Introduction to Bayesian Data Analysis Tutorial 7

(1) Problem 8.1 (Hoff) Components of variance: Consider the hierarchical model where

$$\theta_1, ..., \theta_m | \mu, \tau^2 \stackrel{\text{iid}}{\sim} \text{normal}(\mu, \tau^2)$$

 $y_{1,j}, ..., y_{n_i,j} | \theta_j, \sigma^2 \stackrel{\text{iid}}{\sim} \text{normal}(\theta_j, \sigma^2)$

For this problem, we will eventually compute the following:

$$Var[y_{i,j}|\theta_j,\sigma^2], Var[\bar{y}_{i,j}|\theta_j,\sigma^2], Cov[y_{i_1,j},y_{i_2,j}|\theta_j,\sigma^2]$$

$$Var[y_{i,j}|\mu,\tau^2], Var[\bar{y}_{\cdot,j}|\mu,\tau^2], Cov[y_{i_1,j},y_{i_2,j}|\mu,\tau^2]$$

First let's use our intuition to guess at the answers:

- (a) Which do you think is bigger, $Var[y_{i,j}|\theta_j,\sigma^2]$ or $Var[y_{i,j}|\mu,\tau^2]$? To guide your intuition, you can interpret the first as the variability of the Y's when sampling from a fixed group, and the second as the variability in first sampling a group, then sampling a unit from within the group.
- (b) Do you think $Cov[y_{i_1,j}, y_{i_2,j} | \theta_j, \sigma^2]$ is negative, positive, or zero? Answer the same for $Cov[y_{i_1,j}, y_{i_2,j} | \mu, \tau^2]$. You may want to think about what $y_{i_2,j}$ tells you about $y_{i_1,j}$ if θ_j is known, and what it tells you when θ_j is unknown.
- (c) Now compute each of the six quantities above and compare to your answers in a) and b).
- (d) Now assume we have a prior $p(\mu)$ for μ . Using Bayes' rule show that

$$p(\mu|\theta_1, ..., \theta_m, \sigma^2, \tau^2, \mathbf{y}_1, ..., \mathbf{y}_m) = p(\mu|\theta_1, ..., \theta_m, \tau^2)$$

Interpret in words what this means.

- (2) Problem 8.2 (Hoff) Sensitivity analysis: In this exercise we will revisit the study from Exercise 5.2, in which 32 students in a science classroom were randomly assigned to one of two study methods, A and B, with $n_A = n_B = 16$. After several weeks of study, students were examined on the course material, and the scores summarized by $\{\bar{y}_A = 75.2, s_A = 7.3\}$ and $\{\bar{y}_B = 77.5, s_B = 8.1\}$. We will estimate $\theta = \mu + \delta$ and $\theta_B = \mu \delta$ using the two-sample model and the prior distributions of Section 8.1.
 - (a) Let $\mu \sim N(75, 100)$, $1/\sigma^2 \sim \text{Gamma}(1, 100)$ and $\delta \sim N(\delta_0, \tau_0^2)$. For each combination of $\delta_0 \in \{-4, -2, 0, 2, 4\}$ and $\tau_0^2 \in \{10, 50, 100, 500\}$ obtain the posterior distribution of μ , δ and σ^2 and compute
 - (i) $Pr(\delta < 0|\mathbf{Y})$
 - (ii) a 95% posterior confidence interval for δ
 - (iii) the prior and posterior correlation of θ_A and θ_B
 - (b) Describe how you might use these results to convey evidence that $\theta_A < \theta_B$ to people of a variety of prior opinions.
- (3) Problem 8.3 (Hoff) The files school1.dat through school8.dat give weekly hours spent on homework for students sampled from eight different schools. We want to obtain posterior distributions for the true means for the eight different schools using a hierarchical normal model with the following prior parameters: $\mu_0 = 7, \gamma_0^2 = 5, \tau_0^2 = 10, \eta_0 = 2, \sigma_0^2 = 15, \nu_0 = 2$
 - (a) Run a Gibbs sampling algorithm to approximate the posterior distribution of $\{\theta, \sigma^2, \mu, \tau^2\}$. Assess the convergence of the Markov chain, and find the effective sample size for $\{\sigma^2, \mu, \tau^2\}$. Run the chain long enough so that the effective sample sizes are all above 1,000.
 - (b) Compute posterior means and 95% confidence regions for $\{\sigma^2, \mu, \tau^2\}$. Also, compare the posterior densities to the prior densities, and discuss what was learned from the data.
 - (c) Plot the posterior density for $R = \frac{\tau^2}{\sigma^2 + \tau^2}$ and compare it to a plot of the prior density on R. Describe the evidence for between-school variation.
 - (d) Obtain the posterior probability that θ_7 is smaller than θ_6 , as well as the posterior probability that θ_7 is the smallest of all the θ 's.
 - (e) Plot the sample averages $\bar{y}_1, ..., \bar{y}_8$ against the posterior expectations of $\theta_1,, \theta_8$, and describe the relationship. Also compute the sample mean of all observations and compare it to the posterior mean of μ . Estimate the shrinkage effect for each school.