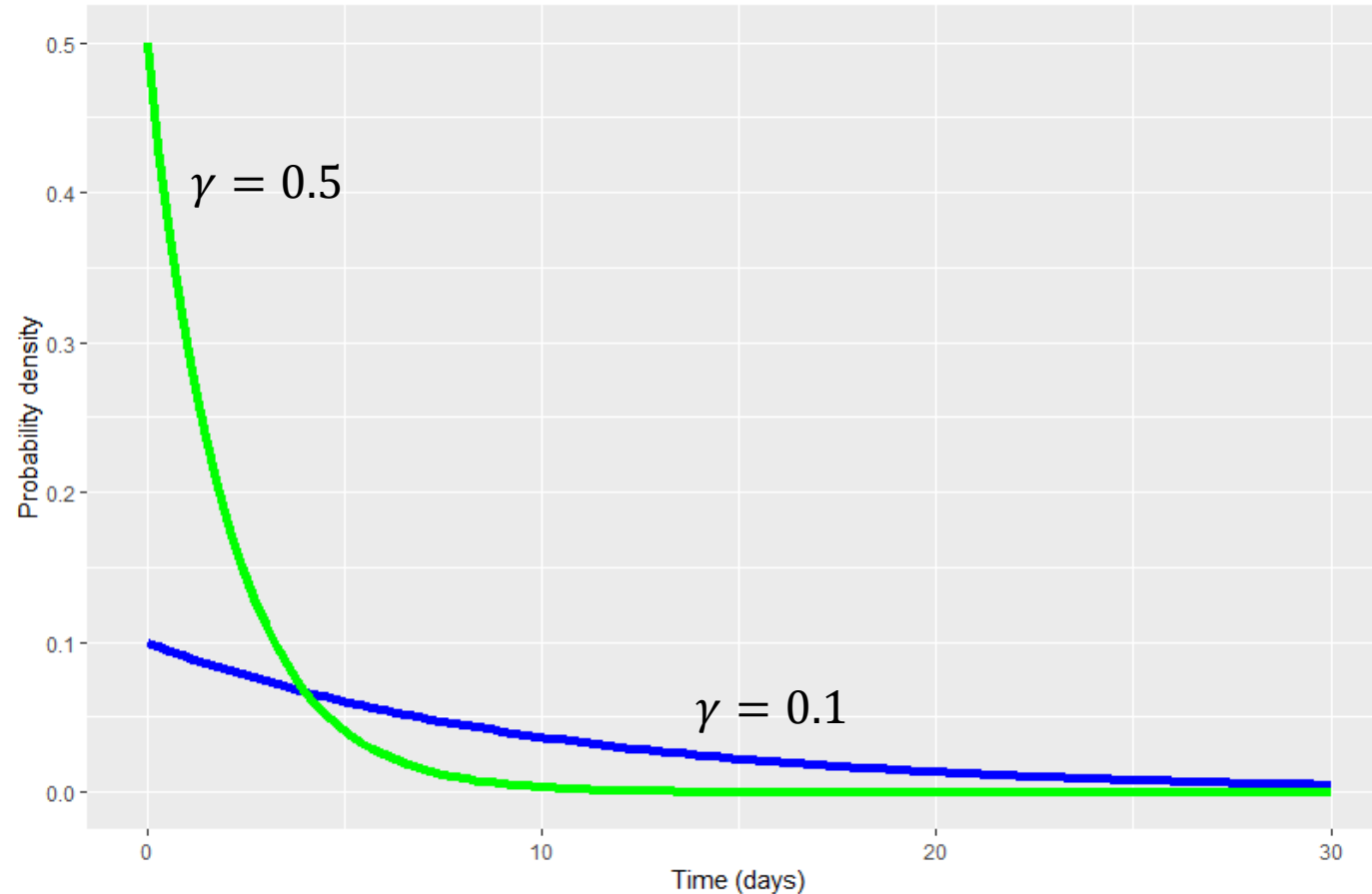


# Behaviour of the exponential distribution



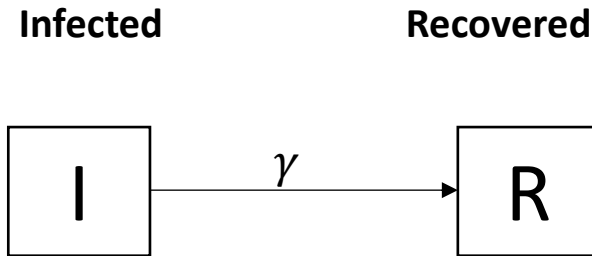
- For  $\gamma = 0.1$  , we can say the mean infectious period is 10 days
- For  $\gamma = 0.5$  , we can say the mean infectious period is 2 days

# Behaviour of the exponential distribution



- The time spent in  $I$  follows an exponential distribution with exponential parameter  $\gamma$
- The mean of that distribution is  $1/\gamma$ , which is in our case the mean infectious period (in days)
- The shorter the infectious period, the larger (quicker) the recovery rate !

# Competing Hazards

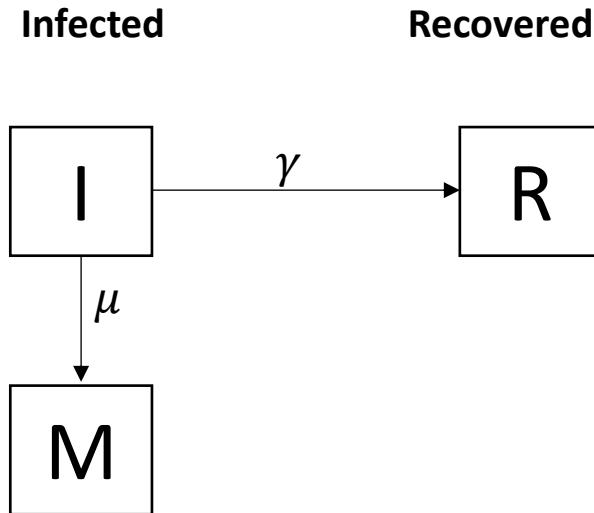


- Let's add some complexity by adding a mortality rate

$$\frac{dI}{dt} = -\gamma I(t)$$

$$\frac{dR}{dt} = \gamma I(t)$$

# Competing Hazards



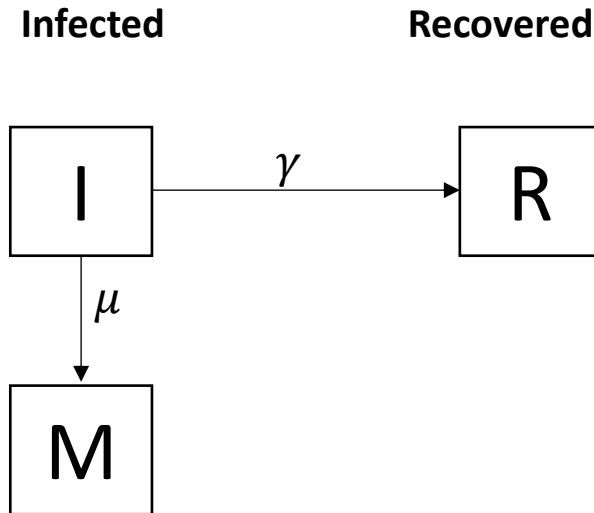
$$\frac{dI}{dt} = -(\gamma + \mu)I(t)$$

$$\frac{dR}{dt} = \gamma I(t)$$

$$\frac{dM}{dt} = \mu R(t)$$

- Let's add some complexity by adding a mortality rate  $\mu$
- More than one event can flow out of  $I$  compartment.
- This is what we call competing hazards:  $\mu$  and  $\gamma$

# Competing Hazards



$$\frac{dI}{dt} = -(\gamma + \mu)I(t)$$

$$\frac{dR}{dt} = \gamma I(t)$$

$$\frac{dM}{dt} = \mu R(t)$$

- if  $\mu > \gamma$ , means more people die before they recover (e.g, Ebola)
- This means that for a particular compartment, two hazard rates are competing
- We need to account for that when we define the value of our rates



# Competing Hazards: CFR

- The case fatality rate is the proportion of people that die before they recover.
- It can be expressed as :

$$\text{CFR} = \mu / (\gamma + \mu)$$

- Similarly, the survival rate is

$$\text{survival} = \gamma / (\gamma + \mu)$$

# Competing Hazards: estimate $\mu$ from CFR

Some algebra

$$\left\{ \begin{array}{l} CFR = \frac{\mu}{\mu + \gamma} \\ \mu = CFR(\mu + \gamma) \\ \mu = \mu CFR + \gamma CFR \\ \mu - \mu CFR = \gamma CFR \\ \mu(1 - CFR) = \gamma CFR \\ \mu = \frac{\gamma CFR}{1 - CFR} \end{array} \right. \quad \left. \vphantom{\frac{\gamma CFR}{1 - CFR}} \right\} \text{ This is what we need}$$

Disease	Pathogen	Eradication status	Deaths per year (in most recent year)	Case fatality rate (if untreated)
Smallpox	Variola virus	eradicated	0	±30%
Rinderpest	Rinderpest virus	eradicated	0	up to 100%
Polio	Poliovirus	eradication under way	0	<0.5%
Guinea worm	Guinea worm (nematode)	eradication under way	not deadly	0%
Yaws	Treponema pallidum (bacterium)	eradication under way	not deadly	0%
Rabies	Lyssavirus	global elimination under way	13,289 (2016)	100%
Tuberculosis	Mycobacterium tuberculosis (bacterium)	possibly eradicable in the future	1.21 million (2016)	70%
HIV/AIDS	Human immunodeficiency virus	possibly eradicable in the future	1.03 million (2016)	up to 100%
Malaria	Plasmodium (unicellular parasite)	possibly eradicable in the future	0.72 million (2016)	±0.3%

source: <https://ourworldindata.org/uploads/2018/06/Key-facts-about-eradicated-and-eradicable-diseases.png>

# What we should know by now

- What are compartmental models
- How we use ODEs to write these models
- What are hazard rates and how can be interpreted
- What are competing hazards and why are these important.
- What is the case fatality rate (CFR)