

# Basic Reproduction Numbers for Compartmental Models

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# Outline

1 Motivations

2 ODE systems

3 Further developments

# A population growth model

**Example 1.** Consider a single species growth model

$$u' = g(u), \quad u(0) = u_0 \geq 0. \quad (1.1)$$

Assume that  $g(0) = 0$ . Linearizing (1.1) at  $u = 0$ , we obtain

$$u' = g'(0)u = bu - au,$$

where  $b$  is the intrinsic birth rate,  $a$  is the mortality rate. Note that  $\frac{1}{a}$  is the average life span. It follows that the basic reproduction number is

$$R_0 = b \cdot 1 \cdot \frac{1}{a} = \frac{b}{a}.$$

Clearly,  $u = 0$  is linearly stable if  $R_0 < 1$ , and linearly unstable if  $R_0 > 1$ .

# An epidemic model

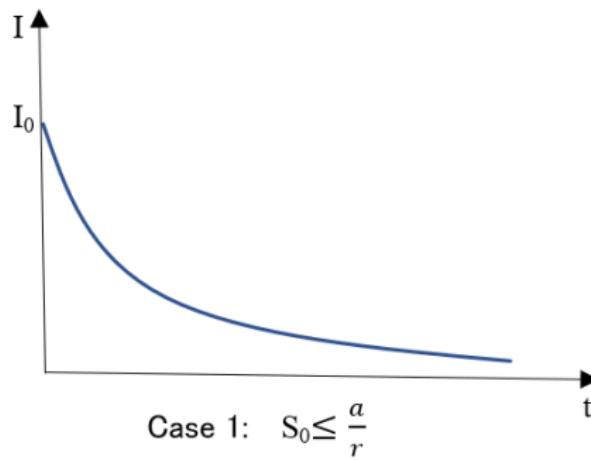
**Example 2.** For the transmission of a disease, let  $S$  be the susceptible,  $I$  be the infected, and  $R$  be the recovered individuals, respectively. Then we have the following Kermack-McKendrick SIR model (1927):

$$\begin{aligned}\frac{dS}{dt} &= -rSI \\ \frac{dI}{dt} &= rSI - aI \\ \frac{dR}{dt} &= aI\end{aligned}\tag{1.2}$$
$$S(0) = S_0 > 0, I(0) = I_0 > 0, R(0) = 0$$

where  $r$  is the contact rate, and  $a$  is the removal rate. Note that  $\frac{1}{a}$  is the average infection period. It then follows that system (1.2) admits the following threshold type dynamics.

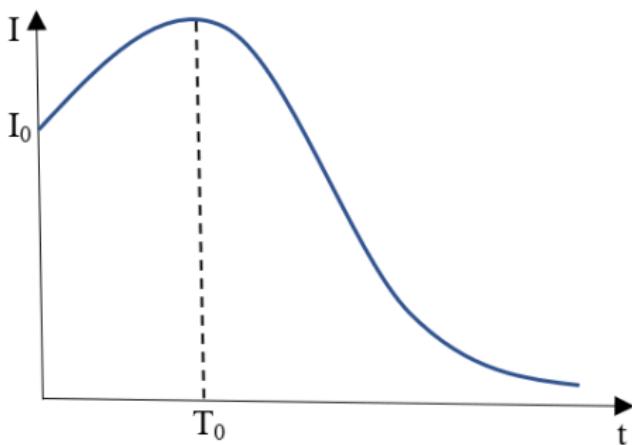
# Threshold dynamics

**Case 1.** If  $S_0 \leq \frac{a}{r}$ , then both  $S(t)$  and  $I(t)$  are decreasing in  $t \in [0, \infty)$ , and  $\lim_{t \rightarrow \infty} S(t) = S_\infty > 0$  and  $\lim_{t \rightarrow \infty} I(t) = 0$ .



This implies that the disease will decrease to extinction.

**Case 2.** If  $S_0 > \frac{a}{r}$ , then there exists  $T_0 > 0$  such that  $S(T_0) = \frac{a}{r}$ , and  $S(t)$  is decreasing in  $t \in [0, \infty)$  and  $\lim_{t \rightarrow \infty} S(t) = S_\infty > 0$ , while  $I(t)$  is increasing in  $t \in [0, T_0]$ , decreasing in  $t \in [T_0, \infty)$ , and  $\lim_{t \rightarrow \infty} I(t) = 0$ .



$$\text{Case 2: } S_0 > \frac{a}{r}$$

This implies that the disease will spread for some period of time, and then decrease to extinction.

# Basic reproduction number

**Initial reproduction number**  $R_0$  is the total number of secondary cases (infections) one average infective can produce if the number of susceptibles remained at its initial size  $S_0$ . Thus, we have

$$R_0 = rS_0 \cdot 1 \cdot \frac{1}{a} = \frac{rS_0}{a}.$$

**K-M Theorem.** *If  $R_0 \leq 1$ , then there is no epidemic outbreak; If  $R_0 > 1$ , then there is an epidemic outbreak.*

Linearizing (1.2) at the disease-free equilibrium  $(S_0, 0)$ , we obtain a linear equation for  $I$ :

$$\frac{dI}{dt} = rS_0I - aI := FI - VI = (F - V)I,$$

where  $F = rS_0$  is the infection rate, and  $V = a$  gives an internal evolution law  $\frac{dI}{dt} = -VI$  when there is no infection.

# Compartmental models

General approach: O. Diekmann, J. A. P. Heesterbeek and J. A. J. Metz, *J. Math. Biol.*, 28(1990), 365-382.

Compartmental models: P. van den Driessche and J. Watmough, *Math. Biosci.*, 180(2002), 29-48.

Linearizing a given compartmental model at the disease-free equilibrium  $(S^*, 0)$ , we obtain a liner system for infected variable  $I$ :

$$\frac{dI}{dt} = FI - VI = (F - V)I,$$

where  $F$  is the infection matrix, and the internal evolution law is  $\frac{dI}{dt} = -VI$  when there is no infection. Usually,  $F$  is nonnegative, and  $-V$  is cooperative. Further, all eigenvalues of  $-V$  have negative real parts (**why?**).

# Next generation matrix

Let  $\psi(0) = (\psi_1(0), \dots, \psi_m(0))$  be the distribution of infected individuals initially in the infected compartments, and  $\psi(t) = (\psi_1(t), \dots, \psi_m(t))$  be the distribution of these initially infected individuals remaining infected after  $t$  time units. Then

$$\frac{d\psi(t)}{dt} = -V\psi(t) \implies \psi(t) = e^{-Vt}\psi(0), \forall t \geq 0.$$

It follows that the expected distribution of new infectives is

$$\int_0^\infty F \cdot \psi(t) dt = \int_0^\infty F \cdot e^{-Vt}\psi(0) dt = FV^{-1}\psi(0).$$

Then  $\psi(0) \rightarrow FV^{-1}\psi(0)$ . Thus,  $FV^{-1}$  is called the next generation matrix, and we define its spectral radius as the **basic reproduction number (ratio)**, that is,

$$R_0 = \rho(FV^{-1}).$$

# Linear stability

Let  $s(A)$  be the **stability modulus** of a matrix  $A$ , that is,  
$$s(A) = \max\{Re\lambda : \det(\lambda I - A) = 0\}.$$

P. van den Driessche and J. Watmough [Math. Biosci., 2002]  
proved the following result.

**Theorem 2.1.**  $R_0 - 1$  has the same sign as  $s(F - V)$ .

This theorem shows that the sign of  $R_0 - 1$  determine the stability of zero solution of  $\frac{dI}{dt} = (F - V)I$ . Thus,  $R_0$  is a threshold value for the disease invasion.

Applications to various autonomous epidemic models...

# An example

Consider a vector-host model for Dengue fever (Feng and Velasco-Hernandez, JMB, 1997):

$$\begin{aligned}\frac{dI}{dt} &= \beta_s SV - (b + \gamma)I, \\ \frac{dV}{dt} &= \beta_m MI - cV, \\ \frac{dS}{dt} &= b - bS + \gamma I - \beta_s SV, \\ \frac{dM}{dt} &= c - cM - \beta_m MI,\end{aligned}\tag{2.1}$$

where  $I$  is the number of infected hosts,  $V$  is the number of infected vectors,  $S$  is the number of susceptible hosts,  $M$  is the number of susceptible vectors,  $\beta_s$  and  $\beta_m$  are disease transmission coefficients. The birth rates have been scaled to  $b > 0$  for the host and  $c > 0$  for the vector.

# Basic reproduction number

It is easy to see that  $(0, 0, 1, 1)^T$  is the disease-free steady state. Then we have

$$F = \begin{pmatrix} 0 & \beta_s \\ \beta_m & 0 \end{pmatrix}, \quad V = \begin{pmatrix} b + \gamma & 0 \\ 0 & c \end{pmatrix}.$$

It follows that

$$R_0 = \rho(FV^{-1}).$$

**Exercise.** Compute the above  $R_0$ .

# Periodic ODE models

N. Baca  r and S. Guernaoui, *J. Math. Biol.*, 53(2006), 421  436.

W. Wang and X.-Q. Zhao, *J. Dyn. Diff. Equ.*, 20(2008), 699  717.

Consider the linear  $\omega$ -periodic ODE system (after linearization):

$$\frac{du}{dt} = F(t)u - V(t)u. \quad (3.1)$$

Let  $Y(t, s)$  be the evolution matrix of the internal evolution systems  $\frac{du}{dt} = -V(t)u$ , and let  $C_\omega$  be the space of continuous  $\omega$ -periodic (vector-valued) functions.

Wang and Zhao (2008) introduced a linear operator  $L$  on  $C_\omega$  by

$$(L\phi)(t) = \int_0^\infty F(t)Y(t, t-s)\phi(t-s)ds, \quad \forall t \in \mathbb{R},$$

and defined  $R_0 := r(L)$ . It turns out that the sign of  $R_0 - 1$  determines the stability of zero solution of system (3.1).

# Temporal and spatial heterogeneity

## Periodic models (continued):

N. Bacaër and E. H. Ait Dads, *J. Math. Biol.*, 65 (2012), 601-621.  
H. Inaba, *J. Math. Biol.*, 22(2012), 113-128.

## Impulsive models:

Y. Yang and Y. Xiao, *Nonlinear Analysis (RWA)*, 13(2012), 224-234.  
Z. Bai and X.-Q. Zhao, *J. Math. Biol.*, 80(2020), 1095-1117.  
(Periodic impulsive systems with delay)

## Infinite-dimensional population models:

H. Thieme, *SIAM J. Appl. Math.*, 70 (2009), 188-211.  
W. Wang and X.-Q. Zhao, *SIAM J. Appl. Dyn. Syst.*, 11(2012), 1652-1673.

# Temporal and spatial heterogeneity

## Almost periodic models:

B.-G. Wang and X.-Q. Zhao, *J. Dyn. Diff. Equ.*, 25(2013), 535–562.

L. Qiang, B.-G. Wang and X.-Q. Zhao, *J. Diff. Equ.*, 269(2020), 4440-4476. (Almost periodic and time delayed systems)

## Periodic and time-delayed models:

X.-Q. Zhao, *J. Dyn. Diff. Equ.*, 29(2017), 67-82.

X. Liang, L. Zhang and X.-Q. Zhao, *J. Dyn. Diff. Equ.*, 31(2019), 1247-1278. (Periodic abstract functional differential equations, and a general algorithm for  $R_0$ )

# Take-home exercise

You may watch 2011 movie “**Contagion**” from the internet.

# Thank you!