## Statistics 330b/600b, Math 330b spring 2013 Homework # 10 Due: Thursday 18 April

Please attempt at least the starred problems.

\*[1] For a fixed n let  $\mu$  denote counting measure on  $\mathfrak{X} = \{0, 1, \dots, n\}$ . For each  $\theta$  in  $\Theta = (0, 1)$  let  $\mathbb{P}_{\theta}$  denote the Bin $(n, \theta)$  probability measure, which has density

$$p(x,\theta) = {\binom{n}{x}} \theta^x (1-\theta)^{n-x}$$
 with respect to  $\mu$ .

Let T denote the map that projects  $\mathfrak{X} \times \Theta$  onto  $\mathfrak{X}$ . That is,  $T(x, \theta) = x$ . Suppose the prior  $\pi$  has density

 $g(\theta) = \theta^{-1}(1-\theta)^{-1}$  with respect to Lebesgue measure  $\mathfrak{m}$  on  $\mathcal{B}(\Theta)$ 

Let  $\lambda = \mu \otimes \mathfrak{m}$  and  $\mathbb{P} = \pi \otimes \mathcal{P}$  where  $\mathcal{P} = \{\mathbb{P}_{\theta} : \theta \in \Theta\}$ . [Despite the choice of letter,  $\mathbb{P}$  is not a probability measure.]

- (i) Find the density of  $\mathbb{P}$  with respect to  $\mu \otimes \mathfrak{m}$ .
- (ii) Define  $\mathbb{Q} = T\mathbb{P}$ . Find  $\mathbb{Q}\{x\}$  for each x in  $\mathfrak{X}$ .
- (iii) Find a collection of measures  $\Gamma = \{\gamma_x : x \in \mathcal{X}\}$  on  $\mathcal{B}(\Theta)$  for which  $\mathbb{P} = \mu \otimes \Gamma$ . That is,  $\mathbb{P}f(x,\theta) = \mu^x \gamma_x^\theta f(x,\theta)$  for each nonnegative, jointly measurable f on  $\mathcal{X} \times \Theta$ . [In other words, the set of measures  $\{\delta_x \otimes \gamma_x : x \in \mathcal{X}\}$  is a  $T, \mu$ -disintegration of  $\mathbb{P}$ .] You should be able to spot a suitable  $\Gamma$  without much calculation.
- (iv) Show that there does not exist a  $T, \mathbb{Q}$ -disintegration of  $\mathbb{P}$ . Hint: Consider what such a disintegration would say about  $\mathbb{P}\{x=0, |\theta|<1/2\}$ .

**Remark.** The measure  $\gamma_x$  plays the role of the posterior distribution. For example, in decision theory using squared error loss, the function  $\delta(x)$  that minimizes  $\gamma_x^{\theta}(\theta - \delta(x))^2$  for each x has useful properties.

\*[2] Let  $\mu$  denote counting measure on  $\{1, 2\}$  and  $\lambda$  denote Lebesgue measure on  $\mathcal{B}(\mathbb{R})$ . For each  $\theta = (\alpha, \gamma_1, \gamma_2)$  in  $\Theta := (0, 1) \times \mathbb{R}^2$  define

$$f(x, y, \theta) = \alpha \{ y = 1 \} \phi(x - \gamma_1) + (1 - \alpha) \{ y = 2 \} \phi(x - \gamma_2),$$

a probability density with respect to  $\lambda \times \mu$ . (Here  $\phi$  is the N(0, 1) density function.) The marginal density for x is the mixture  $\alpha \phi(x - \gamma_1) + (1 - \alpha)\phi(x - \gamma_2)$ .

For  $\mathbf{x} = (x_1, \dots, x_n)$  and  $\mathbf{y} = (y_1, \dots, y_n)$  define

$$p(\mathbf{x}, \mathbf{y}, \theta) = \prod_{i \le n} f(x_i, y_i, \theta)$$

- (i) Find the maximum likelihood estimator for  $\theta$  based on x and y.
- (ii) Now suppose we only observe **x**. The marginal density for **x** is  $\nu^{\mathbf{y}} p(\mathbf{x}, \mathbf{y}, \theta)$  where  $\nu$  is the *n*-fold product measure  $\mu^n$ . The EM algorithm iteratively updates an estimate  $\theta_0$  by the value  $\theta_1$  that maximizes the function

$$H_{\theta_0}(\theta) = \mathbb{P}_{\theta_0}\left(\log p(\mathbf{x}, \mathbf{y}, \theta) \mid \mathbf{x}\right).$$

Find the value for  $\theta_1$ .