

# Infectious Disease Models 3

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# Key Quantities for Infectious Disease Models: Parameters

- Contacts per susceptible per unit time:  $c$ 
  - e.g. 20 contacts per month
  - This is the number of contacts a given susceptible will have with *anyone*
- Per-infective-with-susceptible-contact transmission probability:  $\beta$ 
  - This is the per-contact likelihood that the pathogen will be transmitted from an infective to a susceptible with whom they come into a single contact.

# Recall: Our model

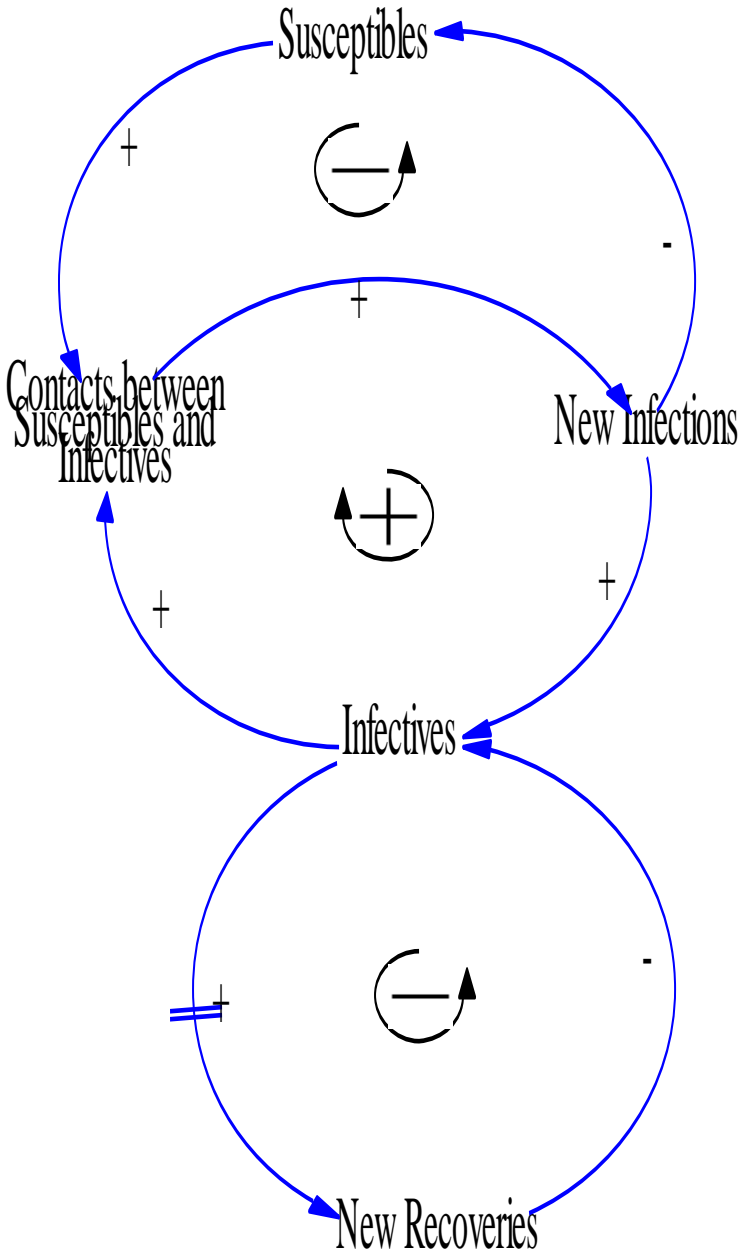
- Set
  - $c=10$  (people/month)
  - $\beta=0.04$  (4% chance of transmission per S-I contact)
  - $\mu=10$
  - Birth and death rate=0
  - Initial infectives=1, other 1000 susceptible

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Reasons

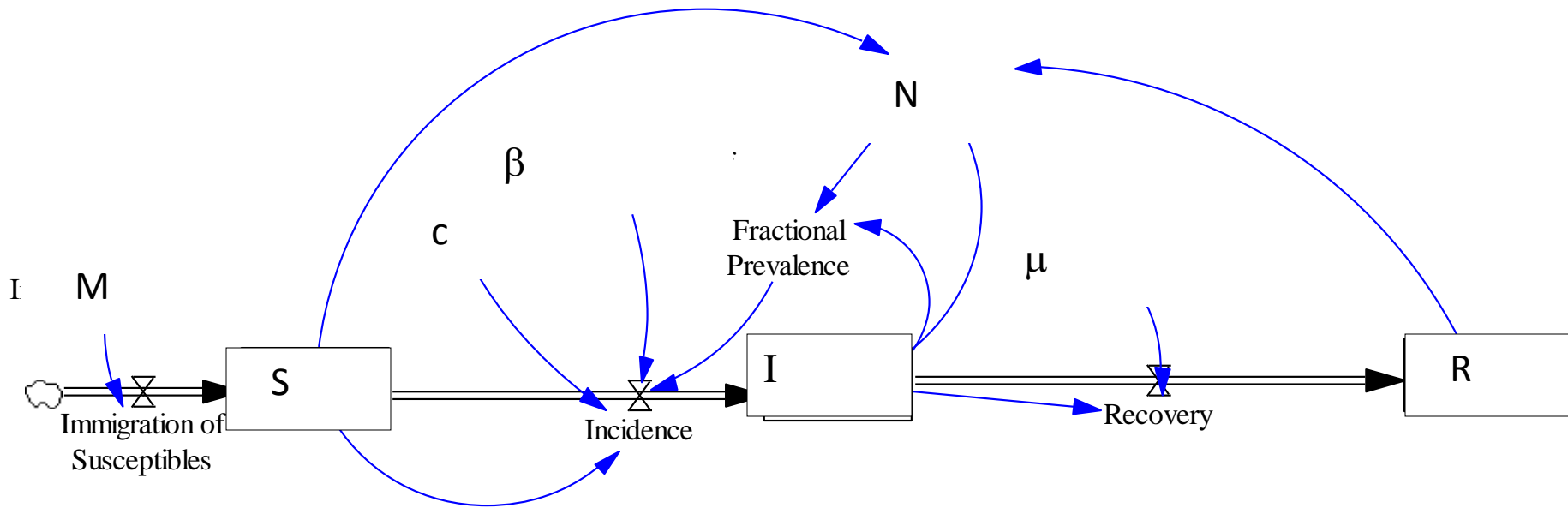
# Let's adjust several things

- Transmissibility (originally 0.04)
  - To 0.10
  - To 0.20
- Duration of infection (originally 10)
  - To 20
  - To 30

# Associated Feedbacks

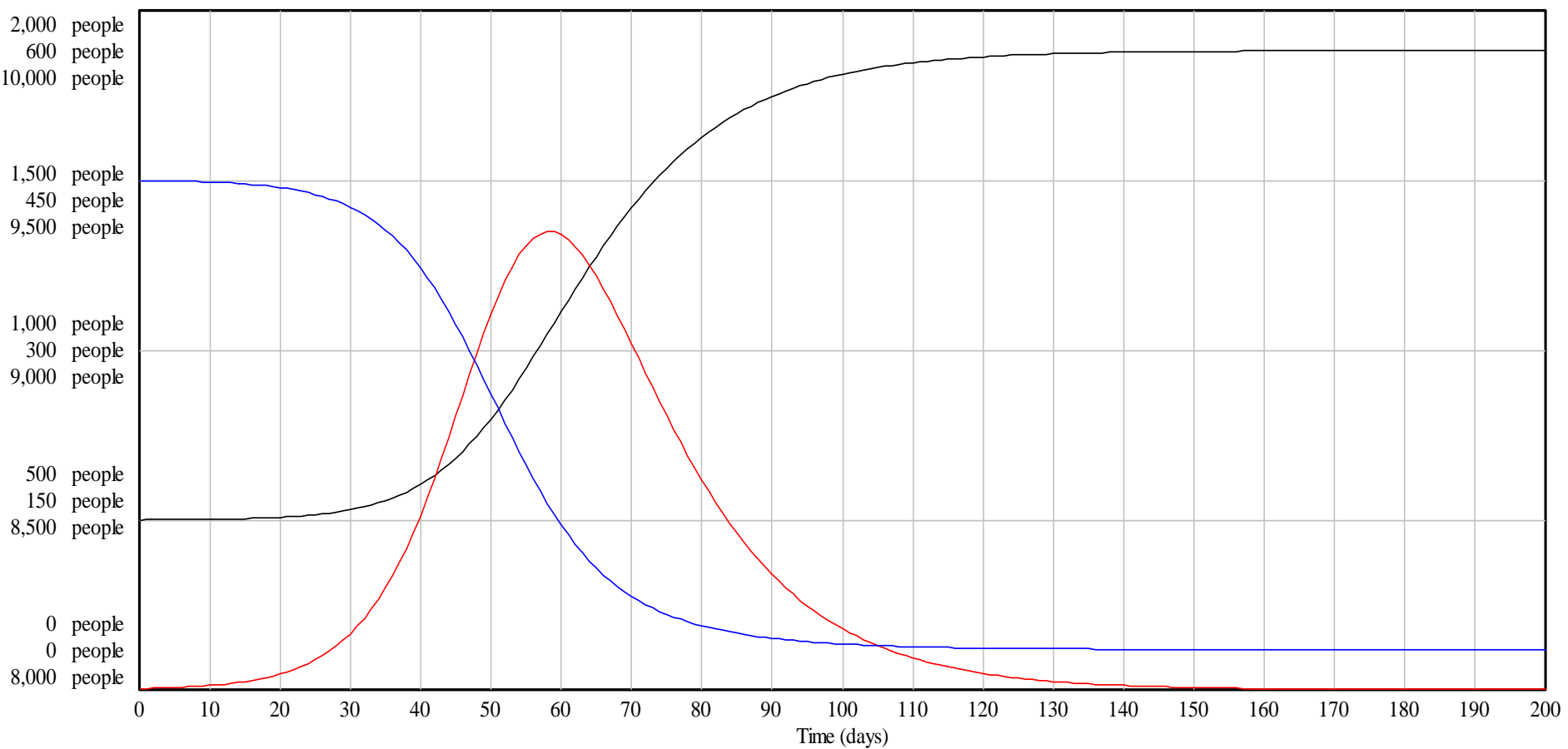


# Mathematical Notation



# Example Dynamics of SIR Model (No Births or Deaths)

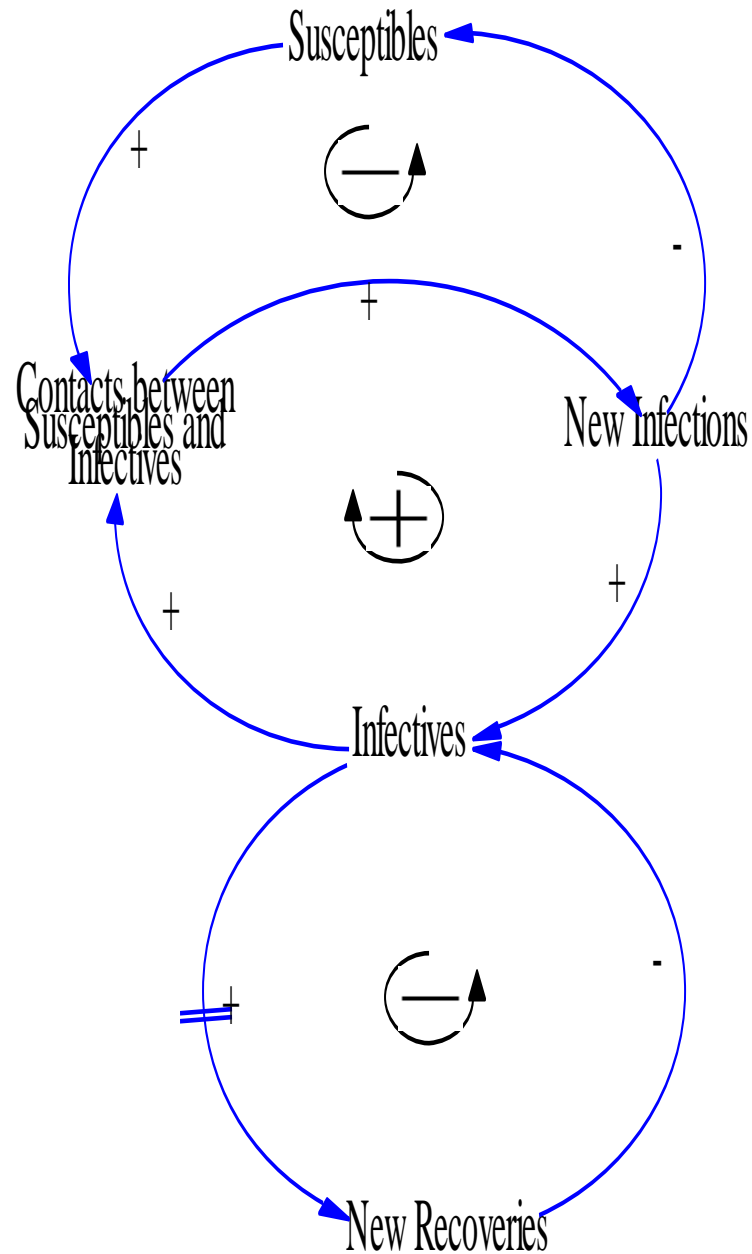
SIR Example



Susceptible Population S : SIR example ————— people  
 Infectious Population I : SIR example ————— people  
 Recovered Population R : SIR example ————— people



# Shifting Feedback Dominance



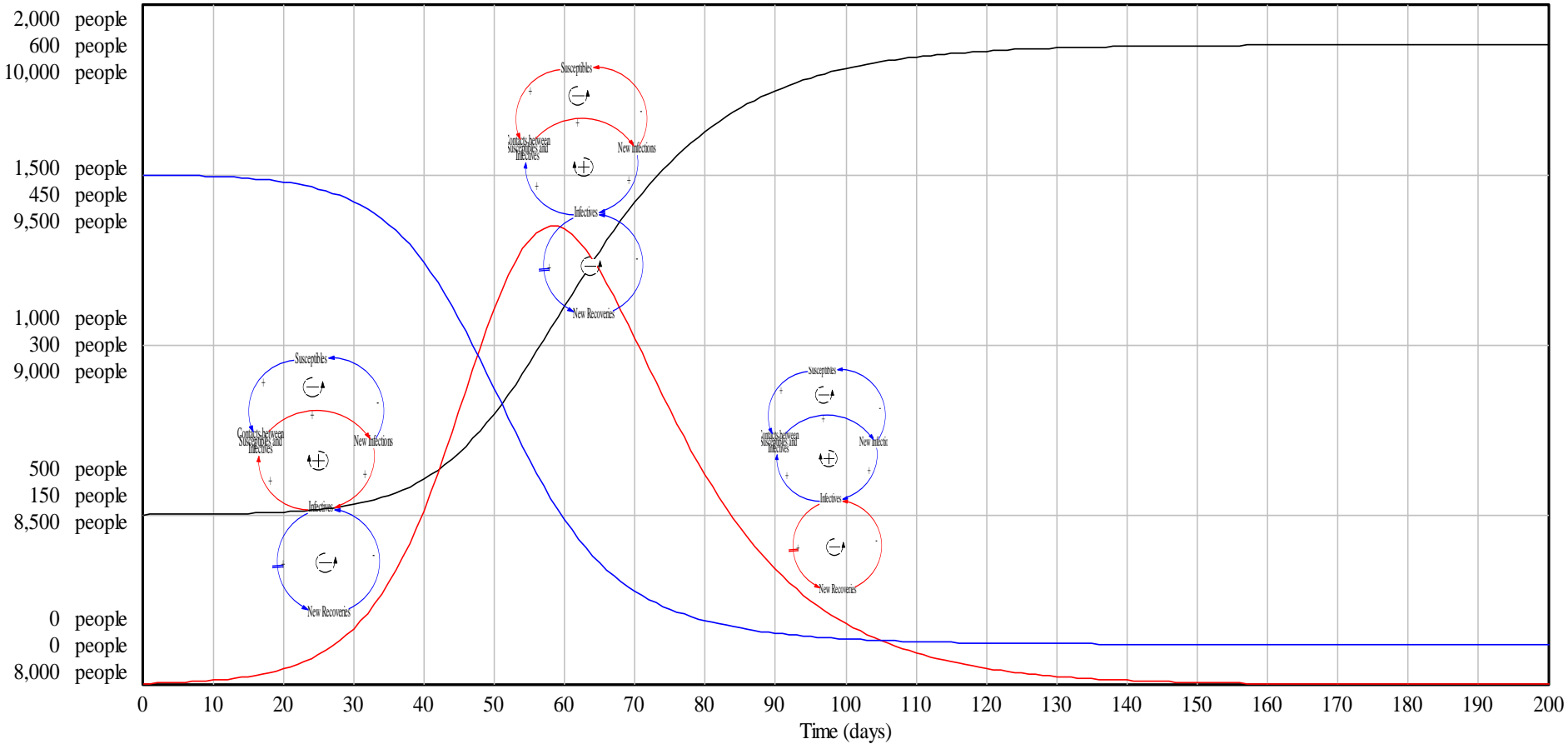
# Explaining the Stock & Flow Dynamics:

## Infectives & Susceptibles

- Initially
  - Each infective infects  $c(S/N)\beta \approx c\beta$  people on average for each time unit – the maximum possible rate
  - The rate of recoveries is 0
- In short term
  - # Infectives grows (quickly) => rate of infection rises quickly
    - (Positive feedback!)
  - Susceptibles starts to decline, but still high enough that each infective is surrounded overwhelmingly by susceptibles, so efficient at transmitting
- Over time, more infectives, and fewer Susceptibles
  - Fewer S around each I => Rate of infections per I declines
  - Many infectives start recovering => slower rise to I
- “Tipping point”: # of infectives plateaus
  - Rate of infections = Rate of recoveries
  - Each infective infects exactly one “replacement” before recovering
- In longer term, declining # of infectives & susceptibles => Lower & lower rate of new infections (negative feedback!)
- Change in I dominated by recoveries => goal seeking to 0 (negative feedback!)

# Shifting Feedback Dominance

SIR Example



Susceptible Population S : SIR example ————— people  
 Infectious Population I : SIR example ————— people  
 Recovered Population R : SIR example ————— people

# Damped Oscillatory Behavior

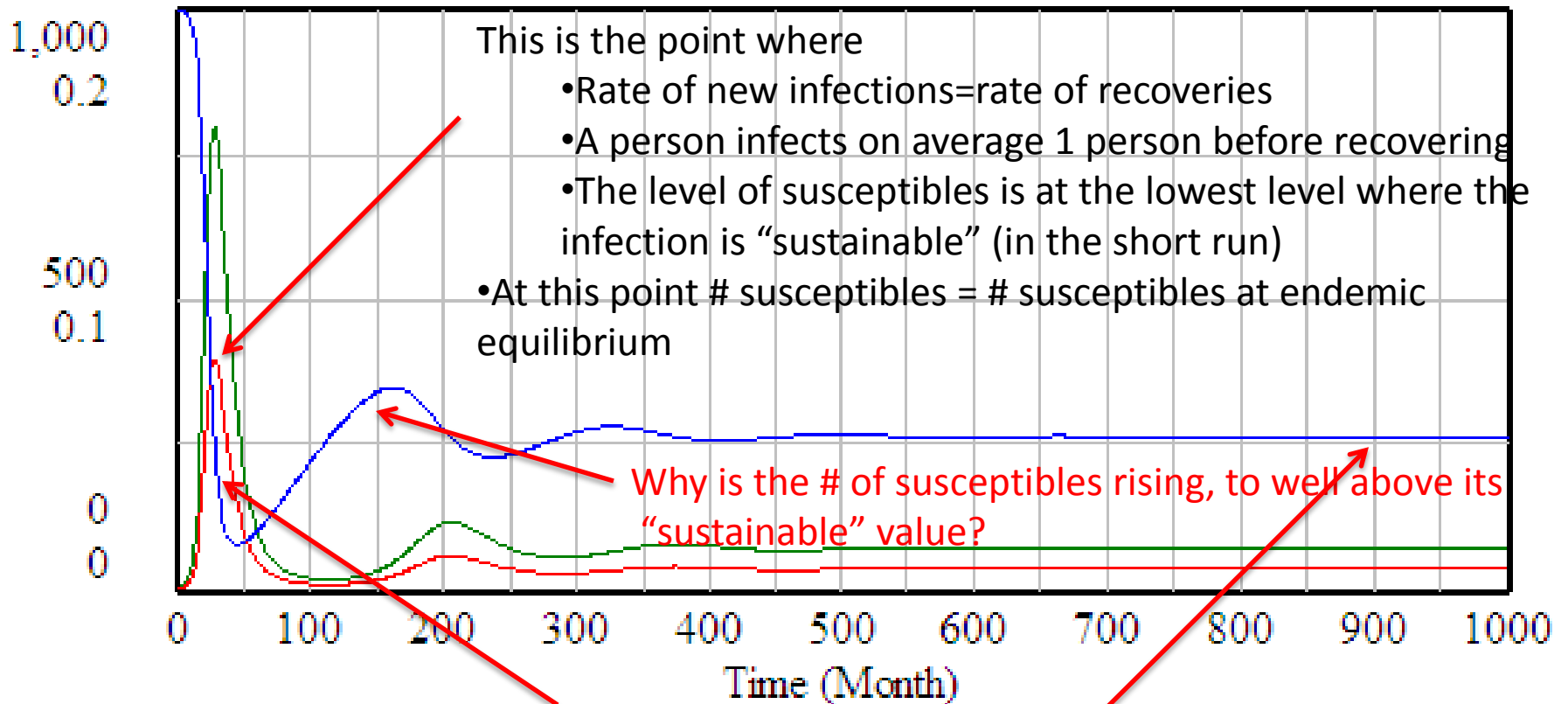
- Modify model to have births and deaths, with an annual birth-and-death rate
- Set Model/Settings/Final Time to 1000 (long time frame)
- In “Synthesim” (“Running man”) mode, set Birth/death rates
  - 0.02
  - 0.05
  - 0.07
  - 0.09

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# Delays

- For a while after infectives start declining (i.e. susceptibles are below sustainable endemic value), they still deplete susceptibles sufficiently for susceptibles to decline
- For a while after susceptibles are rising (until susceptibles=endemic value), infectives will still decline
- For a while after infectives start rising, births  $>$  # of infections  $\Rightarrow$  susceptibles will rise to a peak well above endemic level

# Susceptibles and Infectives

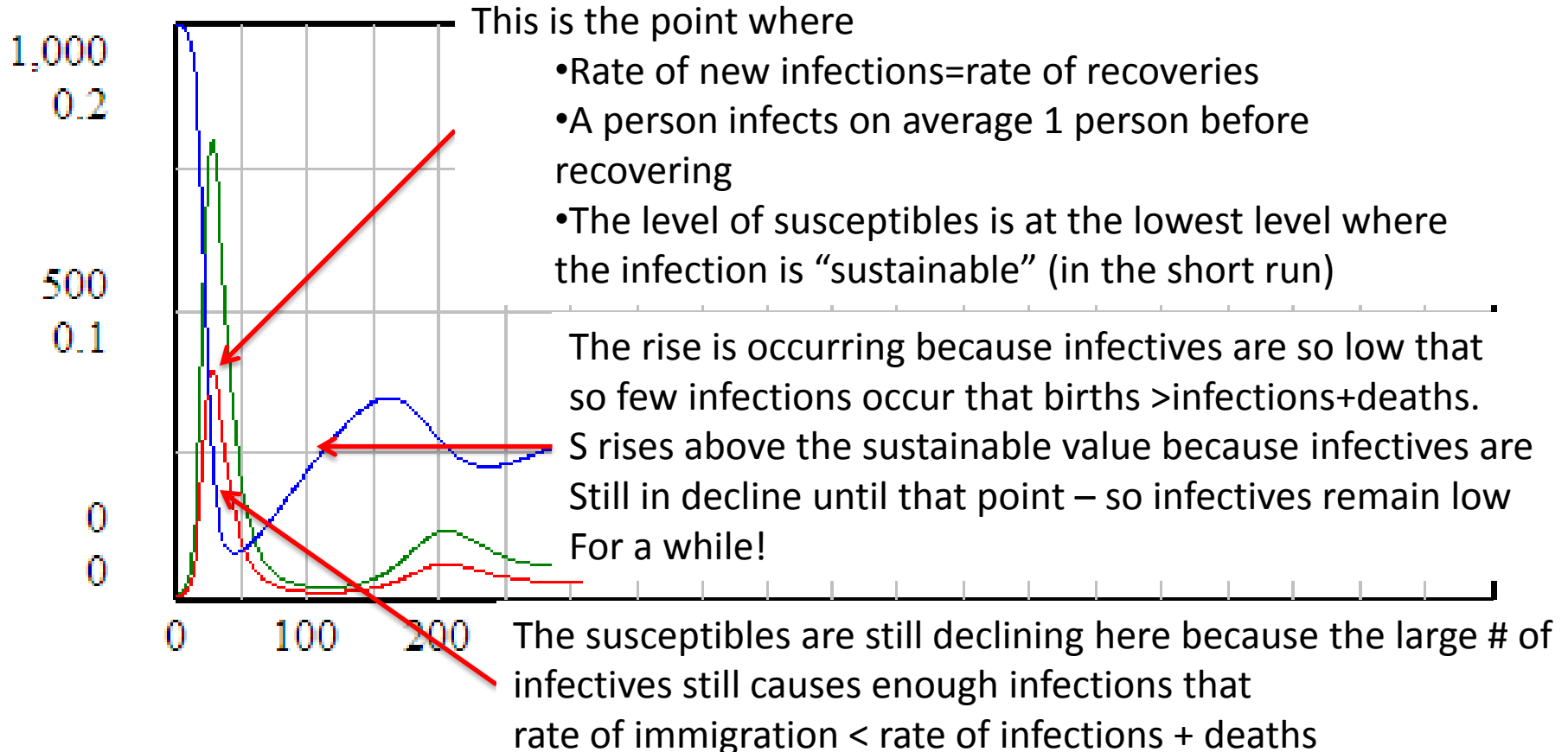


Why is the # of susceptibles still declining?

This fraction of susceptibles at endemic equilibrium is the minimum “sustainable” value of susceptible – i.e. the value where the properties above hold.

- Above this fraction of susceptibles, the # infected will rise
- Below this fraction of susceptibles, the # infected will fall

# Susceptibles and Infectives



Susceptible : Alternate SIR Birth Death \_\_\_\_\_

Infectives : Alternate SIR Birth Death \_\_\_\_\_

Force of Infection : Alternate SIR Birth Death \_\_\_\_\_



# Equilibrium Behaviour

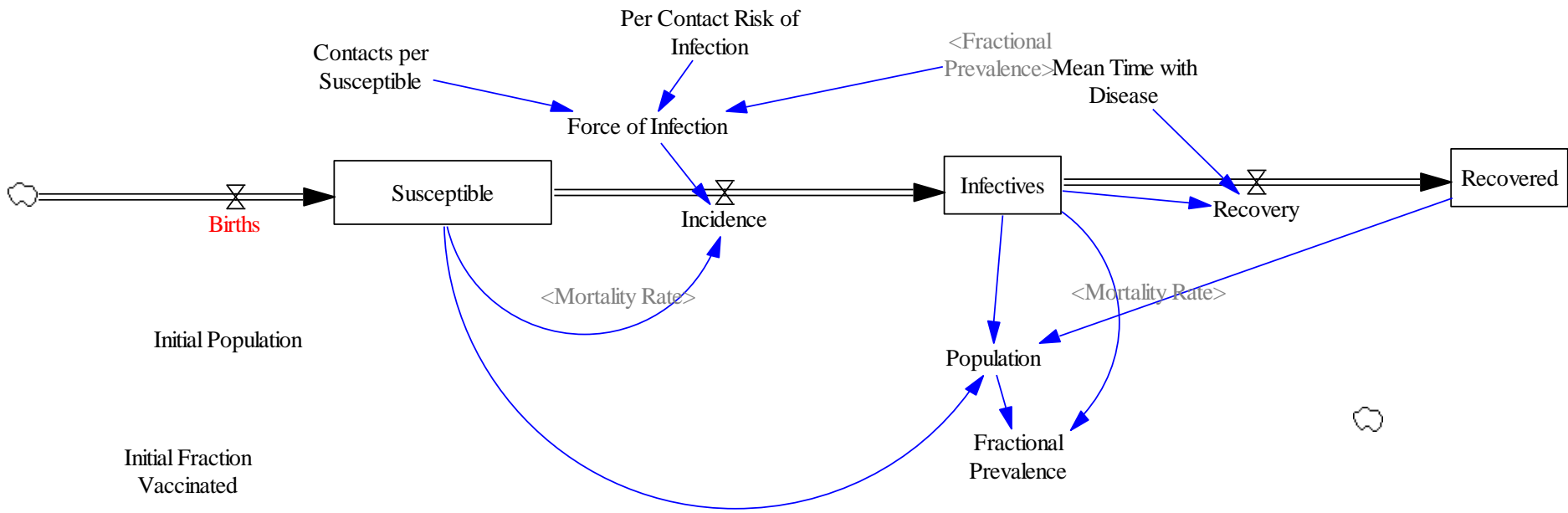
- With Births & Deaths, the system can approach an “endemic equilibrium” where the infection stays circulating in the population – but in balance
- The balance is such that (simultaneously)
  - The rate of new infections = The rate of immigration
    - Otherwise # of susceptibles would be changing!
  - The rate of new infections = the rate of recovery
    - Otherwise # of infectives would be changing!

# Tipping Point

- Now try setting transmission rate  $\beta$  to 0.005

# Recall: Kendrick-McKermack Model

- Partitioning the population into 3 broad categories:
  - Susceptible (S)
  - Infectious (I)
  - Removed (R)



# Shorthand for Key Quantities for Infectious Disease Models: Stocks

- $I$  (or  $Y$ ): Total number of infectives in population
  - This could be just one stock, or the sum of many stocks in the model (e.g. the sum of separate stocks for asymptomatic infectives and symptomatic infectives)
- $N$ : Total size of population
  - This will typically be the sum of all the stocks of people
- $S$  (or  $X$ ): Number of susceptible individuals

# Intuition Behind Common Terms

- $I/N$ : The Fraction of population members (or, by assumption, contacts!) that are infective
  - Important: Simplest models assume that this is also the fraction of a given susceptible's contacts that are infective! Many sophisticated models relax this assumption
- $c(I/N)$ : Number of *infectives* that come into contact with a susceptible in a given unit time
- $c(I/N)\beta$ : “Force of infection”: *Likelihood a given susceptible will be infected per unit time*
  - The idea is that if a given susceptible comes into contact with  $c(I/N)$  infectives per unit time, and if each such contact gives  $\beta$  likelihood of transmission of infection, then that susceptible has roughly a total likelihood of  $c(I/N)\beta$  of getting infected per unit time (e.g. month)

# Key Term: Flow Rate of New Infections

- This is the key form of the equation in many infectious disease models
- Total # of susceptiblesinfected per unit time
  - # of Susceptibles \* “Likelihood” a given susceptible will be infected per unit time =  $S * (\textit{Force of Infection})$   
 $= S(c(I/N)\beta)$
  - Note that this is a term that multiplies both S and I !
    - This is much different than the purely linear terms on which we have previously focused
  - “Likelihood” is actually a likelihood density (e.g. can be  $>1$  – indicating that mean time to infection is  $<1$ )

# Another Useful View of this Flow

- Recall: Total # of susceptibles infected per unit time = # of Susceptibles \* “Likelihood” a given susceptible will be infected per unit time =  $S * (\text{“Force of Infection”}) = S(c(I/N)\beta)$
- The above can also be phrased as the following:  $S(c(I/N)\beta) = I(c(S/N)\beta) = \# \text{ of Infectives} * \text{Average \# susceptibles infected per unit time by each infective}$
- This implies that as # of susceptibles falls  $\Rightarrow$  # of susceptibles surrounding each infective falls  $\Rightarrow$  the rate of new infections falls (“Less fuel for the fire” leads to a smaller burning rate)





# Underlying Equations

$$\dot{S} = M - c \left( \frac{I}{N} \right) \beta S$$

$$\dot{I} = c \left( \frac{I}{N} \right) \beta S - \frac{I}{\mu}$$

$$\dot{R} = \frac{I}{\mu}$$

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