



**Massey University**

# **Spatio-temporal and Network Modelling of Diseases**

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# The next generation matrix and $\mathcal{R}_0$

- We define the **Basic Reproduction Number**  $\mathcal{R}_0$  to be the largest eigenvalue of the next generation matrix  $\mathbf{K}$ , where  $K_{ij}$  is the expected number of infections of Type  $i$  due to a single infection of Type  $j$ , in a fully susceptible population.
- If the system is in state  $\mathbf{w}$ , then (in a susceptible population)

$$\lim_{n \rightarrow \infty} \frac{1}{(\mathcal{R}_0)^n} \mathbf{K}^{n+1} \mathbf{w} = \mathbf{K} \mathbf{v} = \mathcal{R}_0 \mathbf{v}$$

- $\mathbf{K}$  is dimensionless, and relates *infection-generations*.

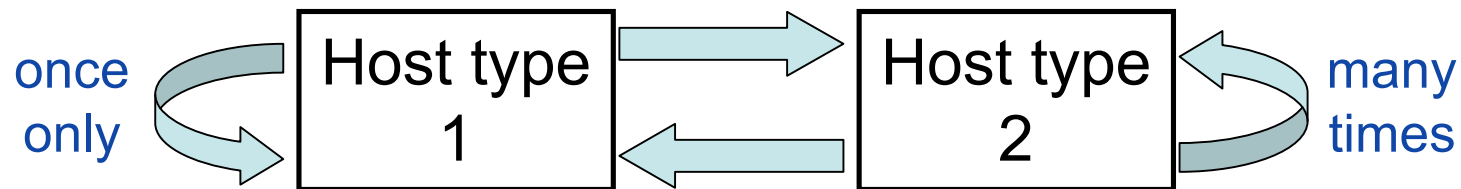
# For example:

- For an STI, Types could be male, female.
- For a vector transmitted infection, Types could be human, mosquito.
- The **Next Generation Matrix** is

$$\mathbf{K} = \begin{pmatrix} 0 & K_{12} \\ K_{21} & 0 \end{pmatrix}$$

- The **Basic Reproduction Number** is  $\mathcal{R}_0 = \sqrt{K_{12}K_{21}}$ .

# The type reproduction number



- The expected number of infected hosts of Type 1 that would arise from a single infected host of Type 1 in an otherwise susceptible population.
- This only makes sense if  $\mathcal{R}_0$  for Type 2 is less than one.

# The type reproduction number: 2 types

- The type reproduction number is like the basic reproduction number, but focuses on a subset of host *types at infection*.
- For example, with two types

$$\begin{aligned}\mathcal{T} &= K_{11} + K_{12}K_{21} + K_{12}K_{22}K_{21} + K_{12}K_{22}^2K_{21} + \dots \\ &= K_{11} + \frac{K_{12}K_{21}}{1 - K_{22}}\end{aligned}$$

This requires  $K_{22} < 1$ .

- If  $K_{22} > 1$  then Type 2 is a **reservoir of infection**.

# The type reproduction number: many types

- To focus on  $\ell$  out of  $n$  types, define a projection matrix  $P_{ij} = 1$  for  $i = 1 \dots \ell$ ,  $P_{ij} = 0$  otherwise.
- Now define a *reduced* next generation matrix:

$$\mathbf{M} = \mathbf{K} (\mathbf{I} - (\mathbf{I} - \mathbf{P}) \mathbf{K})^{-1}$$

- This requires  $\|(\mathbf{I} - \mathbf{P}) \mathbf{K}\| < 1$ . If not, the *out of focus* types form a **reservoir of infection**.
- The type reproduction number is  $\mathcal{T} = \|\mathbf{M}\|$ .

# Infection control

- It has been proved that  $\mathcal{T} < 1 \Leftrightarrow \mathcal{R}_0 < 1$ .
- To eliminate infection you must:
  - protect a proportion  $v$  of all types, where  $v > 1 - \frac{1}{\mathcal{R}_0}$ ;  
or of focus types, where  $v > 1 - \frac{1}{\mathcal{T}}$ ;
  - reduce transmission time of all types by a proportion  $w > 1 - \frac{1}{\mathcal{R}_0}$ ; or of focus types by  $w > 1 - \frac{1}{\mathcal{T}}$ .

# References

- Roberts & Heesterbeek: A new method for estimating the effort required to control infectious diseases. *Proc. R Soc. B* 270 (2003):1359-64.
- Heesterbeek & Roberts: The type-reproduction number  $\mathcal{T}$  in models for infectious disease control. *Math. Biosci.* 206 (2007): 3-10.
- Roberts: The pluses and minuses of  $\mathcal{R}_0$ . *J. R. Soc. Interface* in press.



# Structured integral equation models

- Integral equation models provide an ideal vehicle for analysing epidemics of emerging infectious diseases.
- Contact and transmission functions are explicitly represented, and easily modified to reflect control interventions. We are not restricted to flow rates between compartments.
  - ... most people keep referring to the [*SIR*] model as *the* Kermack and McKendrick model. This should stop! (Diekmann *et al.* 1995)
  - This is the classic Kermack-McKendrick (1927) model (Murray 2002).

# The Kermack-McKendrick model

The incidence of an emerging infection  $\iota(t)$  may be calculated from:

$$\iota(t) = \delta(t) + s(t) \int_0^{\infty} p(\tau) \kappa(\tau) \iota(t - \tau) d\tau$$

- $\delta(t)$  is the incidence of the index case;
- $p(\tau)$  is the probability of transmission given contact;
- $\kappa(\tau)$  is the contact rate;
- $\tau$  is the time since exposure to infection.

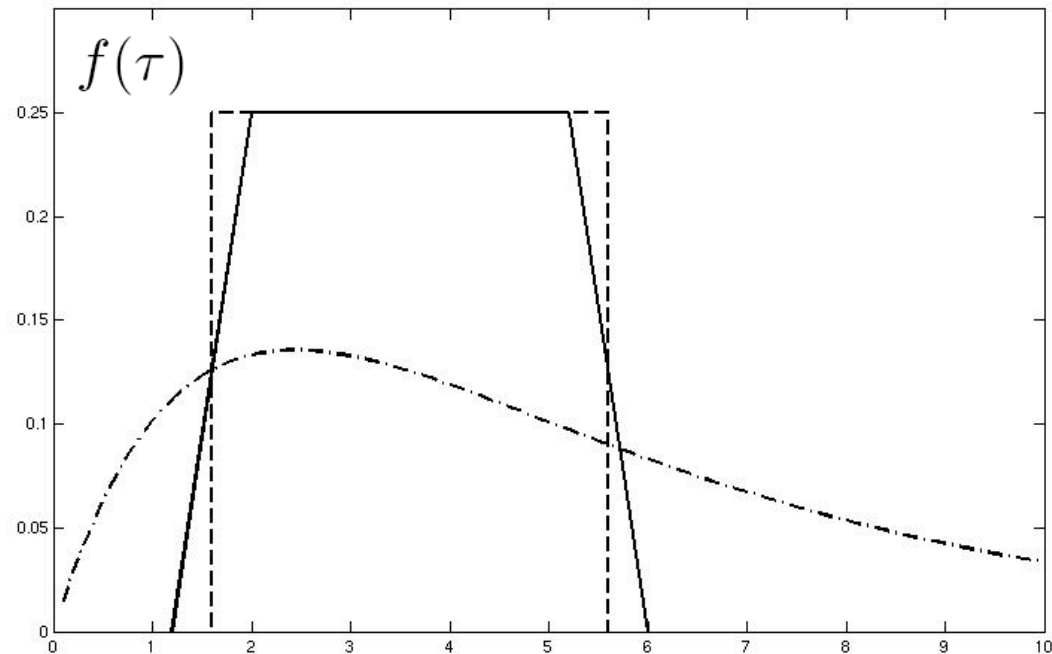
# The Kermack-McKendrick model - 2

$$i(t) = \delta(t) + \mathcal{R}_0 s(t) \int_0^\infty f(\tau) i(t - \tau) d\tau$$

- $\mathcal{R}_0$  is the basic reproduction number;
- $f(\tau)$  is a probability density;
- $N$  is the population size,  $s(t) = S(t)/N$ .
- The numbers in the population susceptible and removed at time  $t$  are

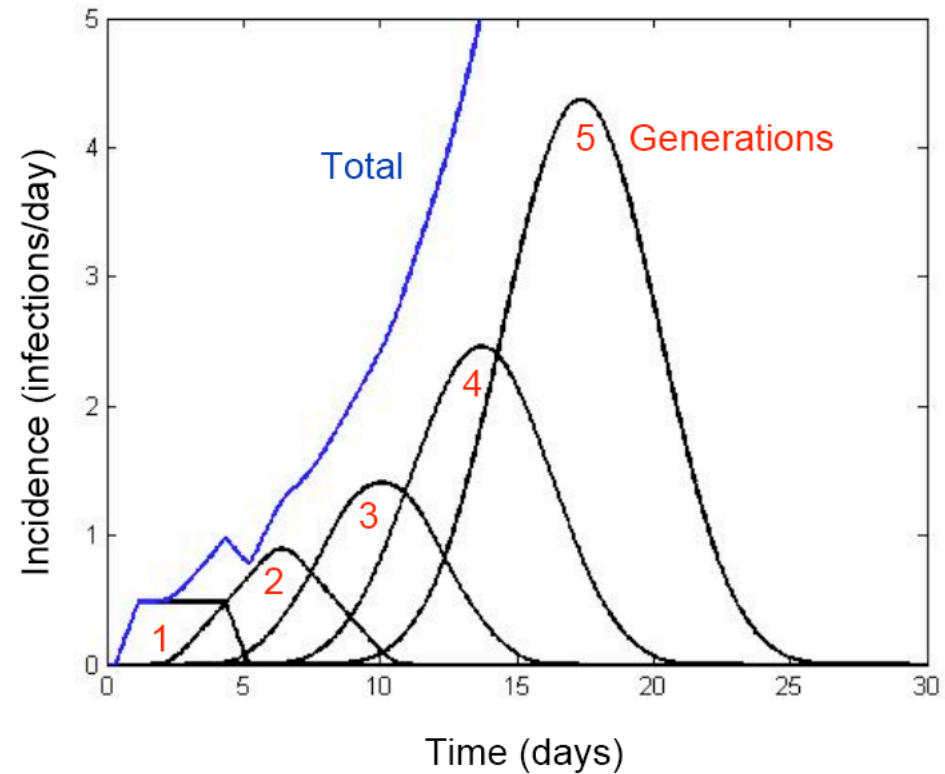
$$S(t) = N - \int_0^t i(u) du \quad R(t) = \int_0^\infty g(\tau) i(t - \tau) d\tau$$

# The integral kernel



$$f(\tau) = \frac{\gamma\nu}{\gamma - \nu} (e^{-\nu\tau} - e^{-\gamma\tau})$$

# Infection generations

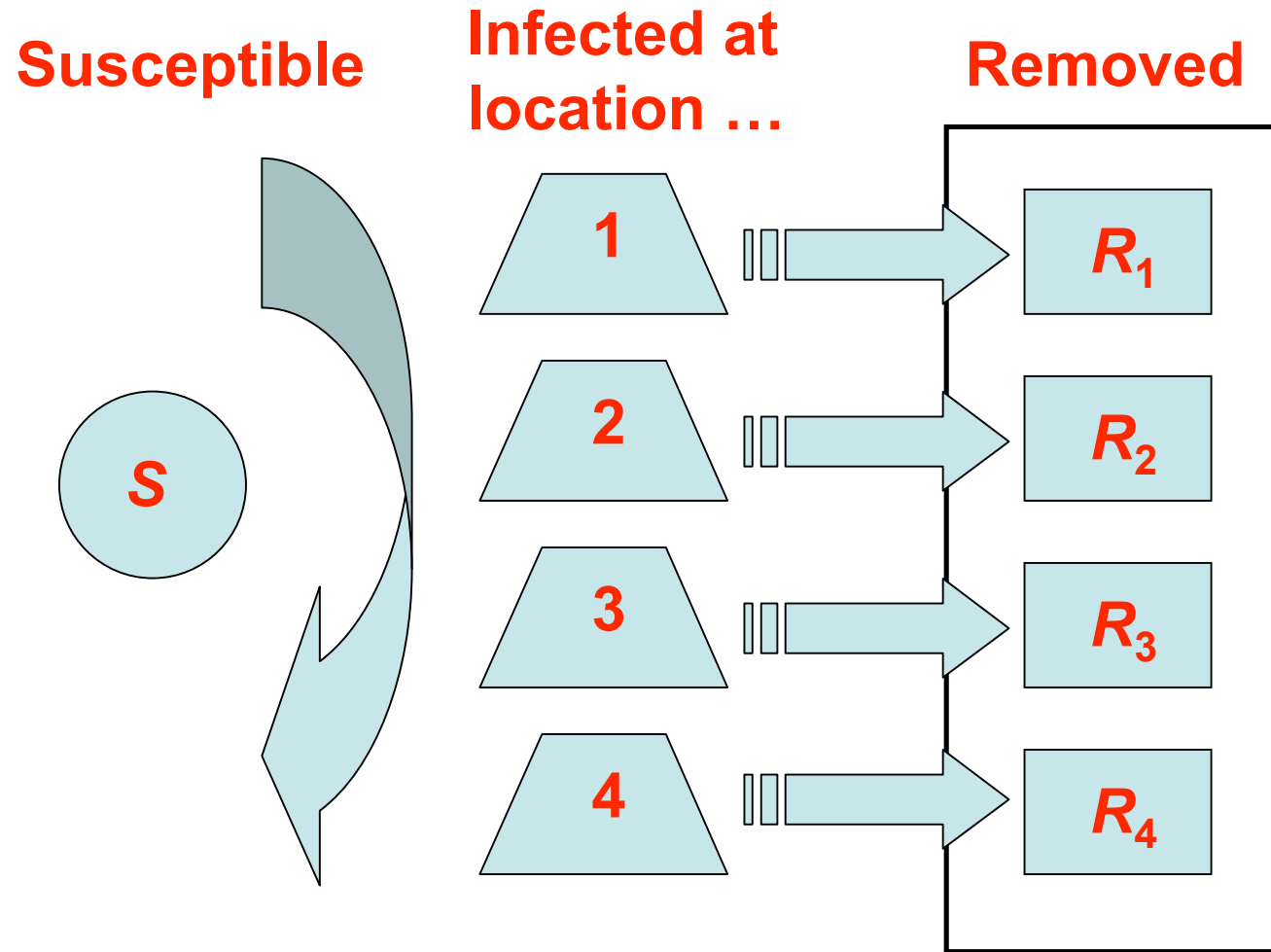


- With  $s(t) \approx 1$  the first generation solution is  $v_1(t) = \mathcal{R}_0 f(t)$ .
- Subsequent infection generations are found from

$$v_{k+1}(t) = \mathcal{R}_0 s(t) \int_0^\infty f(\tau) v_k(t - \tau) d\tau$$

- The result is  $v(t) = \sum_{k=1}^\infty v_k(t)$ .

# Adding some structure



# Adding some structure - 2

- The location of exposure confers a unique infection type.

$$\mathbf{z}(t) = \delta(t)\mathbf{e} + \mathcal{R}_0 s(t) \int_0^\infty f(\tau) \mathbf{W} \mathbf{z}(t - \tau) d\tau$$

where  $\|\mathbf{W}\| = 1$ . For example:

$$\mathbf{W} = \begin{pmatrix} 0 & w_1 & w_1 & w_1 \\ w_2 & w_2 & 0 & w_2 \\ w_3 & 0 & w_3 & w_3 \\ w_4 & w_4 & w_4 & w_4 \end{pmatrix}$$

1. Within the household;
2. At school/work;
3. In the wider community;
4. Within a healthcare facility.

# Adding some structure - 3

- The incidence of infection is

$$\mathbf{z}(t) = \delta(t)\mathbf{e} + \mathcal{R}_0 s(t) \int_0^\infty f(\tau) \mathbf{W} \mathbf{z}(t - \tau) d\tau$$

- In the beginning  $s(t) \simeq 1$ , so

$$\bar{\mathbf{z}}(\omega) \simeq (\mathbf{I} - \mathcal{R}_0 \bar{f}(\omega) \mathbf{W})^{-1} \mathbf{e}$$

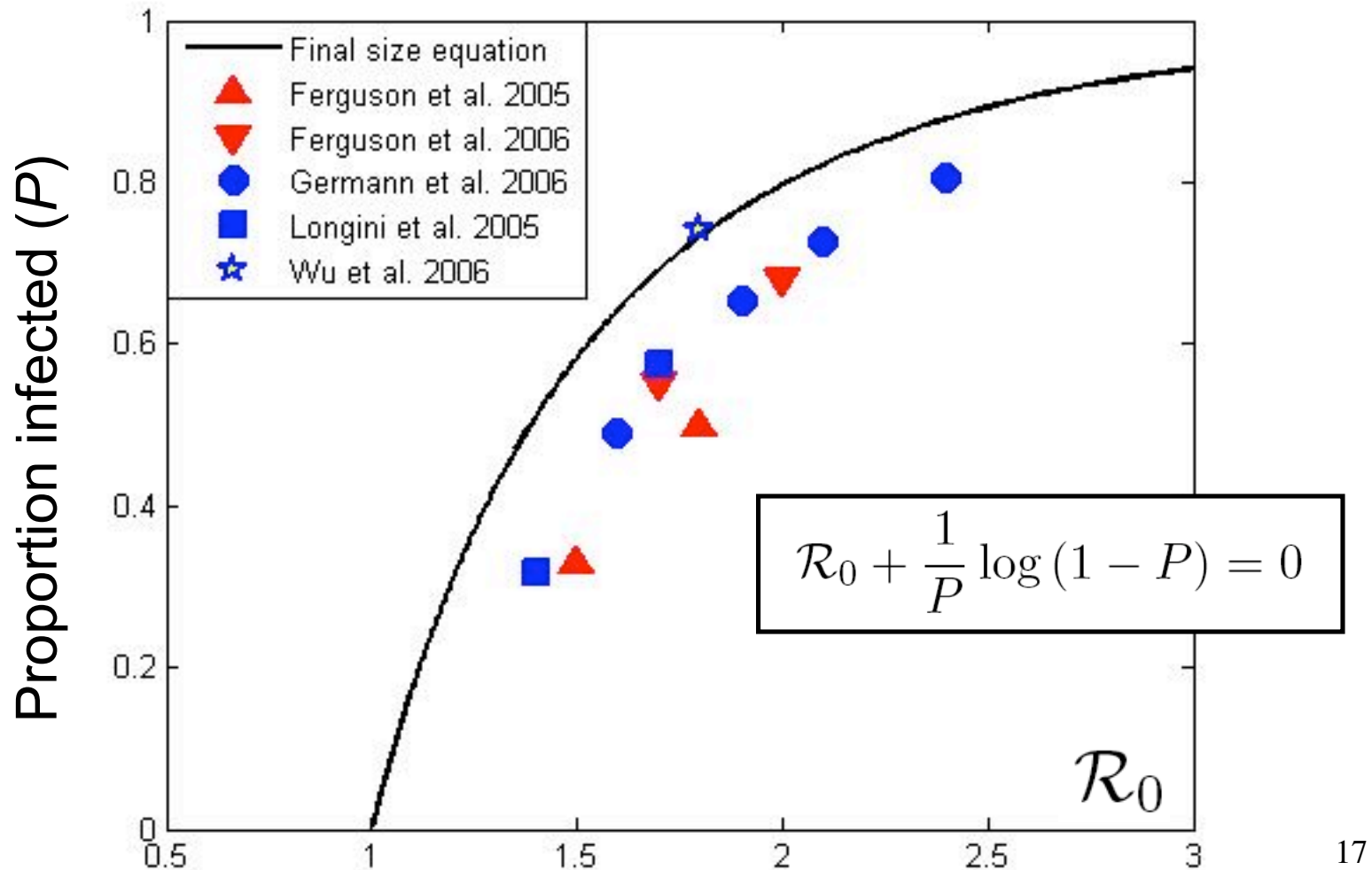
and  $|\mathbf{z}| \simeq e^{rt}$  where  $\mathcal{R}_0 \bar{f}(r) = 1$ .

- The proportion infected in an epidemic is  $|\mathbf{p}|$  where

$$\mathcal{R}_0 + \frac{1}{|\mathbf{p}|} \log(1 - |\mathbf{p}|) = 0 \quad \mathbf{W}\mathbf{p} = \mathbf{p}$$



# The final size equation



# References

- Roberts: Modelling strategies for minimizing the impact of an imported exotic infection. *Proc. R. Soc. B* 271(2004):2411-15.
- Aldis & Roberts: An integral equation model for the control of a smallpox outbreak. *Math. Biosci.* 195(2005):1-22.
- Roberts *et al.*: A model for the spread and control of pandemic influenza in an isolated geographical region. *J. R. Soc. Interface* 4(2007): 325-30.

# What about networks?

- A statistic as useful as  $\mathcal{R}_0$ , but defined on a network must:
  - be easily derived from the network structure;
  - have a biological interpretation;
  - tell us something useful -
    - \* threshold property for (large) epidemics,
    - \* control effort for elimination,
    - \* final size of the epidemic,
    - \* ...
- Is it possible to define and derive such a property?

# The end

