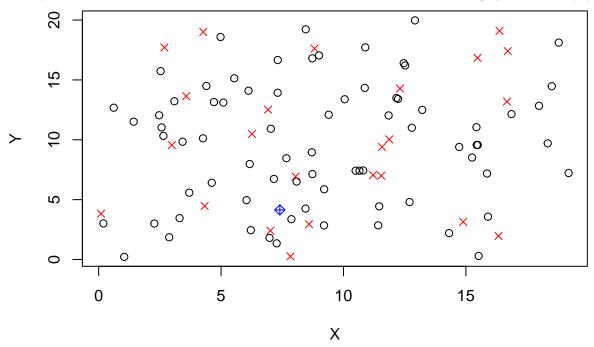
Unpaired Homogeneous General Stochastic Epidemic Inference

Benjamen Simon

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Load in the epidemic

The epidemic we have simulated has a modal number of infected individuals as roughly 25% of the population



Component functions of the MCMC algorithm

The likelihood / posterior

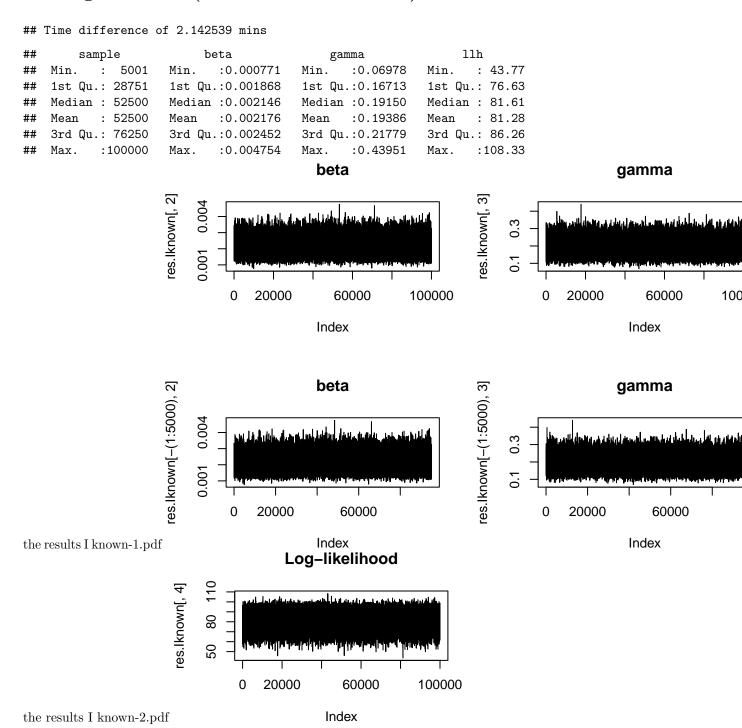
Thus we can now bring the two parts together and calculate the log-likelihood (more accurately the log-posterior for the infection times);

Metropolis-Hastings Acceptance Probability

The MCMC algorithm

Assuming we know the true infection times, we can remove the Metropolis-Hastings step and reduce the algorithm down to just a Gibbs sampler.

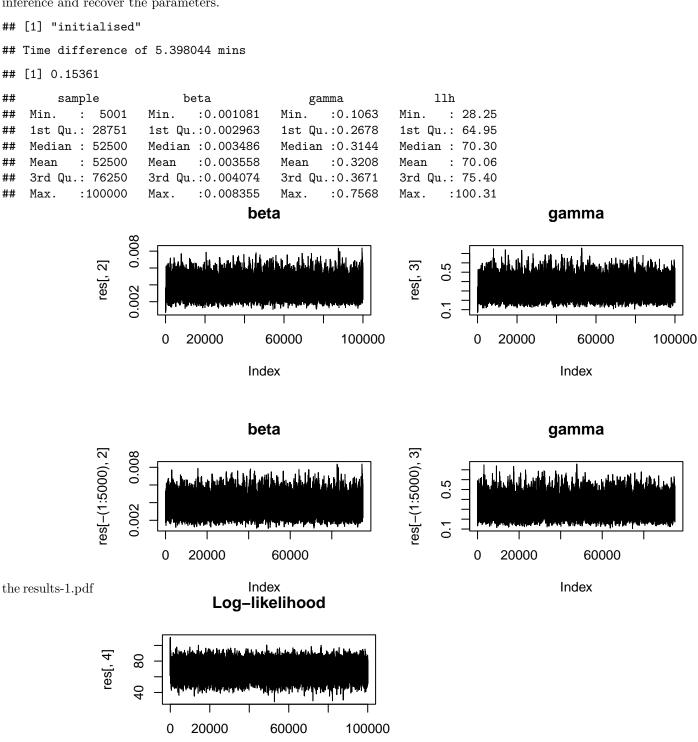
Making inference (Infection times known)



The algorithm estimates the mean, median, and quartiles of β to be 0.0021763 (0.0018677, 0.0021458, 0.0024525) compared to the true value of $\beta = 0.003$, with an effective sample size of 101202. It estimates the mean, median, and quartiles of γ to be 0.1938643 (0.1671271, 0.1915016, 0.2177905) compared to the true values of $\gamma = 0.23$, with an effective sample size of 96643.

Making inference (Infection times unknown)

Using our simulated epidemic from earlier, we will now apply our MCMC algorithm in an effort to make inference and recover the parameters.



The algorithm estimates the mean, median, and quartiles of β to be 0.0035577 (0.0029626, 0.0034858,

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the results-2.pdf

0.0040742) compared to the true value of $\beta=0.003$, with an effective sample size of 1699. It estimates the mean, median, and quartiles of γ to be 0.3207641 (0.2677693, 0.3143977, 0.3670952) compared to the true values of $\gamma=0.23$, with an effective sample size of 1636.