

RENEWAL EQUATIONS

a tribute to Kermack & McKendrick

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Building block

infectiousness as a function
of time τ elapsed since exposure

τ aka "infection-age"

outbreak

no demographic turnover

permanent immunity

second part

the simplest contact structure

in a

heterogeneous host population

but for the time being

every individual subject to
the same force-of-infection
and equally susceptible

$$\frac{dS}{dt} = -FS$$

incidence

f-o-i

KM 27

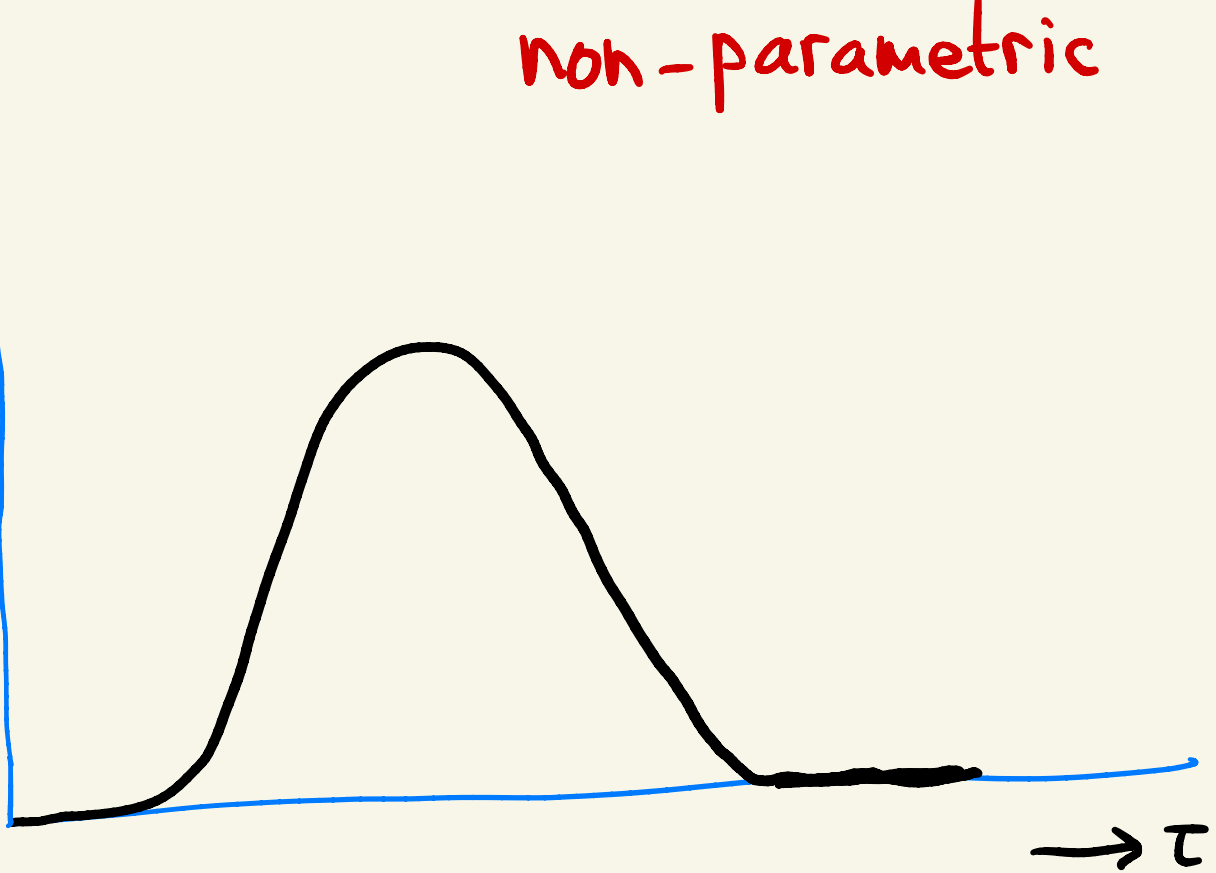
$$F(t) = \int_0^{\infty} A(\tau) \text{ incidence}(t-\tau) d\tau$$

$A(\tau) =$ **expected** contribution
to the f-o-i
at time τ after
becoming infected

input - output

A ↑

non-parametric




parametric example

$$A(\tau) = U e^{\tau \Sigma} V$$

$$U, V \in \mathbb{R}^n \quad \Sigma \in \mathbb{R}^{n \times n}$$

V initial state

individual level



Σ state transitions

U output

i-State representation

↑
fake

Analysis

Linearization:

$$\text{incidence} = F \cancel{S} / N$$

linear Renewal Equation ← RE

$$F(t) = N \int_0^{\infty} A(\tau) F(t-\tau) d\tau$$

Lotka , Feller

Ansatz

$$F(t) = e^{\lambda t}$$

Euler Lotka

$$1 = N \int_0^{\infty} e^{-\lambda \tau} A(\tau) d\tau$$

real root

$$\lambda = r$$



Malthusian parameter

$$R_0 := N \int_0^{\infty} A(\tau) d\tau$$

Basic Reproduction Number

$$\text{sign } r = \text{sign}(R_0 - 1)$$

emerging disease

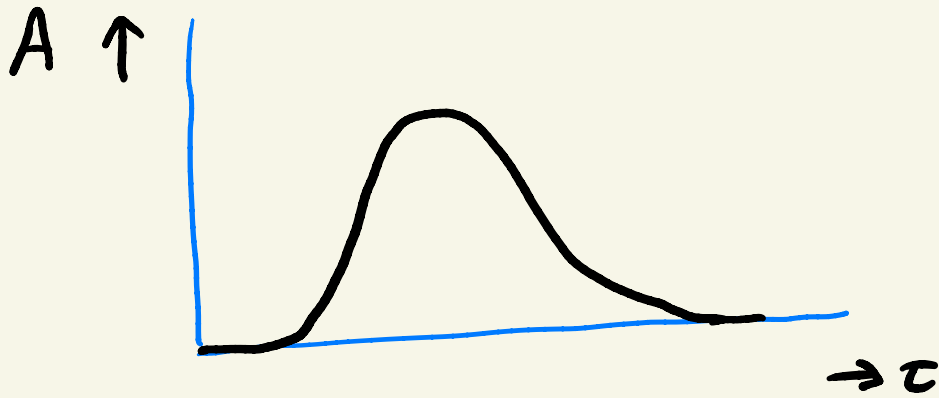
data \Rightarrow estimate of r

wanted

estimate of R_0



in view of control



↑
shape matters

generation interval

Nonlinear

$$F(t) = \int_0^{\infty} A(\tau) \left\{ -S'(t-\tau) \right\} d\tau$$

↑
integrate over $(-\infty, t]$

$$s(t) := \frac{S(t)}{N}$$

$$s = e^{-w}$$

w = cumulative f-o-i

$$w(t) = \int_0^{\infty} A(\tau) \Psi(w(t-\tau)) d\tau$$

scalar nonlinear RE

$$\Psi(w) = N (1 - e^{-w})$$

Interlude

RE

is a

Delay Equation

A **delay equation** is a rule for extending a function of time towards the **future** on the basis of the (assumed to be) **known past**

$$DE = DDE \cup RE$$

A **dynamical system** is defined
by **translation** along the
extended function

$$S_t(\theta) = s(t+\theta) \quad , \quad \theta \leq 0$$

end of interlude

Final Size Equation

$$s(\infty) = e^{-R_0(1-s(\infty))}$$

- probabilistic consistency condition
- "shape" of A is irrelevant

Compartmental Models

J.A.J. Metz
M. Gyllenberg

$$A(\tau) = U e^{\tau \Sigma} V$$

$$U, V \in \mathbb{R}^n \quad \Sigma \in \mathbb{R}^{n \times n}$$

$$w = U \cdot Q$$

$$\frac{dQ}{dt} = \sum Q + \Psi(U \cdot Q) V$$

recall: $\Psi(w) = N(1 - e^{-w})$

integrated
form

Y, V, Σ, U

may incorporate

asymptomatic

quarantained

hospitalized

....

individuals

put

$$Q(t) = \int_{-\infty}^t Y(\sigma) d\sigma$$

then

$$\frac{dS}{dt} = -FS$$

$$\frac{dY}{dt} = \sum \gamma + (FS)V$$

$$F = U \cdot Y$$

standard form

SEIR example

$$\frac{d}{dt} \begin{pmatrix} E \\ I \end{pmatrix} = \begin{pmatrix} -\gamma & 0 \\ \gamma & -\alpha \end{pmatrix} \begin{pmatrix} E \\ I \end{pmatrix} + FS \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

Diagrammatic annotations for the first equation:
- A blue arrow labeled γ points from the bottom element I of the vector $\begin{pmatrix} E \\ I \end{pmatrix}$ to the top element E .
- A blue arrow labeled Σ points from the bottom element I of the matrix $\begin{pmatrix} -\gamma & 0 \\ \gamma & -\alpha \end{pmatrix}$ to the top element $-\gamma$.
- A blue arrow labeled v points from the bottom element 0 of the vector $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ to the top element 1 .

$$F = \begin{pmatrix} 0 & \beta \end{pmatrix} \begin{pmatrix} E \\ I \end{pmatrix} = \beta I$$

Diagrammatic annotation for the second equation:
- A blue arrow labeled u points from the bottom element I of the vector $\begin{pmatrix} E \\ I \end{pmatrix}$ to the top element 0 of the matrix $\begin{pmatrix} 0 & \beta \end{pmatrix}$.

why are compartmental
models

omnipresent

?

i) omnipresence is

self reinforcing

ii) RE are unfamiliar

iii) lack of tools for RE

Discrete time

← M. Kreck

$$s(t+1) = e^{-\hat{f}(t)} s(t)$$

time unit one day

$$\hat{F}(t) = \sum_{k=m_1}^{\infty} A_k (1 - e^{-\hat{F}(t-k)}) s(t-k)$$

*m*₂

*m*₁

essential parameters

M. Bootsma

H. Othmer

R. Planqué

PNAS (2021) 118 39

peak { when ?
how high

Herd Immunity Threshold

$$S_{\text{crit}} = \frac{1}{R_0}$$

vaccin induced } immunity
disease induced }

HIT

disease induced

in the presence of

heterogeneity

(some more susceptible than others)

G. Gomes

T. Britton

F. Jülicher

N. Goldenfeld

M. Bootsma

D. Chan

H. Inaba

(in progress)


Trait $x \in \Omega$ ← static

$$s(t, x) = e^{-N \int_0^t \int_{\Omega} A(\tau, x, \xi) [1 - s(\tau, \xi)] \Phi(d\xi) d\tau}$$

↑
trait
distribution

abstract RE

For general theory see book
by H. Inaba and recent paper
by E. Franco e.a.


arXiv: 2201.05323

Separable Mixing:

$$A(\tau, x, \xi) = a(x) b(\tau) c(\xi)$$

independence makes "life" easy

$$s(t, x) = e^{-a(x)w(t)}$$

$w = \text{cumulative } f_{-o-i}$

side remark (for later use)

$$s(t, x_1) = \left(s(t, x_2) \right)^{\frac{a(x_1)}{a(x_2)}}$$

$$w(t) = \int_0^{\infty} b(\tau) \Psi(w(t-\tau)) d\tau$$

Scalar RE

$$\Psi(w) := N \int_{\Omega} c(\xi) (1 - e^{-a(\xi)w}) \Phi(d\xi)$$

Note $\Psi'(0) = N \int_{\Omega} c(\xi) a(\xi) \Phi(d\xi)$

side remark

$$\text{if } a(x) = 1 \quad \forall x \in \Omega$$

then

$$\Psi(w) = \bar{c} N(1 - e^{-w})$$

$$\bar{c} := \int_{\Omega} c(\xi) \Phi(d\xi)$$

if

heterogeneity only affects
infectiousness

and numbers are large

simply take the

average infectiousness

end of side remark

$$R_0 = \Psi'(0) \int_0^{\infty} b(\tau) d\tau$$

$$R_{\text{eff}} = \Psi'(w) \int_0^{\infty} b(\tau) d\tau$$

HIT

$$\Psi'(w) = \frac{\Psi'(0)}{R_0} \Rightarrow w = \tilde{w}$$

$$\tilde{s} = \int_{\Omega} e^{-a(\xi)\tilde{w}} \Phi(d\xi)$$

Special Case

$$\Omega = (0, \infty)$$

$$a(x) = x$$

$$\int_{\Omega} x \Phi(dx) = 1$$

Gamma Distribution

Novozhilov

Φ has density

$$x \mapsto \frac{p^p}{\Gamma(p)} x^{p-1} e^{-px}$$

$$\text{Variance} = \frac{1}{p}$$

$$\hat{\Phi}(\lambda) = \left(\frac{\lambda}{p} + 1 \right)^{-1}$$

$$\Psi(w) = 1 - \left(\frac{w}{p} + 1\right)^{-q}$$

$$c(\xi) = 1 \quad \Rightarrow \quad q = p$$

$$c(\xi) = \xi \quad \Rightarrow \quad q = p + 1$$

HIT does not depend on $b(\tau)$

$$c(\xi) = 1$$

$$\tilde{s} = R_0^{-1} + \frac{1}{p+1}$$

$$c(\xi) = \xi$$

$$\tilde{s} = R_0^{-1} + \frac{1}{\frac{1}{2}p+1}$$

When $b(\tau) = U e^{\tau \Sigma} V$

then $w = U \cdot Q$

with

$$\frac{dQ}{dt} = \Sigma Q + \Psi(U \cdot Q) V$$

normalisation

$$a(\bar{x}) = 1$$

$$c(\bar{x}) = 1$$

for "representative" $\bar{x} \in \Omega$

standard form

$$\bar{s}(t) = s(t, \bar{x})$$

$$\frac{d\bar{s}}{dt} = -F\bar{s}$$

$$\frac{dY}{dt} = \sum Y + F \boxed{\Psi'(-\ln \bar{s})} V$$

$$F = u \cdot Y$$

$$\Psi'(-\ln \bar{s}) = N \int_{\Omega} c(\xi) a(\xi) \bar{s}^{a(\xi)} \mathbb{I}(d\xi)$$

$$S_{\text{tot.}} = \int_{\Omega} \bar{s}^{a(\xi)} \Phi(d\xi)$$

Example : two types

A not-vaccinated $a=1$ $c=1$

B vaccinated $a=\varepsilon_1$ $c=\varepsilon_2$

$$\Psi'(-\ln \bar{s}) = N_A \bar{s} + N_B \varepsilon_1 \varepsilon_2 \bar{s}^{\varepsilon_1}$$

$$S_{\text{tot.}} = \frac{N_A}{N} \bar{s} + \frac{N_B}{N} \bar{s}^{\varepsilon_1}$$

Gamma Distribution

$$\frac{ds_{\text{tot.}}}{dt} = -F s_{\text{tot.}}^{1+\frac{1}{p}}$$

$$\frac{dY}{dt} = \sum Y + F H(s_{\text{tot.}}) Y$$

$$F = u \cdot Y$$

$$H(s) = \begin{cases} N s^{1 + \frac{1}{p}} & \leftarrow c(\xi) = 1 \\ N \left(1 + \frac{1}{p}\right) s^{1 + \frac{2}{p}} & \leftarrow c(\xi) = \xi \end{cases}$$

Summary

the KM 1927 model

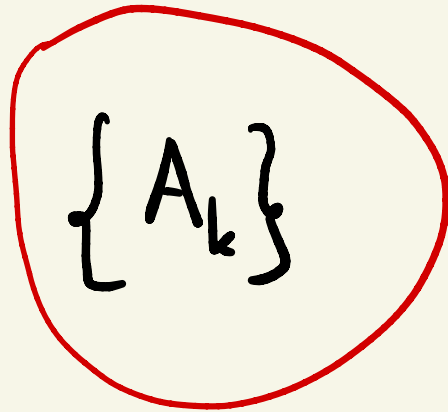
takes the form of a

RE

the kernel

$A(\tau)$

resp.



discrete
time

is the

essential ingredient

compartmental models

miss the mark

sojourn times are not
exponentially distributed

RE

exhibit remarkable

bookkeeping efficiency

also when dealing with

heterogeneity

Appeal

next time you formulate
an epidemic model,
please try a

RE