

# Intentional infection as a method of population level disease control

Newborn infection

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- What are the historical examples?
  - ▶ Variolation of smallpox
  - ▶ Pox party
- This method is out of date, banned in some places, why do we care?
  - ▶ In history, the mechanisms and benefits of intentional infection on a population level was not quite understood.
  - ▶ New application to immunology, e.g. Transmissible vaccine.

# Method

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$$\begin{aligned}\frac{dS}{dt} &= \mu - \beta SI - \mu S, \\ \frac{dI}{dt} &= \beta SI - \gamma I - \mu I, \\ \frac{dR}{dt} &= \gamma I - \mu R.\end{aligned}\tag{1}$$

Where  $S$ ,  $I$  and  $R$  represent susceptible, infected and recovered.

- $\mu$  is birth/death rate
- $\beta$  is transmission rate
- $\gamma$  is recovery rate

# Method

## Assumptions:

- No difference between intentionally infected and naturally infected individuals.
- No disease induced mortality.
- Birth and natural death rate are the same, the total population remains constant.
- Latent period is short enough to be ignored.
- All susceptible individuals are equally likely to be infected, and all infected individuals are equally infectious.



# Method

With intentional infection on newborn,

$$\begin{aligned}\frac{dS}{dt} &= \mu(1 - p) - \beta SI - \mu S, \\ \frac{dI}{dt} &= \beta SI + \mu p - \gamma I - \mu I, \\ \frac{dR}{dt} &= \gamma I - \mu R.\end{aligned}\tag{2}$$

Where  $p$  is the proportion of newborn intentionally infected.

Non-dimensionalize the above system by scaling time, by

$$\tau = (\gamma + \mu)t, \quad (3)$$

which yields

$$\frac{dS}{d\tau} = \epsilon(1 - p) - \mathcal{R}_0 SI - \epsilon S, \quad (4a)$$

$$\frac{dI}{d\tau} = \mathcal{R}_0 SI + \epsilon p - I, \quad (4b)$$

where  $\epsilon = \frac{\mu}{\gamma + \mu}$ ,  $\mathcal{R}_0 = \frac{\beta}{\gamma + \mu}$ .

# Equilibria

Solving equations above to find equilibria, we obtained only one set of solution,

$$\hat{S} = \frac{1}{\mathcal{R}_0} - \frac{2p}{(\mathcal{R}_0 - 1) + \sqrt{(\mathcal{R}_0 - 1)^2 + 4\mathcal{R}_0 p}}, \quad (5a)$$

$$\hat{I} = \frac{\epsilon(\mathcal{R}_0 - 1) + \epsilon\sqrt{(\mathcal{R}_0 - 1)^2 + 4\mathcal{R}_0 p}}{2\mathcal{R}_0}. \quad (5b)$$

- $\hat{I} \neq 0$  for all  $p$  between 0 and 1.
- The equilibrium is an Endemic Equilibrium (EE).

# Stability of Equilibria

Jacobian Matrix,

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$$\mathcal{J} = \begin{bmatrix} -\mathcal{R}_0 I - \epsilon & -\mathcal{R}_0 S \\ \mathcal{R}_0 I & \mathcal{R}_0 S - 1 \end{bmatrix}. \quad (6)$$

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  - ▶ We need to have separate CFP for each of them.

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- It is challenging to compare the results of this model to others for the purpose of observing any advantages.
- Define “Have advantage” by comparing total number of deaths.
- If intentionally infected cases have the same case fatality proportion (CFP) as naturally infected cases, then it will result in more deaths.
  - ▶ We need to have separate CFP for each of them.
- We need to divide  $I$  into two separate infective classes.  $V$  for intentionally infected class,  $I$  for naturally infected class.

# Model with disease induced mortality rate

Therefore, our model becomes,

$$\frac{dS}{d\tau} = \epsilon(1 - p) - \mathcal{R}_0 S(V + I) - \epsilon S, \quad (7a)$$

$$\frac{dV}{d\tau} = \mathcal{R}_0 S V + \epsilon p - V, \quad (7b)$$

$$\frac{dI}{d\tau} = \mathcal{R}_0 S I - I, \quad (7c)$$

$$\frac{dM}{d\tau} = p_V(1 - \epsilon)V + p_I(1 - \epsilon)I, \quad (7d)$$

$$\frac{dR}{d\tau} = (1 - p_V)(1 - \epsilon)V + (1 - p_I)(1 - \epsilon)I - \epsilon R, \quad (7e)$$

Where  $p_V$  and  $p_I$  represent the CFP for intentionally infected and naturally infected cases, respectively.

# Equilibria

If  $p \neq 0$ , the only equilibrium is,

$$\hat{S} = \frac{1}{\mathcal{R}_0} - \frac{2p}{(\mathcal{R}_0 - 1) + \sqrt{(\mathcal{R}_0 - 1)^2 + 4\mathcal{R}_0 p}}, \quad (8a)$$

$$\hat{V} = \frac{\epsilon(\mathcal{R}_0 - 1) + \epsilon\sqrt{(\mathcal{R}_0 - 1)^2 + 4\mathcal{R}_0 p}}{2\mathcal{R}_0}, \quad (8b)$$

$$\hat{I} = 0. \quad (8c)$$

- Naturally infected cases cease to exist at EE.
  - ▶ Helpful for disease eradication.

# Effect of intentional infection on total mortality

## Smallpox

**Table:** Model parameters and smallpox values.

Symbol	Meaning	Value
$\frac{1}{\mu}$	Average lifespan	50 years
$\frac{1}{\gamma}$	Mean infectious period	22 days
$\mathcal{R}_0$	Basic reproduction number	4.5
$p_V$	Intentionally infected CFP	0.01
$p_I$	Naturally infected CFP	0.3

# Effect of intentional infection on total mortality

Mortality rate at EE

$$\left. \frac{dM}{d\tau} \right|_{EE} = \frac{p_V(1-\epsilon)\epsilon(\mathcal{R}_0 - 1) + p_V(1-\epsilon)\epsilon\sqrt{(\mathcal{R}_0 - 1)^2 + 4\mathcal{R}_0\rho}}{2\mathcal{R}_0}, \quad (9)$$

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- Mortality rate at EE increases as  $p$  increases.
- In the long run, a larger proportion of intentional infection will lead to more deaths.



# Initial state being at equilibrium

In history, intentional infection was introduced when the population is at equilibrium, which is the equilibrium for  $p = 0$ .

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$$\hat{S} = \frac{1}{\mathcal{R}_0}, \quad (10a)$$

$$\hat{V} = 0, \quad (10b)$$

$$\hat{I} = \epsilon(1 - \frac{1}{\mathcal{R}_0}) \quad (10c)$$

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Since the new equilibrium has  $\hat{I} = 0$ , define reaching equilibrium by  $I \leq 10^{-6}$ .

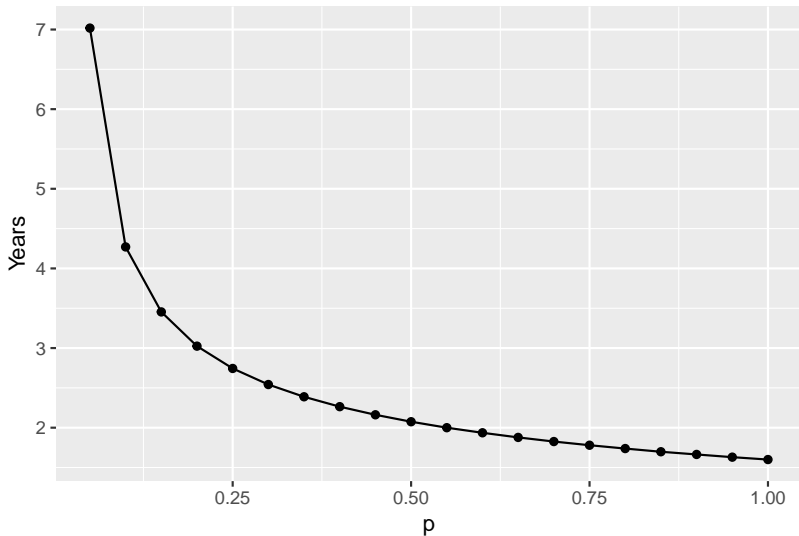
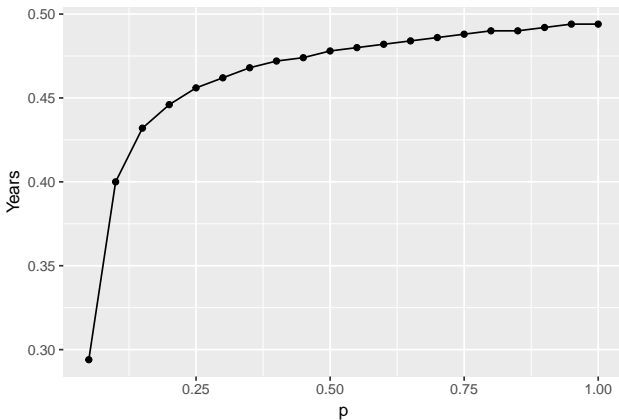


Figure: Determination of time taken to reach equilibrium



**Figure:** Time to advantage, as a function of  $p$

With a lower proportion of intentional infection, we can gain advantages relative faster.

# Possibility of disease eradication

$\hat{I} = 0$  at EE. We can consider the Naturally infected cases already been eradicated at EE.

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For intentionally infected cases, if we stop intentional infection after we reach EE, can they burn out?



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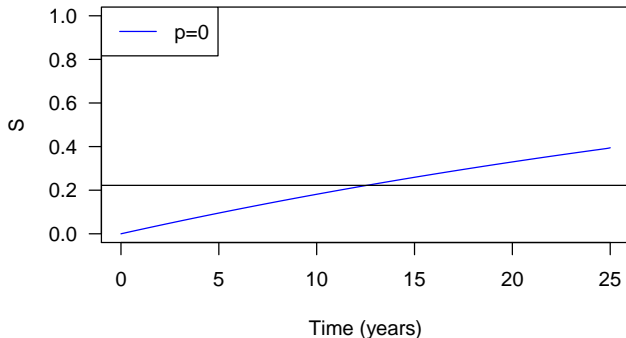
Threshold for susceptible:  $S = \frac{1}{\mathcal{R}_0}$

If  $S$  remains below this threshold until  $V$  goes extinct, then we can achieve complete eradication of this disease.

# Possibility of disease eradication

Example:  $p = 1$  initially, stop intentional infection after reaching EE,

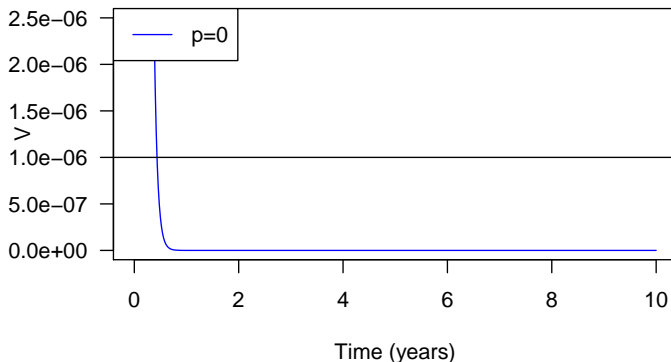
## Increase of $S$ after we stop intentional infection



**Figure:** For more than 10 years after we stop intentional infection,  $S < \frac{1}{R_0}$

# Possibility of disease eradication

**V as a function of time, after we stop intentional infection**



**Figure:** It takes less than 1 year for  $V$  to fall below  $1 \times 10^{-6}$

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# Challenges to this method

- Identify susceptible individuals

# Conclusion

- Intentional infection has positive effects on population level disease control.
- Further study is required for determining intentional infection strategies to optimize this method.

# References

-  Pauline van den Driessche *Reproduction numbers of infectious disease models*, Infectious Disease Modelling, Volume 2, Issue 3, August 2017, Pages 288-303.
-  R.M. Anderson, R.M. May *Infectious diseases of humans: Dynamics and control*, Oxford University Press, Oxford, UK (1991)