## **Inference Tutorial 10**

https://people.maths.bris.ac.uk/~sw15190/TOI/jags\_user\_manual.pdf links to the JAGS manual. The str function in R is useful for examining the structure of an R object that you are unsure of. To use JAGS you have to produce a text file containing your model specification. On the university windows computers the editors 'notepad' or 'wordpad' can be used for this ('word' tends to create problems with hidden characters). It is probably best to create the file wherever you usually store your R script files, for example on the network drive accessed by opening the file manager (3rd icon from lower left in windows), and clicking on 'Computer' followed by 'MyFiles'. If you use this location then setwd("O:") (that's the letter, not zero) within R will cause R to look for jags (and other) files in this location (If you stored your files in a directory called, e.g., 'foo' within 'Myfiles' then use setwd("O:/foo")). Don't forget library(rjags).

The data loaded with the command

give the number of laboratory confirmed cases of H1N1 influenza in the Southern Hemisphere each week starting in the 17th week of 2009. The data cover the period of the southern hemisphere winter and are from the WHO. It may be reasonable to assume that the expected number of lab confirmed cases is proportional to the actual number of cases each week, but of course most cases will not be lab tested. One possible model for these data is that they result from a S-I-R (susceptible-infected-recovered) model, as follows. Let  $C_t$  denote the observed number of new cases in week t of the data, while  $S_t$  and  $I_t$  are the number of individuals susceptible to infection and actually infected, respectively. Then the model is

$$\begin{split} S_{t+1} &= S_t e^{-a_t I_t S_t}, \quad a_t \sim \operatorname{lognorm}(\alpha, \tau_a) \\ I_{t+1} &= (I_t - S_{t+1} + S_t) e^{-d_t I_t}, \quad d_t \sim \operatorname{lognorm}(\delta, \tau_d) \end{split}$$

The idea is that the susceptible population at t+1 is made up the susceptible population at t not infected between t and t+1, while the infected population is made up of the infected population at t plus those infected between t and t+1, who have not recovered (or died) between t and t+1.

The data contain no information to determine the scale of S alongside the other parameters, so w.l.o.g. we set  $S_1=1$ . For the sake of simplicity also use the prior  $I_1 \sim \text{lognormal}(-5,1)$  (-5 is the expectation of  $\log I_1$ , here). The observed cases are then related to the infected population via

$$C_t \sim \text{Poi}(\phi I_t)$$
.

Use dgamma (1, .01) priors for  $\tau_a$ ,  $\tau_d$  and  $\phi$  and  $N(0, 10^2)$  priors for  $\alpha$  and  $\delta$  (noting that in JAGS the normal is parameterized using the precision, rather than the variance, and dlnorm is the lognormal).

- 1. Implement the model in JAGS. jags.model has difficulty initialising the model. To help it along use the argument init=list (phi=1000, alpha=1, delta=.1).
- Generate 10000 samples from the posterior distribution of the model unknowns (remember to tell jags.samples which variables to output).
- 3. Produce trace plots for  $\alpha$ ,  $\delta$  and  $\phi$ , and an ACF for the most problematic (you'll probably need to use the argument max.lag=320 to get a sensible ACF plot). Estimate how many samples you would need to get an effective sample size around 500.
- 4. Produce plots showing 20 (approximately independent) random trajectories for  $S\phi$  and  $I\phi$  overlaid on the case data.
- 5. Produce a plot to visualize the average of the infection rate  $a_t$  at each time, according to the posterior, and interpret it.