

SOME STOCHASTIC MODELS IN EPIDEMICS

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INTRODUCTION

- stochastic models
 - the variables considered in the model are random and with a probability distribution
 - the distribution and model depend on some parameters
- Markov chain models (SIR)
 - model de evolution of the epidemics (number of Susceptible, Infected and Recovered individuals)
 - configuration of the population depends on present state only (given present the past and future history are independent)
 - also depend on some parameters (model assumed)

- Bayesian formulation and estimation
 - the parameters of the model are also considered random quantities with a probability distribution (density)
 - estimate the parameters using the so-called posterior distribution of the parameters
 - based on the relation among the posterior and prior distributions and the likelihood of the model, e.g.,
 - * θ : vector of parameters of the model
 - * \mathbf{Y} : observed data
 - * $P(\theta | \mathbf{Y}) \propto L(\mathbf{Y} | \theta) P(\theta)$

- Markov chain Monte Carlo methods (MCMC methods)
 - posterior distribution does not have closed form
 - dimension of the vector of parameters is very large
 - use MCMC algorithms and the Law of Large Number

$$\int f(x) g(x) dx \sim \frac{1}{M} \sum_{i=1}^M f(X_i)$$

- other quantities may also be estimated
 - * density distribution
 - * variance and standard deviation

– MCMC algorithms

* $\pi(\cdot)$: target distribution

$$\pi(x) = \frac{h(x)}{c}$$

* construct a Markov chain with $\pi(\cdot)$ as its stationary distribution

* allow the chain run for a suitable period and from that moment on the values simulated are drawn approximately from $\pi(\cdot)$

– Metropolis-Hastings algorithm

* $X_n = x$ then generate a value y using a transition probability P_{xy}

* calculate

$$\alpha(x, y) = \min \left\{ 1, \frac{\pi(y) P_{yx}}{\pi(x) P_{xy}} \right\}$$

* with probability $\alpha(x, y)$ accept the new value for the chain. i.e., $X_{n+1} = y$, otherwise no change occurs, i.e., $X_{n+1} = x$

– Gibbs sampling algorithm

* $P(x_1, x_2, \dots, x_d)$

* $P(x'_i | x'_1, x'_2, \dots, x'_{i-1}, x_{i+1}, x_d)$: easy to simulate value from it

* successively simulate the corresponding values

* Markov chain which under appropriate conditions is ergodic

EXAMPLE

- applied to influenza epidemics (theoretical part) - Cauchemez *et al.*, 2004
- Bayesian approach
- parameters of the models are estimated using Markov chain Monte Carlo methods
 - Metropolis-Hastings
- data from 334 households during winter 1999-2000
- absence/presence of influenza reported daily for 15 days

MATHEMATICAL DESCRIPTION OF THE PROBLEM

- k : number of households
- $N^{(k)}$: size of the k th household
- $N^{(f)}$: number of families
- $i \in \{1, 2, \dots, N^{(k)}\}$: individual in family k
- $j \in \{0, 1, 2, \dots, 14\}$: day at which individuals are observed

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$$Y_{ij}^{(k)} = \begin{cases} 1, & \text{clinical influenza was observed} \\ 0, & \text{otherwise} \end{cases}$$

- $I^{(k)}$: individuals report at least one day with clinical influenza in family k
- $S^{(k)}$: remaining members of the k family
- $Z_i^{(k)}$: first day of influenza for individual $i \in I^{(k)}$
- $\nu_i^{(k)}$: start of infectious period for individual i in family k
- $\psi_i^{(k)}$: end of infectious period for individual i in family k
- $\nu = \{\nu_i^{(k)}, i = 1, 2, \dots, N^{(k)}, k = 1, 2, \dots, N^{(f)}\}$
- $\psi = \{\psi_i^{(k)}, i = 1, 2, \dots, N^{(k)}, k = 1, 2, \dots, N^{(f)}\}$

- I : set of individuals in the sample such that the starts of its infectious period fell between 1 and 3 days
- transmission level
 - s : susceptible individual just before time t
 - α_s : instantaneous risk of infection from the community (depends on characteristics of individuals s)
 - ϵ_s : susceptibility to infection of individual s
 - β_i : ability of infective individual i to infect others
 - n : size of a given household where individual s belongs to

- contribution of infective individual i in a household of size n
 - β_i/n
 - β_i/n^η

where η is a parameter that also needs to be estimated

- $I^{(k)}(t) = \{i \in I^{(k)} : \nu_i^{(k)} < t \leq \psi_i^{(k)}\}$: group of infective just before time t
- $Y^{(k)} = \{Y_{ij}^{(k)}, i \in \{1, 2, \dots, N^{(k)}\}, j \in \{0, 1, 2, \dots, 14\}\}$: all observations in family k
- $Y = \{Y^{(k)}, k = 1, 2, \dots, N^{(f)}\}$: all observations

- $\lambda_s(t)$: instantaneous risk of infection at time t for the individual s

$$\lambda_s(t) = \alpha_s + \epsilon_s \sum_{i \in I(t)} \frac{\beta_i}{n}$$

BAYESIAN FORMULATION

- θ vector of parameter
- $P(\theta, \nu, \psi | Y) \propto L(Y | \nu, \psi) P(\nu, \psi | \theta) P(\theta)$

- Likelihood function of observed data

- $L(Y | \nu, \psi) = \prod_{i=1}^{N^{(f)}} P(Y^{(k)} | \nu^{(k)}, \psi^{(k)})$

- incubation period: 1 – 3 days (augmented data were compatible with observations)

- $\prod_{i \in I} I(Z_i - 3 < \nu_i < Z_i \text{ and } \nu_i < \psi_i)$

- Likelihood for augmented data

- $P(\nu, \psi | \theta) = \prod_{i=1}^{N(f)} P(\nu^{(k)} \psi^{(k)} | \theta)$

- $\psi_i - \nu_i \stackrel{\mathcal{D}}{=} \text{Gamma distribution with } \mu_i \text{ and standard deviation } \sigma_i \text{ and density } d_{\mu_i, \sigma_i}$

- condition on ν'

$$P(\nu, \psi | \theta) = \left(\prod_{i \in I} d_{\mu_i, \sigma_i}(\psi_i - \nu_i) \right)$$

$$\left(\prod_{i \in I - \{1\}} \lambda_i(\nu_i) e^{\int_{\nu'}^{\nu_i} \lambda_i(t) dt} \right)$$

$$\left(\prod_{s \in S} e^{\int_{\nu'}^{15} \lambda_s(t) dt} \right)$$

where

- * ν' : start of the infectious period for the first infected individual (in the entire community)
- * $I - \{1\}$: infective individuals without the first infective
- * 15: number of observed days

- vector of parameters

- $P(\theta)$: prior distribution of the vector of parameters
- $\epsilon = (\epsilon_C, \epsilon_A)$
- $\alpha = (\alpha_C, \alpha_A)$
- $\theta = (\mu, \sigma, \alpha, \beta, \epsilon, \eta)$

- $CPI(i)$: probability for individual i to be infected from the community in the 15-day follow-up (Community Probability of Infection)

$$CPI(i) = 1 - e^{-15 \alpha_i}$$

- $SAR(i \rightarrow s)$: (Secondary Attack Rate)

P(infective i infects susceptible s in household of size n)

$$SAR(i \rightarrow s) = 1 - \int_0^{\infty} e^{-\frac{\epsilon_s \beta_i}{n} t} d\mu_{i, \sigma_i}(t) dt$$

SIMULATION

- prior distributions used
 - $\mu, \sigma \stackrel{\mathcal{D}}{=} \text{Gamma, mean 3 and SD 2}$
 - $\alpha, \beta \stackrel{\mathcal{D}}{=} \text{Exp}(0.001)$
 - $\eta \stackrel{\mathcal{D}}{=} U[-3, 3]$
 - $\log(\epsilon_C) \stackrel{\mathcal{D}}{=} \text{Logistic with scale parameter 1}$
 - adults were reference category $\rightarrow \epsilon_A = 1$

- initial values

- $\nu_i^{(k)} \stackrel{\mathcal{D}}{=} U[Z_i^{(k)} - 3, Z_i^{(k)}]$

- $\psi_i^{(k)} - \nu_i^{(k)} \stackrel{\mathcal{D}}{=} U[0, 20]$

- $\alpha, \beta \stackrel{\mathcal{D}}{=} U[0, 1]$

- $\mu, \sigma \stackrel{\mathcal{D}}{=} U[0, 10]$

- $\epsilon_C, \eta \stackrel{\mathcal{D}}{=} \text{respective prior distributions}$

COMMENTS

- several hypothesis were tested by the authors
- selection of the best hypothesis for the model fitting the data was made using Bayes Factor
- substantial evidence that children are more susceptible than adults $\epsilon_C > 1$
- substantial evidence that community risk is larger for children than adults $\alpha_C > \alpha_A$
- poor evidence against the hypothesis that on average children are infectious for a longer period than adults $\mu_C > \mu_A$

- strong evidence favoring that household risk is larger with infectious child than with infectious adult
 $\beta_C > \beta_A$
- decisive evidence that SAR is larger with infectious child than with infectious adult, i.e., for $n \in \{2, 3, 4, 5\}$, $SAR_C(n) > SAR_A(n)$
- modifications and further generalisations may be considered
- simplified by using the software Winbugs (Spiegelhalter *et al.*, 1999)
- http://www.mrc_bsu.cam.ac.uk/bug/winbugs/