

Basics of Epidemic Models

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The compartments

- Most of the coming lectures will study epidemic compartment SIR / SEIR models, Here
 - S stands for Susceptible, $S(t) = \#$ of susceptible indiv. at time t ;
 - E for Exposed, $E(t) = \#$ of exposed indiv. at time t ;
 - I for Infectious, $I(t) = \#$ of infectious indiv. at time t ;
 - R for Recovered, $R(t) = \#$ of recovered indiv. at time t .
- We shall mainly consider no demography. Hence if $N = S(0) + E(0) + I(0) + R(0)$, $S(t) + E(t) + I(t) + R(t) = N$ for all $t > 0$.
- A susceptible who gets infected is first E, then I, then R, at which point he/she is immune, and in most of the lectures we shall assume that he/she remains immune for ever.
- However, we shall also consider SIS / SEIS and SIRS / SEIRS models, where infectious either become susceptible as soon as they cure, or lose their immunity after some time.

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Evolution of the epidemic

- While I, an individual has contacts with others at rate β . The contacted individual is chosen uniformly at random in the population. If the contacted individual is S, then the contact results in a new infection with probability p . Otherwise, nothing happens.
- In other words, an I has (potentially) infectious contacts at rate $\lambda = \beta \times p$. Such a contact results at time t in a new infection with probability $\bar{S}^N(t) = S(t)/N$.
- After being infected, an individual is first E for a random duration \mathcal{L} , then I for a random duration \mathcal{I} , after which he/she recovers.
- “At rate λ ” means according to a rate λ Poisson process. We always assume that the various Poisson processes are mutually independent, and independent of the $(\mathcal{L}, \mathcal{I})$ attached to the various indiv. (those are i.i.d.).
- An assumption which is not realistic, but makes the analysis easier is that the r.v. \mathcal{L} and \mathcal{I} are independent and both follow an exponential distribution.

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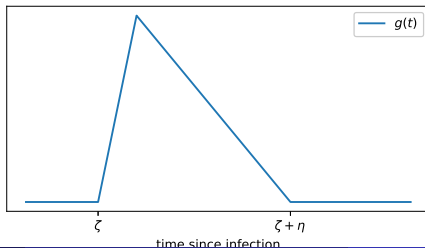
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The basic reproduction number R_0

- In our homogeneous model, R_0 is the mean number of infections produced by one I *at the start of the epidemic*, i.e. while $\bar{S}^N(t) = S(t)/N \simeq 1$.
- In the above described model, $R_0 = \lambda \mathbb{E}[\mathcal{I}]$. If the law of \mathcal{I} is $\text{Exp}(\mu)$, then $R_0 = \lambda/\mu$.
- **Generalisation : VI** Suppose now that the infectivity is a random function $\lambda(t)$ of the time since infection, those random functions being i.i.d. for the various individuals. Then $R_0 = \int_0^\infty \mathbb{E}[\lambda(t)] dt$. Note that in the previous case $\lambda(t) = \lambda 1_{\mathcal{I} > t}$.

Exercise : The two formulas are consistent !



$R_0 \leq 1$ vs. $R_0 > 1$

- If $R_0 < 1$, then in average each infectious infects less than one individual. The number of infected individuals tend to decrease. If we start with a small number of infected, there will be no major epidemic : the total number of infected will be small.
- It follows from the results on branching processes that the same is true in case $R_0 = 1$.
- If $R_0 > 1$ however, in average the number of infected individuals increases. As a result, starting from a small number of infected individuals, there is a positive probability that a major epidemic develops.

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The escape probability

- It follows from the above description that the probability that one given S escapes infection from one particular infectious individual is

$$\mathbb{E} [\exp (-\lambda \mathcal{I} / N)] .$$

- or in the VI model

$$\mathbb{E} \left[\exp \left(-N^{-1} \int_0^{\infty} \lambda(t) dt \right) \right] .$$

- **Exercise** Consider a SIR / SEIR epidemic with $\lambda = 1.8$ and \mathcal{I} follows the $\text{Exp}(1)$ distribution, in a village of size $N = 100$. What is the probability that a given S escapes the infection by a given I .

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The final size 1

- Let us now deduce a famous formula by an intuitive argument (to be justified later). Consider a SIR / SEIR epidemic in a population of size N with N large. The epidemic ends soon or later. What is the proportion \bar{I}^N of the population which gets infected at some stage?
- The probability of escaping the infection from all infected equals

$$\left\{ \mathbb{E} \left[\exp \left(-N^{-1} \int_0^\infty \lambda(t) dt \right) \right] \right\}^{N\bar{I}^N}.$$

- That probability should equal $1 - \bar{I}^N$. Hence

$$\begin{aligned} 1 - \bar{I}^N &= \left\{ \mathbb{E} \left[\exp \left(-N^{-1} \int_0^\infty \lambda(t) dt \right) \right] \right\}^{N\bar{I}^N} \\ &\simeq \left\{ 1 - N^{-1} \int_0^\infty \mathbb{E}[\lambda(t)] dt \right\}^{N\bar{I}^N} \end{aligned}$$

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The final size 2

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$$\begin{aligned}\log(1 - \bar{I}^N) &\simeq -\bar{I}^N \int_0^\infty \mathbb{E}[\lambda(t)] dt, \\ 1 - \bar{I}^N &\simeq \exp\left(-\bar{I}^N \int_0^\infty \mathbb{E}[\lambda(t)] dt\right).\end{aligned}$$

- We thus expect to have in the limit $N \rightarrow \infty$

$$1 - \bar{I} = \exp(-R_0 \bar{I}).$$

- Two remarks :
 - No all the population gets infected.
 - The final size depends only upon R_0 , and not upon further details of the model.
 - The final size is zero for $R_0 = 1$, and increases with R_0 .

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The SIR ODE model

Consider the SIR model in the Markovian case. One of the results which we will show in these lectures is that as $N \rightarrow \infty$, $N^{-1}(S^N(t), I^N(t), R^N(t)) \rightarrow (\bar{S}, \bar{I}, \bar{R})$, where $(\bar{S}, \bar{I}, \bar{R})$ solves the ODE

$$\begin{aligned}\frac{d\bar{S}}{dt}(t) &= -\lambda \bar{S}(t) \bar{I}(t), \\ \frac{d\bar{I}}{dt}(t) &= \lambda \bar{S}(t) \bar{I}(t) - \mu \bar{I}(t), \\ \frac{d\bar{R}}{dt}(t) &= \mu \bar{I}(t).\end{aligned}$$

λ = contact rate, μ = rate of recovery.

The start of the epidemic

- At the start of the epidemic, $\bar{S}(t) \simeq 1$, so we can linearize the second equation, yielding, after multiplication by N ,

$$\frac{dI^N}{dt}(t) = (\lambda - \mu)I^N(t),$$

whose solution is

$$I^N(t) = I^N(0) \exp[(\lambda - \mu)t].$$

- $R_0 > 1 \iff \rho := \lambda - \mu > 0$
- ρ is the growth rate. It is easy to estimate in practice (easily deduced from the doubling time = number of days necessary for the number of infected to double). What additional information is necessary for estimating R_0 once ρ is known?
- In the VI case, ρ is such that $\int_0^\infty \mathbb{E}[\lambda(t)]e^{-\rho t}dt = 1$, i.e., if we define $g(t) := R_0^{-1}\mathbb{E}[\lambda(t)]$, then ρ and R_0 are related by $\int_0^\infty g(t)e^{-\rho t}dt = R_0^{-1}$. $g(t)$ is the “generation time distribution”.
Exercise Verify the consistency with the above Markov/ODE model.

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