# Mathematics of Infectious Diseases

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# Happy St. Patrick's Day!









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- 2012-present: Assistant Professor of Mathematics, UCF Mathematical biology, ecology and epidemiology





## Core Research

- New models, new approaches, new results, and new applications to complex biological systems
- Developed new graph-theoretic approaches to investigate dynamics of coupled systems on networks
- Established sharp threshold results for many heterogeneous infectious disease models
- Formulated and analyzed new mathematical models for waterborne diseases such as cholera
- Defined a new concept target reproduction number to mathematically measure intervention and control strategies in order to eradicate infectious diseases in heterogeneous host populations



Central Florida Math Circle

- Once you receive the paper, you have choices to
  - hold the paper, or
  - pass the paper to one person with whom you shake hands, or
  - rip it into two or more pieces and pass them to different persons with whom you shake hands
- Repeat whenever hearing "NEXT"



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- Now consider the paper that you have passed after handshaking (effective contact) is an infectious disease
- How many persons have been infected by the "paper" disease?



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- An immune response against a pathogen, which can include a high fever, inflammation, and other damaging symptoms, can be more devastating than the direct damage caused by the pathogen
- Infectious diseases require a mode of transmission (direct or indirect transmission, waterborne, airborne, vector-borne, food-borne, etc.) to be transmitted to other individuals (infectious)

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- spread among heterogeneous host groups (heterogeneous model)



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- MacDonald [1957]
  - included adult and larval mosquitoes
  - showed that control of adult mosquitoes is more effective than control of larvae



## Gonorrhea

A sexually transmitted bacterial disease



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A sexually transmitted bacterial disease

- Hethcote and Yorke [1984]
  - found that a core group of very highly active individuals maintains the disease at an epidemic level
  - contact tracing is more efficient for control than routine screening



## Measles

A viral childhood disease



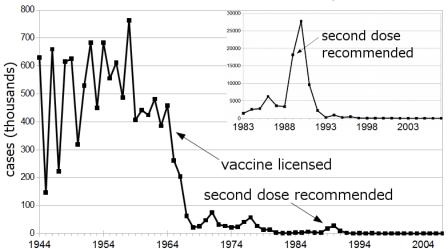
## Measles

#### A viral childhood disease

- Anderson and May [1992]
  Infectious Diseases of Humans: Dynamics and Control
  - guided design of vaccination programs (two dose strategy in UK)
  - two dose vaccination strategy started in US in 1989



# Measles cases in the United States, 1944-2007





# Why Should We Do Mathematical Modeling of Infectious Diseases?

- Mathematical models and computer simulations can be used as experimental tools for testing control measures and determining sensitivities to changes in parameter values e.g.,
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- The modeling process can contribute to the design of public health surveys especially by suggesting data that should be collected e.g.,
  - contact tracing for gonorrhea
- Mathematical modeling of epidemics can lead to and motivate new results in mathematics e.g.,
  - ruling out periodic orbits in higher dimensional ODE systems



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In 1902 Sir Ronald Ross was awarded the Noble prize for medicine for proving that malaria is transmitted by mosquitoes





# The Kermack-McKendrick Epidemic Model [1927]

Constant population divided into 3 compartments:

- S(t) = number of individuals *susceptible* to disease
- I(t) = number of individuals *infected* by disease
- R(t) = number of individuals recovered from disease

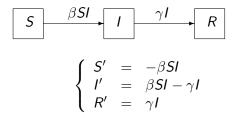
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- $\beta$  = effective contact coefficient
- $1/\gamma$  = average time of infection (depends on disease) Length of the infective period is exponentially distributed



# Dynamics of Kermack-McKendrick Model

- Asymptotic analysis for  $I' = \beta SI \gamma I$ 
  - If  $\beta S(0) < \gamma$  then I(t) decreases monotonically to 0
  - If  $\beta S(0) > \gamma$  then  $I(t) \to I_{max} \to 0$  as  $t \to \infty$  leaving a positive number of susceptibles  $S(\infty) > 0$

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$$I_{max} = I(0) + S(0) - \frac{\gamma}{\beta} \left( 1 + \ln \frac{\beta S(0)}{\gamma} \right)$$

- Basic reproduction number (threshold):  $\mathcal{R}_0 = \beta S(0) \frac{1}{\gamma}$  is the average number infected when one infective enters a susceptible population
  - $\mathcal{R}_0 < 1$ : disease dies out
  - $\mathcal{R}_0 > 1$ : there is a disease epidemic



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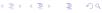
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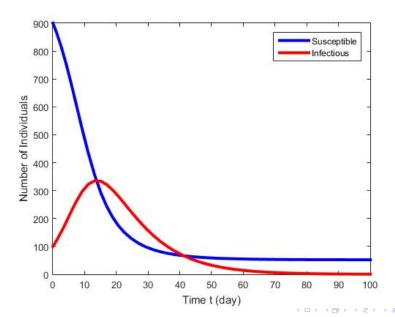
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- The final size relation

$$1 - R(\infty) = \exp(-\mathcal{R}_0 R(\infty))$$





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# Simulations with Other $\mathcal{R}_0$ Values

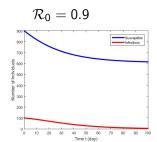
$$\mathcal{R}_0 = 0.9\,$$

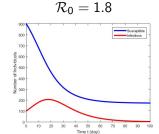
$$\mathcal{R}_0 = 1.8$$

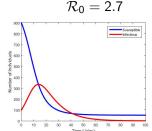
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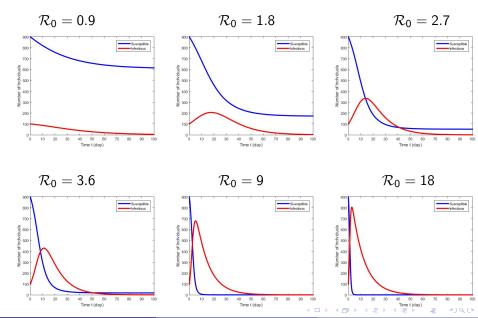
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  - For measles in urban areas  $\mathcal{R}_0 \approx 10\text{-}16$  need to vaccinate over 90%
  - ▶ For ebola  $\mathcal{R}_0 \approx 2$  need to vaccinate over only 50% (hypothetically)



## Thanks!



