Deterministic Models for the Spread of Cholera

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- An infection of the small intestine caused by the bacterium Vibrio cholerae
- Infection causes mild diarrhea in most cases
- But some cases develop severe diarrhea and vomiting which if untreated may lead to death within a few hours due to dehydration and electrolyte imbalance



How is Cholera Transmitted?

- Human-environment (water)-human transmission Indirect, Slow
- Other transmission pathways
 Direct, Fast
- Incubation period is short, usually 2-3 days

Methods of control: purify water, provide sanitation, vaccination (oral vaccine is 50-60% effective)

Treatment: rehydration and antimicrobial





John Snow traced the source of a cholera outbreak in England in 1854 to a contaminated water pump





John Snow's memorial and pub



Where does Cholera Occur? http:healthmap.org 08.05.2013





 1817-1824: First recorded pandemic: started in the Bay of Bengal and spread to most of the world

- Zimbabwe outbreak in 2008-2009 (with 98,585 reported cases and 4,287 deaths, Mukandavire et al. PNAS 2011)
- Haiti outbreak since October 2010, with 712,330 reported cases and 8,655 deaths as of 24 November 2014, giving 1.2% fatality ratio (*ProMed data*)
- WHO estimates 3m-5m cholera cases per year with 100,000 deaths spread over 40-50 countries, mainly in underdeveloped countries,

thus modeling of cholera was largely neglected at least until the

Haiti outbreak

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Map of Haiti





Map of the Artibonite River



Possible source of outbreak: The United Nations Stabilization Mission in Haiti (MINUSTAH)

Piarroux et al. Emerging Infectious Diseases 2011



How to Model the Spread of Cholera?

"All models are wrong, but some are useful" George E.P. Box

"A scientific theory should be as simple as possible, but no simpler" Albert Einstein

Some ODE deterministic models assuming spatial homogeneity



Basic Model: Indirect Transmission, Saturating Incidence





Calculation of \mathcal{R}_0 for Codeço's model

There is a disease free equilibrium (DFE) at S = A/d, I = R = B = 0

$$\frac{dI}{dt} = \lambda S \frac{B}{K+B} - (d+\gamma+\alpha)I$$
$$\frac{dB}{dt} = \xi I - \delta B$$

By the next generation matrix method:

$$F = \begin{bmatrix} 0 & \frac{A\lambda}{dK} \\ 0 & 0 \end{bmatrix} \quad V = \begin{bmatrix} d + \gamma + \alpha & 0 \\ -\xi & \delta \end{bmatrix}$$

giving

$$\mathcal{R}_{0} =
ho(FV^{-1}) = rac{A\lambda\xi}{dK\delta(d+\gamma+lpha)}$$

The DFE is locally asymptotically stable if $\mathcal{R}_0 < 1$, whereas it is unstable if $\mathcal{R}_0 > 1$ [vdD & Watmough 2002]



Remark

If shedding is though of as a new infection, then ξ is placed in the ${\it F}$ matrix giving

$$F = \begin{bmatrix} 0 & \frac{A\lambda}{dK} \\ \xi & 0 \end{bmatrix} \quad V = \begin{bmatrix} d + \gamma + \alpha & 0 \\ 0 & \delta \end{bmatrix}$$

and

$$\mathcal{R}_0 = \sqrt{rac{A\lambda\xi}{dK\delta(d+\gamma+lpha)}}$$

agreeing with the previous value (no square root) at the threshold value $1 \$



Hyperinfectivity Model

Infectivity of freshly shed vibrios can be up to 700-fold greater than not freshly shed vibrios [Sengupta et al. *FEMS Microbiol Lett* 1998]



Hartley et al. PLOS Medicine 2006



Direct/Indirect Transmission Model: Mass Action Incidence

In certain situation, fast transmission may contribute 41-95% of all transmission [Mukandavire et al. *PNAS* 2011]

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Fast transmission: \beta SI (mass action)
Slow transmission: \lambda SB (mass action)
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Hyperinfectivity ignored





Tien & Earn Bull Math Biol 2010



Direct/Indirect Transmission Model: Mass Action Incidence Stage-Progression Model

The infectious individuals are categorized into n different compartments according to infectivity and infection age



Tien & Earn Bull Math Biol 2010



More General Cholera Model

[Shuai & vdD Math Biosc (2011) 234:118-126]

- Infectious individuals have multiple infectious stages I_1, I_2, \ldots, I_n
- ▶ Pathogen has multiple infectious states B_1, B_2, \ldots, B_m
- ► Fast transmission (contact between S and I_i) and Slow transmission (contact between S and B_k)
- Nonlinear incidence functions f_i(S, I_i) and g_k(S, B_k) under some biologically reasonable assumptions that include mass action incidence, saturating incidence, frequency dependent incidence for one infectious stage
- Nonlinear shedding functions h_i(l_i) including linear
- For different n, m, f_i, g_k, h_i our model includes those described on previous slides and by Tian & Wang Math Biosc 2011



Flow Diagram





Model Formulation

A cholera model is formulated as the following n + m + 1 ODEs:

$$\frac{dS}{dt} = A - \sum_{j=1}^{n} f_j(S, l_j) - \sum_{j=1}^{m} g_j(S, B_j) - dS$$

$$\frac{dI_1}{dt} = \sum_{j=1}^{n} f_j(S, l_j) + \sum_{j=1}^{m} g_j(S, B_j) - (d + \gamma_1 + \alpha_1) l_1$$

$$\frac{dI_i}{dt} = \gamma_{i-1} l_{i-1} - (d + \gamma_i + \alpha_i) l_i \quad \text{for } i = 2, \dots, n$$

$$\frac{dB_1}{dt} = \sum_{j=1}^{n} h_j(l_j) - \delta_1 B_1$$

$$\frac{dB_k}{dt} = \delta_{k-1} B_{k-1} - \delta_k B_k \quad \text{for } k = 2, \dots, m$$

with nonnegative initial conditions $S(0), I_i(0), B_k(0) \ge 0$



Feasible Region and Equilibria

Take biologically reasonable assumptions on f, g, h including sufficient smoothness, increasing in S, I_i, B_i and concave down in I_i, B_i

There is a positively invariant feasible region

$$\Gamma = \left\{ (S, I_1, \dots, I_n, B_1, \dots, B_m) \in \mathbb{R}^{n+m+1}_+ \mid S + I_1 + \dots + I_n \leq \frac{A}{d} \\ B_k \leq M_k, k = 1, \dots, m \right\}$$

- ► The disease-free equilibrium (DFE) denoted by $P_0 = (S_0, 0, ..., 0)$ with $S_0 = \frac{A}{d}$ lies on $\partial \Gamma$
- An endemic equilibrium (EE) in $\overset{\circ}{\Gamma}$, if it exists, is denoted by $P^* = (S^*, l_1^*, \dots, l_n^*, B_1^*, \dots, B_m^*)$



Basic Reproduction Number

The disease free equilibrium (DFE) is

$$S_0=\frac{A}{d}, I_i=B_k=0$$

Assume $p_i = \frac{\partial f_i}{\partial l_i}(S_0, 0)$, $q_k = \frac{\partial g_k}{\partial B_k}(S_0, 0)$, and $r_i = h'_i(0)$ are nonnegative and set $\nu_i = d + \gamma_i + \alpha_i$

Using the next generation matrix method, the basic reproduction number is derived as

$$\mathcal{R}_{0} = \frac{p_{1}}{\nu_{1}} + \dots + \frac{p_{n}\gamma_{1}\cdots\gamma_{n-1}}{\nu_{1}\cdots\nu_{n}} + \left(\frac{q_{1}}{\delta_{1}} + \dots + \frac{q_{m}}{\delta_{m}}\right) \left(\frac{r_{1}}{\nu_{1}} + \dots + \frac{r_{n}\gamma_{1}\cdots\gamma_{n-1}}{\nu_{1}\cdots\nu_{n}}\right)$$

direct transmission

indirect transmission

The DFE is locally asymptotically stable if $\mathcal{R}_0 < 1$, whereas it is unstable if $\mathcal{R}_0 > 1$ [vdD & Watmough 2002]



Global Stability of the DFE and Persistence

Theorem

- (a) If $\mathcal{R}_0 < 1$ then the DFE is globally asymptotically stable in the feasible region Γ
- (b) If $\mathcal{R}_0 > 1$ then the system is uniformly persistent and there exists at least one endemic equilibrium

Let $\omega^T \ge 0$ be a left eigenvector of $V^{-1}F$ corresponding to $\rho(V^{-1}F) = \mathcal{R}_0$ Then $Q = \omega^T V^{-1}x$ where x is the vector of infected compartments is a Lyapunov function for the system when $\mathcal{R}_0 < 1$

(no guessing in finding Q !)

This result holds for mass action, saturating and frequency dependent incidence functions f_i, g_k



Global Stability of the Endemic Equilibrium S^* , I_i^* , B_k^*

Theorem

If $\mathcal{R}_0 > 1$, then the system has a unique endemic equilibrium that is globally asymptotically stable in the interior of the feasible region Proof uses the Lyapunov function

$$V = c_1 \int_{S^*}^{S} \frac{\Phi(\xi) - \Phi(S^*)}{\Phi(\xi)} d\xi + \sum_{i=1}^{n} c_i \Big(I_i - I_i^* - I_i^* \ln \frac{I_i}{I_i^*} \Big) \\ + \sum_{k=1}^{m} c_{n+k} \Big(B_k - B_k^* - B_k^* \ln \frac{B_k}{B_k^*} \Big)$$

where $\Phi:(0,\mathit{S}_0]\to\mathrm{R}_+$ is increasing and satisfies certain inequalities

Constants $c_i > 0$ are chosen by using Kirchhoff's Matrix Tree theorem [1847] and the tree cycle identity [Li & Shuai JDE 2010] giving

$$\frac{dV}{dt} \leq 0 \quad \forall \quad (S, I_1, \cdots I_n, B_1, \cdots, B_m) \in \Gamma$$



Conclusions for this Cholera Model

For the ODE cholera model \mathcal{R}_0 is a sharp threshold

$$\mathcal{R}_{0} = \frac{p_{1}}{\nu_{1}} + \dots + \frac{p_{n}\gamma_{1}\cdots\gamma_{n-1}}{\nu_{1}\cdots\nu_{n}} + \left(\frac{q_{1}}{\delta_{1}} + \dots + \frac{q_{m}}{\delta_{m}}\right) \left(\frac{r_{1}}{\nu_{1}} + \dots + \frac{r_{n}\gamma_{1}\cdots\gamma_{n-1}}{\nu_{1}\cdots\nu_{n}}\right)$$

- \blacktriangleright If $\mathcal{R}_0 < 1$ then the disease dies out
- \blacktriangleright If $\mathcal{R}_0 > 1$ the disease persists at an endemic level

To control cholera: reduce \mathcal{R}_0 below 1

- ► vaccination (decrease p_i, q_k) herd immunity vaccinate more than 1 - 1/R₀ of the host population
- provide clean water (decrease q_k)
- improve sanitation (decrease r_i and/or increase δ_i)

Target/type reproduction numbers can be used to assess different strategies, Shuai et al. *J. Math Biol.* 2013, but accurate data is needed to evaluate control strategies



Final Size of an Epidemic

Assume

- ▶ Recruitment and death are zero (A = d = α_i = 0) so wlog N = 1
- Mass action direct transmission $(f_i(S, I_i) = \beta_i S I_i)$
- Linear pathogen shedding rate $(h_i(I_i) = \xi_i I_i)$
- ► Indirect transmission is either mass action or saturating incidence $(g_k(S, B_k) = \lambda_k SB_k \text{ or } \lambda_k S\frac{B_k}{K_k + B_k})$

Lemma

Under the above assumptions $I_i(t), B_k(t) \to 0$ as $t \to \infty$ (disease burns out)



Mass Action $g_k(S, B_k) = \lambda_k SB_k$

If a small number of infectious individuals or pathogens is initially introduced then the final size of an epidemic is

$$1-R(\infty)=\exp(-\mathcal{R}_0R(\infty))$$

where

$$\mathcal{R}_{0} = \frac{\beta_{1}}{\gamma_{1}} + \frac{\beta_{2}}{\gamma_{2}} + \dots + \frac{\beta_{n}}{\gamma_{n}} + \left(\frac{\lambda_{1}}{\delta_{1}} + \dots + \frac{\lambda_{m}}{\delta_{m}}\right) \left(\frac{\xi_{1}}{\gamma_{1}} + \dots + \frac{\xi_{n}}{\gamma_{n}}\right)$$

This is equivalent to

$$\ln \frac{S_0}{S(\infty)} = \mathcal{R}_0[1 - \frac{S(\infty)}{N}]$$

with $S_0\simeq N=1, R(\infty)=1-S(\infty)$



Saturating Incidence $g_k(S, B_k) = \lambda_k S \frac{B_k}{K_k + B_k}$

An attempt to model in a continuous way the fact that a relatively high level of pathogen is needed to develop cholera

The final size of an epidemic satisfies the inequality

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

where \mathcal{R}_0 is as before but with δ_k replaced with $\delta_k K_k$ giving an upper bound on $R(\infty)$



Simulations

For simulations consider saturating incidence with n = m = 2Infectivity of hyperinfectious state B_1 is 700 times that of B_2 Parameters are taken from cholera data in Hartley et al. *PLOS Medicine* 2006 giving $\mathcal{R}_0 = 15.83$

Simulations show

- Direct transmission provides a fast route for cholera spread and produces a higher peak in prevalence
- Heterogeneity in infectious hosts may decelerate disease spread and reduce the peak





Outbreak of a cholera epidemic keeping $\mathcal{R}_0=15.83$ and same initial conditions



More realistic cholera models should include spatial heterogeneity, age structure, optimal disease control, \cdots

Now consider spatial heterogeneity



Cholera Outbreak in Haiti as of November 2010



Piarroux et al. Emerging Infectious Diseases 2011



How to Include Spatial Heterogeneity in a Cholera Model?

A multigroup model with cross infection between groups

Tuite et al. Ann. Internal Medicine 2011 incorporate 10 groups corresponding to the populations in the 10 departments of Haiti and numerically simulate control strategies (vaccination and clean water)

Theoretical work with Zhisheng Shuai, SIAP 2013

- A partial differential equation (PDE) model for spread by humans and/or by a river Bertuzzo et al. J. Royal Society Interface 2010
- A discrete patch (metapopulation) ODE model for spread due to human and/or water movement

Work with Maria Eisenberg (U Michigan), Zhisheng Shuai (U Central Florida), and Joe Tien (Ohio State U) Math. Bios. 2013, J. Math Biol. 2014



Challenges in Modeling Cholera and Other Neglected Infectious Diseases

- How much epidemiological realism to include in a model
- How to obtain accurate data for model parameters
- How to include social aspects, such as lack of public information, lack of access to heath clinics
- How to include economic aspects, such as optimizing treatment strategies

▶

Answers depend in part on the particular question being addressed

THANK YOU

