

## ECE 313: Exam I

Monday, July 7, 2014

7:00 p.m. — 8:15 p.m.

269 Everitt Lab

1. (a) Let  $L$  denote the restaurant quality level. Then  $\mathbb{P}[L = \ell] = \frac{1}{5}$  for  $\ell = 1, \dots, 5$ . Let  $Y_\ell$  denote the Poisson distribution with mean  $10\ell$  for  $\ell = 1, \dots, 5$ . Let  $X$  denote the number of people dining in the restaurant. According to the total probability formula,

$$\mathbb{P}[X = 40] = \sum_{\ell=1}^5 \mathbb{P}[X = 40|L = \ell]\mathbb{P}[L = \ell] = \frac{1}{5} \sum_{\ell=1}^5 \mathbb{P}[Y_\ell = 40] = \frac{1}{5} \sum_{\ell=1}^5 \frac{(10\ell)^{40}}{40!} e^{-10\ell}.$$

- (b) By the Bayes formula,

$$\mathbb{P}[L \geq 3|X = 40] = \frac{\sum_{\ell=3}^5 \mathbb{P}[X = 40|L = \ell]\mathbb{P}[L = \ell]}{\mathbb{P}[X = 40]} = \frac{\sum_{\ell=3}^5 \frac{(10\ell)^{40}}{40!} e^{-10\ell}}{\sum_{\ell=1}^5 \frac{(10\ell)^{40}}{40!} e^{-10\ell}}.$$

2. (a) Define the likelihood ratio  $\Lambda(k) = \frac{\mathbb{P}[X=k|H_1]}{\mathbb{P}[X=k|H_0]}$ . Then  $\Lambda(0) = \frac{1-q}{1-p} = \frac{2}{3}$  and  $\Lambda(1) = \frac{q}{p} = \frac{3}{2}$ . Hence, the ML decides  $H_0$  when  $X = 0$  and decides  $H_1$  when  $X = 1$ .
- (b) By definition,

$$p_{\text{false alarm}} = \mathbb{P}[\text{ML says } H_1|H_0] = \mathbb{P}[X = 1|H_0] = p = 0.4$$

$$p_{\text{miss}} = \mathbb{P}[\text{ML says } H_0|H_1] = \mathbb{P}[X = 0|H_1] = 1 - q = 0.4.$$

- (c) The MAP decides  $H_0$  if  $\Lambda(k) < \frac{P(H_0)}{P(H_1)}$  and decides  $H_1$  if  $\Lambda(k) > \frac{P(H_0)}{P(H_1)}$ . In this problem,  $\frac{P(H_0)}{P(H_1)} = \frac{1}{3}$  and therefore the MAP always say  $H_1$  no matter what values the observation of  $X$  takes.
- (d) By definition

$$p_e = P(H_0)p_{\text{false alarm}} + P(H_1)p_{\text{miss}} = \frac{1}{4} \times 1 + \frac{3}{4} \times 0 = \frac{1}{4}.$$

- (e) Note that  $p_{\text{false alarm}}$  for the MAP rule is zero means that the MAP always say  $H_0$ , which implies that  $\frac{P(H_0)}{P(H_1)} > \frac{3}{2}$  (here is the inequality is strict, because we assume the tie is broken in favor of  $H_1$ ), and thus  $0.6 < P(H_0) \leq 1$ .
3. (a) The solution is  $\frac{d}{m}$ . We made an error only if  $F(x)$  is not identical to  $G(x)$  and  $n$  is a root of  $F(x) - G(x)$ . Since there are at most  $d$  roots of  $F(x) - G(x)$  and  $n$  is uniformly chosen from  $\{1, 2, \dots, m\}$ , the probability that we made an error is at most  $\frac{d}{m}$ .
- (b) The solution is  $\left(\frac{d}{m}\right)^r$ . We made an error if  $F(x)$  is not identical to  $G(x)$  and all  $n_1, \dots, n_r$  are roots of  $F(x) - G(x)$ . By part (a), the probability that  $n_i$  is the root of  $F(x) - G(x)$  is at most  $\frac{d}{m}$  for  $i = 1, \dots, r$ . Since  $n_1, \dots, n_r$  are independently chosen, the probability that all  $n_1, \dots, n_r$  are roots of  $F(x) - G(x)$  is at most  $\left(\frac{d}{m}\right)^r$ .

4. (a)  $X \sim \text{Binomial}(n, \frac{1}{m})$ ; the limiting distribution of  $X$  is  $\text{Pois}(\alpha)$  whose probability mass function is  $p(k) = \frac{\alpha^k}{k!} \exp(-\alpha)$ .
- (b) Since  $X \sim \text{Bin}(n, \frac{1}{m})$ , so  $\mathbb{P}(X = 0) = (1 - \frac{1}{m})^n$ . Define Bernoulli random variable  $Y_i$  such that  $Y_i = 1$  if bin  $i$  is empty and  $Y_i = 0$  if bin  $i$  is not empty. Then  $\mathbb{P}[Y_i = 1] = \mathbb{P}[X = 0] = (1 - \frac{1}{m})^n$ . Since the fraction of empty bins is given by  $\frac{1}{m} \sum_{i=1}^m Y_i$ , by the linearity of the expectation, we get the the fraction of empty bins on average is  $\frac{1}{m} \sum_{i=1}^m \mathbb{E}[Y_i] = (1 - \frac{1}{m})^n$ .  
In the limit as  $m \rightarrow \infty$ ,  $X \sim \text{Pois}(\alpha)$  and thus  $\mathbb{P}(X = 0) = \exp(-\alpha)$ . Then the fraction of empty bins on average is  $\exp(-\alpha)$ .
- (c) First consider all the possible assignments of balls into bins. For the  $i$ -th ball, it could land in any one of  $m$  bins and thus there are  $m$  possible assignments for it. In total, there are  $m^n$  possible assignments of  $n$  balls into  $m$  bins. Next consider all the possible assignments of balls into bins such that all bins contain exactly  $\alpha$  balls. For the first bin, since it contains  $\alpha$  balls, there are  $\binom{n}{\alpha}$  different choices of balls into the first bin. Then consider the second bin, since it contains  $\alpha$  balls and there are  $n - \alpha$  balls left unassigned, there are  $\binom{n-\alpha}{\alpha}$  different choices of balls into the second bin. Continue this argument for all the bins, and in total there are  $\binom{n}{\alpha} \binom{n-\alpha}{\alpha} \cdots \binom{\alpha}{\alpha} = \frac{n!}{(\alpha!)^m}$  possible assignments of balls into bins such that all bins contain exactly  $\alpha$  balls. In conclude, the probability that all bins contain exactly  $\alpha$  balls is  $\frac{n!}{(\alpha!)^m m^n}$ .
- (d) The probability of seeing at least one ball from every bin after  $m$  rounds is  $\frac{\alpha^m m!}{n^m} = \frac{m!}{m^m}$ . For each round, there are  $n$  different choices of balls, so in  $m$  rounds there are  $n^m$  different choices of balls. To see at least one ball from every bin after  $m$  rounds, we need exactly 1 ball from every bin. Since each bin contains exactly  $\alpha$  balls, there are  $\alpha^m$  different choices of  $m$  balls. Furthermore, the order of seeing these  $m$  balls could be arbitrarily and there are  $m!$  different orders of these  $m$  balls.  
Another way to derive is as follows. To see at least one ball from every bin after  $m$  rounds means that each round we pick a ball from a bin that is different from what we've picked, so the answer is given by  $\frac{m}{m} \frac{m-1}{m} \frac{m-2}{m} \cdots \frac{1}{m} = \frac{m!}{m^m}$ .
- (e) Let  $T_i$  denote the number of additional rounds to see at least one ball from  $i$  bins, given that we have seen at least one ball from  $i-1$  bins. Then  $T_i$  is distributed as geometric distribution with parameter  $\frac{m-(i-1)}{m}$ . Therefore,  $\mathbb{E}[\sum_{i=1}^m T_i] = \sum_{i=1}^m \frac{m}{m-(i-1)} = m \sum_{i=1}^m \frac{1}{i}$ .
5. (a) Each needle is lying in the circle with probability  $\frac{\pi}{4}$  and thus  $X \sim \text{Bin}(n, \frac{\pi}{4})$ . Hence,  $\mathbb{P}[X = k] = \binom{n}{k} (\frac{\pi}{4})^k (1 - \frac{\pi}{4})^{n-k}$  for  $1 \leq k \leq n$ .
- (b) Let  $\hat{\pi}$  denote the maximum likelihood estimator of  $\pi$ . Then  $\hat{\pi} = \arg \max_x \binom{n}{k} (\frac{x}{4})^k (1 - \frac{x}{4})^{n-k}$ . Hence,  $\hat{\pi} = \frac{4k}{n}$ .
- (c) Note that the mean of  $X$  is  $n\frac{\pi}{4}$ . To estimate  $\pi$  within 0.1 means to estimate  $\frac{\pi}{4}$  with 0.025. Using Chebychev inequality, we have

$$\mathbb{P} \left\{ \frac{\pi}{4} \in \left[ \frac{X}{n} - \frac{a}{2\sqrt{n}}, \frac{X}{n} + \frac{a}{2\sqrt{n}} \right] \right\} \geq 1 - \frac{1}{a^2}.$$

Pick  $a = 5$  such that  $1 - \frac{1}{a^2} = 96\%$ , and we need  $\frac{a}{2\sqrt{n}} = 0.025$ . Solving for  $n$ , we get  $n = 10^4$ .