

Indian Buffet Epidemics

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RSC 2010

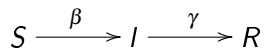


Outline

- 1 Epidemic Models and Inference
 - SIR Models

- 2 Indian Buffet Epidemics
 - MCMC Inference

SIR Models



$$N_t^S \quad N_t^I \quad N_t^R$$

$$N = N_t^S + N_t^I + N_t^R$$

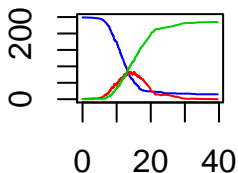
$$P(N_{t+dt}^S = j - 1 | N_t^S = j) = \beta N_t^S N_t^I dt$$

$$P(N_{t+dt}^I = j + 1 | N_t^I = j) = \beta N_t^S N_t^I dt$$

$$P(N_{t+dt}^I = j - 1 | N_t^I = j) = \gamma N_t^I dt$$

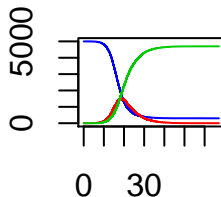
SIR examples

SIR N=250



$\beta = .75$, $\gamma = .25$

SIR N=1000



$\beta = .75$, $\gamma = .25$

Inference

- What data is available ?
 - Epidemic complete ?
 - Infection times ?
- MLE well known with full data
 - see Andersson and Britton (2000)
- Martingale estimator Becker and Hasofer (1997)
- MCMC estimates O'Neill and Roberts (1999)

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Indian Buffet Epidemics

- Need a model between homogeneous mixing and over complex models.
- Aim to fit the heterogeneity with two or three parameters that measure the departure from homogeneity.

Places and People

- Model heterogeneity in an epidemic amongst N people
- Each person belongs to 1 or more of many classes
 - e.g. households, schools, clubs, buses etcetera
- The classes are not specified
- A prior is put on class membership
 - represented as an $N \times K$ binary matrix Z
- An Indian Buffet Process

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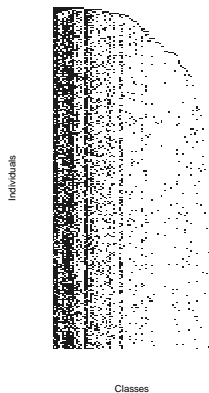
Indian Buffet Process

- Introduced by Griffiths and Ghahramani (2005)
- For each k ψ_k is the probability that an individual is in class k
- $\psi_k \sim \text{Beta}(\alpha/K, 1)$ with α being the strength parameter of the IBP.
- The model for Z is: $z_{ik} | \psi_k \sim \text{Bernoulli}(\psi_k)$ independently
- The process is obtained as $K \rightarrow \infty$

A culinary metaphor

- N customers enter a restaurant one after another.
- The j th customer selects each dish with probability m_k/j
 - where m_k is the number of previous customers who have chosen a dish.
- He then tries $\text{Poisson}(\alpha/j)$ new dishes.

Indian Buffet Process example



IBP Z generated with $N = 260$, $K = 260$, $\alpha = 15$

Indian Buffet Epidemic

- The state of individual j is at time t is $x_{j,t} \in \{S, I, R\}$.
- Given Z , infections are independent with transition rates given by
 - $P(x_{j,t+dt} = I | x_{j,t} = S, Z) = \sum z_{jk} \lambda_k N_{k,t}^I dt$
 - where $N_{k,t}^I$ is the number that are in class k and infective at time t and λ_k is the infection rate within group k .
- $N_{k,t}^I = \sum_j z_{jk} \mathbf{1}(x_{j,t} = I)$
- A basic model has λ_k the same for all k .
- Intuitively it is reasonable to assume a greater per person infection rate in a small group such as a household
 - so take $\lambda_k = \lambda N_k^{-\nu}$

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MCMC Inference

- Augmented data
- Parameterisation
- Proposal

Summary

- The limitations of existing epidemic models have been explored.
- A new model for epidemics incorporating heterogeneity has been introduced.
- Initial steps towards inference taken.

- Planned developments
 - Develop MCMC algorithms
 - Apply to real data
- Questions