

## ECON 546: Extra Bayes Exercises

### Question 1

Suppose that we have  $n$  independent observations from a Poisson distribution, whose p.m.f. is:

$$p(y_i | \lambda) = \lambda^{y_i} \exp\{-\lambda\} / y_i! \quad ; y_i = 0, 1, 2, \dots; \quad \lambda > 0$$

Suppose that, *a priori*, we are totally ignorant about the value of  $\lambda$ .

Prove that the Bayes estimator of  $\lambda$ , when the loss function is quadratic, is  $\bar{y} = (\sum_{i=1}^n y_i / n)$ .

[**Hint:** The p.d.f. for a Gamma distribution is  $p(x) \propto x^{\alpha-1} \exp\{-\beta x\}$ ;  $\alpha, \beta, x > 0$ . The mean of this distribution is  $(\alpha / \beta)$ .]

### Question 2

Suppose that  $Y$  is **uniform** on the interval  $[0, \theta]$ , and that we have a random sample of  $n$  observations on  $Y$ .

Let the prior p.d.f. for  $\theta$  be:

$$p(\theta) = ak^a \theta^{-(a+1)} \quad ; \quad \theta \geq k; \quad a > 0.$$

(This is the p.d.f. for a Pareto distribution.)

- Show that this is the natural conjugate prior for  $\theta$ .
- Obtain the full posterior density for  $\theta$ , including the normalizing constant.
- Find the mean of the prior distribution.
- What is the Bayes estimator of  $\theta$  if the loss function is quadratic?

### Question 3

You probably know that J. M. Keynes made many important contributions to probability theory and statistics (as well as to economics, of course). His *Treatise on Probability* is a classic work that makes seminal contributions to the “subjective” theory of probability used by Bayesians. He also provided Keynes, 1911) the first modern treatment of “Laplace’s (1774) first law” - if we have an odd number of observations, “ $n$ ”, then the value of  $\theta$  that minimizes the expression

$\sum_{i=1}^n |y_i - \theta|$  is the median of the  $y_i$ ’s. (An odd number is needed to ensure that the median is

unique.) Now suppose that we have a random sample of “ $n$ ” (which you can assume to be an odd number of) observations from a Laplace (or “double exponential”) distribution. That is, the density function for an individual  $y_i$  is:

$$p(y_i | \theta, \lambda) = (2\lambda)^{-1} \exp\{-|y_i - \theta| / \lambda\}; \quad -\infty < y_i < \infty \quad ; \quad \lambda > 0.$$

- (a) Prove that the MLE for  $\theta$ , say  $\hat{\theta}$ , is the median of the sample, and that the MLE for  $\lambda$  is

$$\hat{\lambda} = \frac{1}{n} \sum_{i=1}^n |y_i - \hat{\theta}|.$$

- (b) It can be shown that  $E(y_i) = \theta$  and  $E(y_i^2) = (2\lambda^2 + \theta^2)$ . Suppose that we want to provide a unitless measure of the variability of the data. One such measure is the “coefficient of variation”,  $cv = \sqrt{\text{var.}(y_i)} / E(y_i)$ . Provide a consistent estimator for  $cv$ . What else can you say about the asymptotic properties of your estimator?
- (c) Now, suppose we **know**  $\theta$ , but that we are totally **ignorant** about  $\lambda$ . Obtain the posterior density for  $\lambda$ , and derive the Bayes’ estimator of this parameter when we have a zero-one loss function. Do this estimator and the MLE converge in probability to  $\lambda$  at the same rate as  $n \rightarrow \infty$ ?

**References:** Keynes, J. M. (1911), “The principal averages and the laws of error which lead to them”, *Journal of the Royal Statistical Society, Series A*, 74, 322-328.

Laplace, P. (1774), “Mémoire sur la probabilité des causes par les évènements”, *Mémoires de Mathématique et de Physique*, 6, 621-656.