

# Simulation Journal for the *augSIR* package

August 1, 2015

## 7/24/2015 - Simulations to assess the effect of population size and binomial sampling probability on mixing

### Simulation parameters

- Population size: 50, 100, 150, 200, 300, 400, 500
- $R_0$ : 2, 5, 10
- $\rho$ : 0.05, 0.2, 0.5
- $\beta = \frac{R_0}{\text{population size}}, \mu = 1$
- Census interval = 0.2
- Three initializations for each scenario, diffuse priors for model parameters

### Measures of interest

- Proportion of proposed trajectories accepted
- Posterior distributions of model parameters
- Complete data log-likelihood

## Summary of results

- In small populations (50-200ish) the mcmc mixed well and all three initializations settled around roughly the same log-likelihood. The parameters were better recovered in smaller populations.
- In larger populations, the chains mixed poorly and were stuck in different modes of the likelihood.

## Next steps and other notes

- It was thought that the priors, while perhaps appropriate for each parameter separately, could jointly pull the value of  $R_0$  away from the true value. Will explore reparameterizing the model in terms of  $R_0$  and sampling parameters using M-H.
- Will run simulations to determine if problems persist with parameters fixed at the true values.
- Will include posterior samples of  $R_0$  in future simulations.

## 7/30/2015 - Simulations to assess overshooting behavior with fixed parameters

### Simulation parameters

- Population size: 50, 150, 300
- $R_0$ : 4, 8
- $\rho$  : 0.1, 0.4
- $\beta = \frac{R_0}{\text{population size}}, \mu = 1$
- Census interval = 0.05, 0.2
- Ten different initializations for each scenario

## Measures of interest

- Correspondence between true trajectory and augmented population level trajectories

## Summary of results

- Overshooting behavior is more apparent in larger populations, and when the observation times are further apart from one another.

## Next steps and other notes

- The fact that observation times and population size lead to overshooting behavior suggests that the method is sensitive to the amount of missing data.
- Will run Geweke style simulations alternating between simulating X—Y and Y—X, to compare the distribution of X via our method to the distribution of X simulated via Gillespie.

## 7/30/2015 - Simulations to assess whether reparameterizing model in terms of $R_0$ affects parameter posteriors

### Simulation parameters

- Population size: 50, 150, 300
- $R_0$ : 4, 8
- $\rho$ : 0.1, 0.4
- $\beta = \frac{R_0}{\text{population size}}, \mu = 1$
- Census interval = 0.05, 0.2
- Three initializations for each scenario.  $\log R_0 \sim N(0, 1.8)$ , diffuse priors for other parameters.

## Measures of interest

- Posterior distributions of model parameters
- Overshooting behavior of latent trajectories

## Summary of results

- Parameters are better recovered than in naively parameterized model.
- There is generally upward bias in  $R_0$ .
- Latent trajectories still overshoot the true trajectory.