# AMS/CIMPA School KENYA 2023

Mathematical Models of Chemical Reaction Networks;Basic tools, applications and open challenges.

University of Nairobi, Kenya

### Introduction to Epidemiological Models

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SI Model

SIS Model

The Basic Reproductive Number  $(R_0)$ 

SIR Model

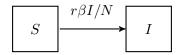
SEIR Model

- Population-based models
  - Can be deterministic or stochastic
  - Continuous time
    - Ordinary differential equations
    - Partial differential equations
    - Delay differential equations
    - Integro-differential equations
  - Discrete time
    - Difference equations
- Agent-based/individual-based models
  - Usually stochastic
  - Usually discrete time

# • Population-based models

- Can be deterministic or stochastic
- Continuous time
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Susceptible-Infectious Model: applicable to HIV.



$$\begin{split} \frac{\mathrm{d}S}{\mathrm{d}t} &= -r\beta S\frac{I}{N} \\ \frac{\mathrm{d}I}{\mathrm{d}t} &= r\beta S\frac{I}{N} \end{split}$$

- S: Susceptible humans
- I: Infectious humans
- r: Number of contacts per unit time
- $\beta$ : Probability of disease transmission per contact
- N: Total population size: N = S + I.

# Analyzing the SI Model

The system can be reduced to one dimension,

$$\frac{\mathrm{d}I}{\mathrm{d}t} = r\beta(N-I)\frac{I}{N},$$

with solution,

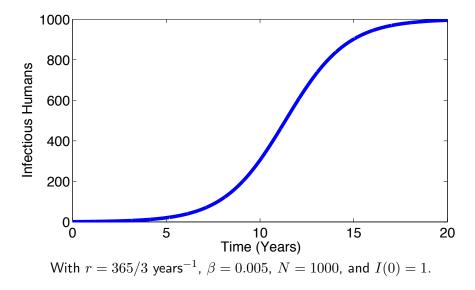
for  $I(0) = I_0$ . Equilibrium Points:

$$I(t) = \frac{I_0 N}{(N - I_0)e^{-r\beta t} + I_0},$$

$$I_{dfe} = 0$$
  
 $I_{ee} = N$ 

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### Numerical Solution of SI Model



- Note that in some models, usually of diseases where contacts are not well defined, rβ (the number of contacts per unit time multiplied by the probability of disease transmission per contact) are combined into one parameter (often also called β the number of adequate contacts per unit time).
- For diseases where a contact is well defined (such as sexually transmitted diseases like HIV or vector-borne diseases like malaria), it is usually more appropriate to separate the contact rate, r, and the probability of transmission per contact, β.
- For diseases where contacts are not well defined (such as air-borne diseases like influenza) it is usually more appropriate to combine the two into one parameter.

Outline

SI Model

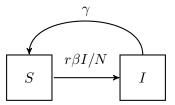
# SIS Model

The Basic Reproductive Number  $(R_0)$ 

SIR Model

SEIR Model

Susceptible-Infectious-Susceptible Model: applicable to the common cold.



$$\frac{\mathrm{d}S}{\mathrm{d}t} = -r\beta S \frac{I}{N} + \gamma I$$
$$\frac{\mathrm{d}I}{\mathrm{d}t} = r\beta S \frac{I}{N} - \gamma I$$

 $\gamma$ : Per-capita recovery rate

### Analyzing the SIS Model

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The system can be reduced to one dimension,

$$\frac{\mathrm{d}I}{\mathrm{d}t} = r\beta(N-I)\frac{I}{N} - \gamma I,$$

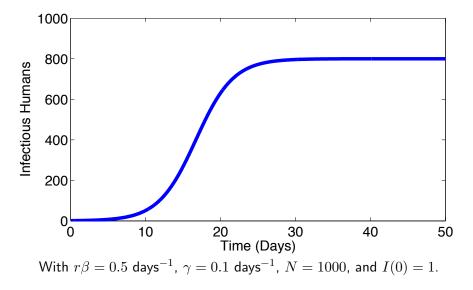
with solution,

$$I(t) = \frac{\frac{N}{r\beta} \cdot (r\beta - \gamma)}{1 + \left(\frac{N}{r\beta} \frac{(r\beta - \gamma)}{I_0} - 1\right) e^{-(r\beta - \gamma)t}},$$

for  $I(0) = I_0$ . Equilibrium Points:

$$\begin{split} I_{\rm dfe} &= 0\\ I_{\rm ee} &= \frac{(r\beta - \gamma)N}{r\beta} \end{split}$$

### Numerical Solution of SIS Model



SI Model

SIS Model

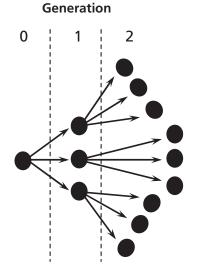
The Basic Reproductive Number  $(R_0)$ 

SIR Model

SEIR Model

# The Basic Reproductive Number $(R_0)$

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Initial phase of epidemic  $(R_0 = 3)$ 

Pan-InfORM (2009) J. Kagunda, UON, AMS Kenya 2023 ~~~18

- The basic reproductive number,  $R_0$ , is the number or secondary infections that one infected person would produce in a fully susceptible population through the entire duration of the infectious period.
- $R_0$  provides a threshold condition for the stability of the disease-free equilibrium point (for most models):
  - The disease-free equilibrium point is locally asymptotically stable when  $R_0 < 1$ : the disease dies out.
  - ► The disease-free equilibrium point is unstable when R<sub>0</sub> > 1: the disease establishes itself in the population or an epidemic occurs.
  - For a given model,  $R_0$  is fixed over all time.
- This definition is only valid for simple homogeneous autonomous models.
- Can define similar threshold conditions for more complicated models that include heterogeneity and/or seasonality but the basic definition no longer holds.

 $R_0$  can be expressed as a product of three quantities:

$$R_0 = \begin{pmatrix} \mathsf{Number of} \\ \mathsf{contacts} \\ \mathsf{per unit time} \end{pmatrix} \begin{pmatrix} \mathsf{Probability of} \\ \mathsf{transmission} \\ \mathsf{per contact} \end{pmatrix} \begin{pmatrix} \mathsf{Duration of} \\ \mathsf{infection} \end{pmatrix}$$

For SIS model:

$$R_0 = r \times \beta \times \frac{1}{\gamma}$$

# **Reproductive Numbers**

- The (*effective*) reproductive number,  $R_e$ , is the number of secondary infections that one infected person would produce through the entire duration of the infectious period.
- Typically, but not always,  $R_e$  is the product of  $R_0$  and the proportion of the population that is susceptible.
- $R_e$  describes whether the infectious population increases or not. It increases when  $R_e > 1$ ; decreases when  $R_e < 1$  and is constant when  $R_e = 1$ . When  $R_e = 1$ , the disease is at equilibrium.
- $R_e$  can change over time.
- The control reproductive number,  $R_c$ , is the number of secondary infections that one infected person would produce through the entire duration of the infectious period, in the presence of control interventions.

$$\begin{aligned} R_e(t) &= \begin{pmatrix} \mathsf{Number of} \\ \mathsf{contacts} \\ \mathsf{per unit time} \end{pmatrix} \begin{pmatrix} \mathsf{Probability of} \\ \mathsf{transmission} \\ \mathsf{per contact} \end{pmatrix} \begin{pmatrix} \mathsf{Duration of} \\ \mathsf{infection} \end{pmatrix} \\ &\times \begin{pmatrix} \mathsf{Proportion of} \\ \mathsf{susceptible} \\ \mathsf{population} \end{pmatrix} \end{aligned}$$

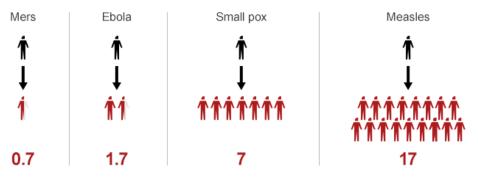
For SIS model:

$$R_e(t) = R_0 \times \frac{S(t)}{N(t)}$$
$$= \frac{r\beta S(t)}{\gamma N(t)}.$$

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# How quickly does it spread?

Basic reproduction value



Source: ECDC, UMICH, Lancet

http://www.cameroonweb.com/CameroonHomePage/NewsArchive/Ebola-How-does-it-compare-316932

SI Model

SIS Model

The Basic Reproductive Number  $(R_0)$ 

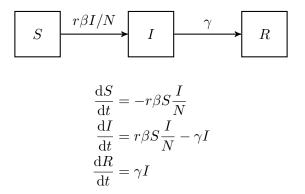
SIR Model

SEIR Model

### SIR Model

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Susceptible-Infectious-Recovered Model: applicable to measles, mumps, rubella.



R: Recovered humans with N = S + I + R.

- Can reduce to two dimensions by ignoring the equation for R and using R = N S I.
- Can no longer analytically solve these equations.
- Infinite number of equilibrium points with  $I^* = 0$ .
- Perform phase portrait analysis.
- Estimate final epidemic size.

$$R_0 = \begin{pmatrix} \mathsf{Number of} \\ \mathsf{contacts} \\ \mathsf{per unit time} \end{pmatrix} \begin{pmatrix} \mathsf{Probability of} \\ \mathsf{transmission} \\ \mathsf{per contact} \end{pmatrix} \begin{pmatrix} \mathsf{Duration of} \\ \mathsf{infection} \end{pmatrix}$$

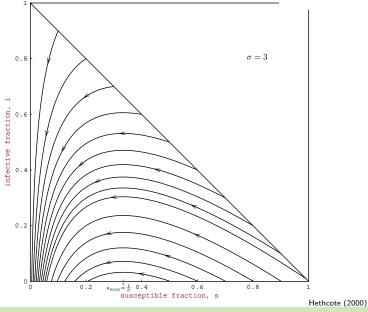
$$R_0 = r \times \beta \times \frac{1}{\gamma}$$
$$= \frac{r\beta}{\gamma}$$

- If  $R_0 < 1$ , introduced cases do not lead to an epidemic (the number of infectious individuals decreases towards 0).
- If R<sub>0</sub> > 1, introduced cases can lead to an epidemic (temporary increase in the number of infectious individuals).

$$R_e(t) = \frac{r\beta}{\gamma} \frac{S(t)}{N}$$

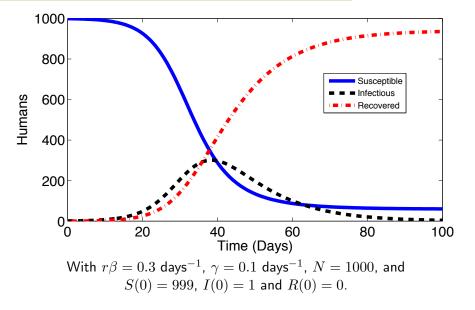
### Phase Portrait of SIR Model

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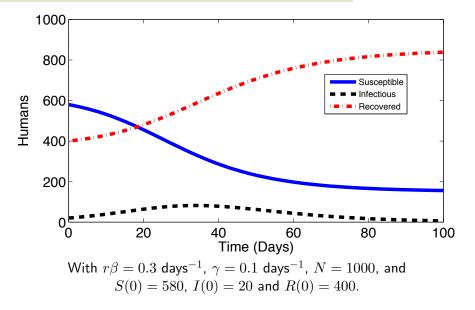
### Numerical Solution of SIR Model

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Numerical Solution of SIR Model

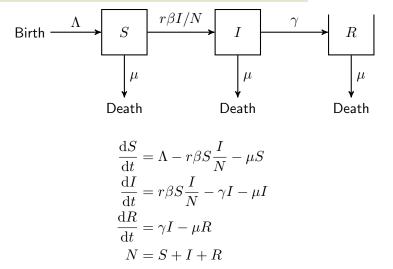
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- Need to include human demographics for diseases where the time frame of the disease dynamics is comparable to that of human demographics.
- There are many different ways of modeling human demographics
  - Constant immigration rate
  - Constant per-capita birth and death rates
  - Density-dependent death rate
  - Disease-induced death rate.

# Endemic SIR Model

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- $\Lambda$ : Constant recruitment rate
- $\mu$ : Per-capita removal rate

- Can no longer reduce the dimension or solve analytically.
- There are two equilibrium points: disease-free and endemic

$$\begin{split} S_{\rm dfe} &= \frac{\Lambda}{\mu} & \qquad S_{\rm ee} &= \frac{\Lambda(\gamma + \mu)}{r\beta\mu} \\ I_{\rm dfe} &= 0 & \qquad I_{\rm ee} &= \frac{\Lambda(r\beta - (\gamma + \mu))}{r\beta(\gamma + \mu)} \\ R_{\rm dfe} &= 0 & \qquad R_{\rm ee} &= \frac{\gamma\Lambda(r\beta - (\gamma + \mu))}{r\beta\mu(\gamma + \mu)} \end{split}$$

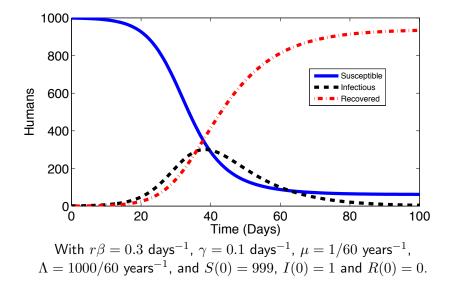
• Can perform stability analysis of these equilibrium points and draw phase portraits.

$$R_0 = \begin{pmatrix} \mathsf{Number of} \\ \mathsf{contacts} \\ \mathsf{per unit time} \end{pmatrix} \begin{pmatrix} \mathsf{Probability of} \\ \mathsf{transmission} \\ \mathsf{per contact} \end{pmatrix} \begin{pmatrix} \mathsf{Duration of} \\ \mathsf{infection} \end{pmatrix}$$

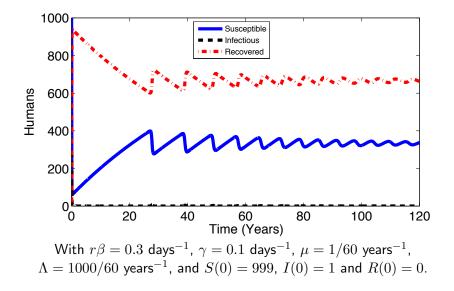
$$R_0 = r \times \beta \times \frac{1}{\gamma + \mu}$$
$$= \frac{r\beta}{\gamma + \mu}$$

- If  $R_0 < 1$ , the disease-free equilibrium point is globally asymptotically stable and there is no endemic equilibrium point (the disease dies out).
- If R<sub>0</sub> > 1, the disease-free equilibrium point is unstable and a globally asymptotically stable endemic equilibrium point exists.

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SI Model

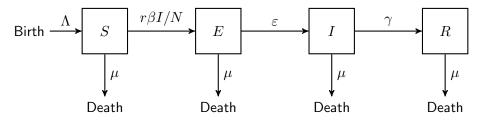
SIS Model

The Basic Reproductive Number  $(R_0)$ 

SIR Model

SEIR Model

Susceptible-Exposed-Infectious-Recovered Model: applicable to measles, mumps, rubella.



- *E*: Exposed (latent) humans
  - $\varepsilon$ : Per-capita rate of progression to infectious state

# SEIR Model



$$\begin{aligned} \frac{\mathrm{d}S}{\mathrm{d}t} &= \Lambda - r\beta S \frac{I}{N} - \mu S \\ \frac{\mathrm{d}E}{\mathrm{d}t} &= r\beta S \frac{I}{N} - \varepsilon E \\ \frac{\mathrm{d}I}{\mathrm{d}t} &= \varepsilon E - \gamma I - \mu I \\ \frac{\mathrm{d}R}{\mathrm{d}t} &= \gamma I - \mu R \end{aligned}$$

with

$$N = S + E + I + R.$$

# $R_0$ for the Endemic SEIR Model

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$$\begin{split} R_0 = \begin{pmatrix} \mathsf{Number of} \\ \mathsf{contacts} \\ \mathsf{per unit time} \end{pmatrix} \begin{pmatrix} \mathsf{Probability of} \\ \mathsf{transmission} \\ \mathsf{per contact} \end{pmatrix} \begin{pmatrix} \mathsf{Duration of} \\ \mathsf{infection} \end{pmatrix} \\ \times \begin{pmatrix} \mathsf{Probability of} \\ \mathsf{surviving} \\ \mathsf{exposed stage} \end{pmatrix} \end{split}$$

$$R_0 = r \times \beta \times \frac{1}{\gamma + \mu} \times \frac{\varepsilon}{\varepsilon + \mu}$$
$$= \frac{r\beta\varepsilon}{(\gamma + \mu)(\varepsilon + \mu)}$$

- If  $R_0 < 1$ , the disease-free equilibrium point is globally asymptotically stable and there is no endemic equilibrium point (the disease dies out).
- If R<sub>0</sub> > 1, the disease-free equilibrium point is unstable and a globally asymptotically stable endemic equilibrium point exists.

- Basic compartmental models assume a homogeneous population.
- Divide the population into different groups based on infection status:
  - M: Humans with maternal immunity
    - S: Susceptible humans
    - E: Exposed (infected but not yet infectious) humans
    - I: Infectious humans
    - *R*: Recovered humans.
- Can include time-dependent parameters to include the effects of seasonality.
- Can include additional compartments to model vaccinated and asymptomatic individuals, and different stages of disease progression.
- Can include multiple groups to model heterogeneity, age, spatial structure or host species.

O. DIEKMANN, H. HEESTERBEEK, AND T. BRITTON, *Mathematical Tools for Understanding Infectious Disease Dynamics.* 

Princeton Series in Theoretical and Computational Biology. Princeton University Press, Princeton, (2013).

H. W. HETHCOTE, "The mathematics of infectious diseases", *SIAM Review* **42**, 599–653 (2000).

 ${\rm M.}$  J. KEELING AND P. ROHANI, Modeling Infectious Diseases in Humans and Animals.

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