

# STATS 579 – INTERMEDIATE BAYESIAN MODELING

## Assignment # 3

1. An hypothetical study considers the lifespan of fluorescent light bulbs. Let  $y_1, \dots, y_n$  be the duration (in years) it takes for each of  $n$  light bulbs to fail. Assume that all tests are performed under laboratory conditions and observations are *iid*. Researchers are interested in whether bulb lifespan is better modeled with model  $M_1$ , an  $\text{Exp}(\lambda)$  distribution, or with model  $M_2$ , a Weibull  $(3, \lambda)$  distribution. For both models, the researchers assume  $p(\lambda) = e^{-\lambda}$ .

For this problem, please use the Exponential and Weibull parameterizations from your textbook, which give

$$\begin{aligned} y_E &\sim \text{Exp}(\lambda) \\ f(y_E | \lambda) &= \lambda \exp(-\lambda y) I_{(0, \infty)}(y) \\ y_W &\sim \text{Weibull}(\alpha, \lambda) \\ f(y_W | \alpha, \lambda) &= \lambda \alpha y^{\alpha-1} \exp(-\lambda y^\alpha) I_{(0, \infty)}(y) \end{aligned}$$

- (a) Obtain the marginal density for these data under each model. (HINT: Take advantage of conjugacy.)
  - (b) Obtain an expression for the Bayes factor comparing  $M_1$  to  $M_2$ .
  - (c) Evaluate the Bayes factor when the data are:  $\{8.05, 6.56, 3.20, 6.85, 5.67\}$ .
  - (d) Explain which model seems preferable based on the Bayes factor. Explain which model would be preferable if you had a prior belief that  $M_1$  were nine times more likely to be correct than  $M_2$ .
2. Using the same set-up as in Problem 1, the researchers want to compare these models in terms of AIC.
    - (a) Find the MLE for  $\lambda$  under  $M_1$  and  $M_2$ . Note that although both models have a parameter named  $\lambda$ , these parameters are not the same and may maximize at different values for each model. It may be helpful to write the MLEs as  $\hat{\lambda}_1$  and  $\hat{\lambda}_2$  to help distinguish them.
    - (b) Calculate AIC for each model using the data provided above. Which model seems preferable based on AIC?

3. Let  $y_1, \dots, y_n$  be independent conditional on some model parameter  $\theta$ . Let  $y_i \sim f(y_i | \theta)$  and let  $\theta$  have prior  $p(\theta)$ . Consider the conditional predictive ordinate for an observation  $y_j$ ,

$$CPO_j = f(y_j | y_{(j)}),$$

where  $y_{(j)}$  denotes the set  $\{y_1, \dots, y_{j-1}, y_{j+1}, \dots, y_n\}$ .

- (a) Show that

$$CPO_j = \frac{\int \prod_{i=1}^n f(y_i | \theta) p(\theta)}{\int \prod_{i \neq j} f(y_i | \theta) p(\theta)}.$$

- (b) Now show that

$$CPO_j^{-1} = \int \left[ \frac{1}{f(y_j | \theta)} \right] p(\theta | y) d\theta.$$