## BASKIN SCHOOL OF ENGINEERING

## Department of Applied Mathematics and Statistics

First Year Exam: June 2014

## Problem AMS 206B:

Assume a Bernoulli experiment in which you perform n independent trials with success probability  $\theta$  and X counts the number of successes. Then,  $X|\theta \sim Binomial(n,\theta)$  with n known.

- 1. (20 points) Assume you observe X = x. Find the posterior density of  $\theta$ ,  $\pi(\theta|x)$ , under a uniform prior  $\theta \sim U(0,1)$ .
- 2. (40 points) Consider the loss function defined by

$$L(\theta, \hat{\theta}) = \frac{(\hat{\theta} - \theta)^2}{\theta(1 - \theta)}.$$

Find  $\hat{\theta}(x)$  the estimator that minimizes the Bayesian expected posterior loss under the scenario described in part 1. Note: Assume  $x \neq 0$ .

3. (40 points) Find Jeffreys prior on  $\theta$  and the corresponding posterior distribution under such prior.

Some useful information:

• If  $X \sim Binomial(n, \theta)$ , X has probability mass function given by

$$Pr(X = x) = \binom{n}{x} \theta^x (1 - \theta)^{n-x}, \quad x = 0, 1, \dots, n.$$

• Beta distribution. If  $X \sim Beta(\alpha, \beta)$  its pdf is given by

$$p(x|\alpha,\beta) = \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \theta^{\alpha-1} (1-\theta)^{\beta-1}, \quad 0 \le x \le 1.$$

In addition,  $E(X) = \alpha/(\alpha + \beta)$ , and  $V(X) = \alpha\beta/((\alpha + \beta)^2(\alpha + \beta + 1))$ .

•  $\Gamma(\alpha + 1) = \alpha \Gamma(\alpha)$ .

## **SOLUTION**

• *Part* (a):

$$\pi(\theta|x) \propto \theta^x (1-\theta)^{n-x}$$

therefore, the posterior is a Beta(x+1, n-x+1).

• Part (b): The expected posterior loss is

$$E(L(\theta, \hat{\theta})|X = x) = \int_0^1 c \frac{(\hat{\theta} - \theta)^2}{\theta(1 - \theta)} \theta^{x+1-1} (1 - \theta)^{n-x+1-1} d\theta,$$

with  $c = \Gamma(n+2)/(\Gamma(x+1)\Gamma(n-x+1))$ . To minimize this function we take derivatives w.r.t.  $\hat{\theta}$  and so,

$$\frac{dE(L(\theta,\hat{\theta})|X=x)}{d\hat{\theta}} = 2c \int_0^1 (\hat{\theta} - \theta)\theta^{x-1} (1-\theta)^{n-x-1} d\theta = 0,$$

implies that the optimal Bayes estimator is  $\hat{\theta} = x/n$ .

Note that the second derivative is

$$2c \int_0^1 \theta^{x-1} (1-\theta)^{n-x-1} d\theta > 0$$

• Part (c): Jeffreys prior is given by  $\pi(\theta) = |I(\theta)|^{1/2}$  with

$$I(\theta) = -E_{X|\theta} \frac{d^2 \log f(x|\theta)}{d\theta^2}.$$

In this case we obtain that  $\pi(\theta) \propto \theta^{-1/2} (1-\theta)^{1/2}$ , which corresponds to a Beta(1/2, 1/2). The corresponding posterior is a Beta(x+1/2, n-x+1/2).