

# Introduction to Bayesian Data Analysis

## Tutorial 8

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

$$\begin{aligned}\theta_1, \dots, \theta_m | \mu, \tau^2 &\stackrel{\text{iid}}{\sim} \text{normal}(\mu, \tau^2) \\ y_{1,j}, \dots, y_{n_j,j} | \theta_j, \sigma^2 &\stackrel{\text{iid}}{\sim} \text{normal}(\theta_j, \sigma^2)\end{aligned}$$

For this problem, we will eventually compute the following:

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

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

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  $\theta_j$  is known, and what it tells you when  $\theta_j$  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  $\mu$ . Using Bayes' rule show that

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

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  $\mu$ ,  $\delta$  and  $\sigma^2$  and compute

(i)  $Pr(\delta < 0 | \mathbf{Y})$

(ii) a 95% posterior confidence interval for  $\delta$

(iii) the prior and posterior correlation of  $\theta_A$  and  $\theta_B$

(b) Describe how you might use these results to convey evidence that  $\theta_A < \theta_B$  to people of a variety of prior opinions.

$$\tau_n^2 = \left[ \frac{1}{\tau_0^2} + \frac{n_1 + n_2}{\sigma^2} \right]^{-1}$$

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$$\tau_0^2 \Rightarrow \tau_n^2 \rightarrow \tau_n^2$$

$$\frac{\tau_n^2 - \tau_n^2}{\tau_n^2 + \tau_n^2}$$

when  $\delta_0 = 0$

then poster prob.

would be  $\gg 50\%$

even if people believe

no difference between

$\theta_A$  &  $\theta_B$

$\Rightarrow \theta_A$  in fact much smaller than  $\theta_B$ .