

## Answers of Master's Exam, Fall, 2007

### Probability

- 1) a) There are 1 C, 5 A's, 2 R's, 1 D, 2 B's

$$\binom{1}{1} \binom{5}{1} \binom{2}{1} \binom{1}{1} / \binom{11}{4} = (1)(5)(2)(1) / 330$$

b)  $(9/11)^5 (2/11)$

c) Let  $X = (\# \text{ cards with R})$ ,  $X \sim \text{Binomial}(605, 2/11)$ ,  $E(X) = 110$ ,  $\text{Var}(X) = 90$ ,  $z = (131.5 - 110) / \sqrt{90} = 2.2663$ ,  $P(X > 131) = 1 - \Phi(z) = 0.0117$ .

d) 0.1103

e)  $5! 2! 2! 1! 1! / 11! = (5)(2)(2)(4)(1)(3)(1)(2)(2)(1)(1) / (11!)$

- 2) a)  $X_1 = U^{1/2}$

b) For  $-1 \leq s < 0$ , let  $z = 1 + s$ .  $F_S(s) = z^2 - z^4 / 2 + (4/3)z^3 y - z^2 y^2 = z^3(1-s) - z^4 / 2 + (4/3)z^3 y$

For  $0 \leq s \leq 1$ ,  $F_S(s) = 1 - F_S(-s)$  by symmetry around zero.

Comment: This problem can take much time to solve accurately. I advise students to indicate the double integrals involved (over triangles) without simplifying them.

c)  $F_Y(y) = F_X((1 + y^{1/2})/2) - F_X((1 - y^{1/2})/2) = (1/4)[(1 + y^{1/2})^2 - (1 - y^{1/2})^2] = y^{1/2}$  for  $0 \leq y \leq 1$ , 0 for  $y < 0$ , 1 for  $y > 1$

d) Let  $\sigma^2 = \text{Var}(X_1) = \text{Var}(X_2) = \text{Var}(X_3)$ ,  $Z_1 = X_2 + 2 X_3$ ,  $Z_2 = 5 X_1 + 4 X_2 - 3 X_3$ . Then  $\text{Cov}(Z_1, Z_2) = -2 \sigma^2$ ,  $\text{Var}(Z_1) = 5 \sigma^2$ ,  $\text{Var}(Z_2) = 50 \sigma^2$ ,  $\rho(Z_1, Z_2) = -2 \sigma^2 / \sqrt{(5)(50) \sigma^4} = -2 / \sqrt{250}$ . Notice that the correlation coefficient does not depend on  $\sigma^2$ . I saved some time in writing by defining the symbols  $Z_1, Z_2$ .

e) By independence  $E(W) = E(X_1^2) E(1/X_2) E(1/X_3) = (2/3)2^2 = 8/3$ . **Don't make the mistake** of thinking that  $E(1/X_2)$  is equal to  $1/E(X_2)$ .

- 3) a)  $\lim_{n \rightarrow \infty} F_n(x) = F(x)$  for every  $x$  on the real line at which  $F$  is continuous.

b)  $\lim_{n \rightarrow \infty} F_n(x) = 0$  for  $x < 0$ , 1 for  $x \geq 1$ . The distribution with cdf  $F(x) = 0$  for  $x < 1$ , 1 for  $x \geq 1$  (mass one at 1) is therefore the limiting distribution of the sequence  $\{F_n\}$ . Though  $F$  is discontinuous at  $x = 1$ , we need not have  $\lim_{n \rightarrow \infty} F_n(1)$ , though we do for this example. If, for example,  $F_n(x) = 0$  for  $x \leq 1$ ,  $n(x-1)$  for  $1 < x \leq 1 + 1/n$ , 1 for  $x \geq 1 + 1/n$ , then  $\lim_{n \rightarrow \infty} F_n(1) = 0$ , but the limiting cdf is  $F$ .

- c) Let  $X_1, \dots, X_n$  be iid with mean  $\mu$ , variance  $\sigma^2$ . Let  $S_n = X_1 + \dots + X_n$ .

Let  $Z_n = (S_n - n \mu) / \sqrt{n \sigma^2}$ . Then for any real number  $z$   $\lim_{n \rightarrow \infty} P(Z_n \leq z) = \Phi(z)$ , the cdf of the  $N(0, 1)$  distribution.

- 4) a)  $f_X(x) = (2/\pi)\sqrt{1-x^2}$  for  $0 \leq x \leq 1$   
 b)  $f_{Y|X}(y|x) = (1/\pi)/f_X(x)$  for  $0 \leq y \leq \sqrt{1-x^2}$ ,  $-1 < x < 1$ .  
 Don't forget to give the domain of functions!  
 c) Given  $X = x$ ,  $Y$  is uniformly distributed on the interval  $[-\theta, \theta]$ , where  $\theta = \sqrt{1-x^2}$ . The expected square of such the  $U(-\theta, +\theta)$  distribution is  $\theta^2/3$ .  
 Thus  $E(Y^2 | X = x) = (1-x^2)/3$ , so  $E(Y^2 | X) = (1-X^2)/3$ .  
 $V = E(X + Y^2 | X) = X + (1-X^2)/3$ .

5) a)  $Y$  takes the values  $-2$  and  $+2$ .  $P(Y = +2) = \int_0^1 3x^2 E(Y|X=x) dx$   
 $= \int_0^1 3x^2 x dx = 3/4$ .  $P(Y = -2) = 1/4$ .

b)  $f_{X|Y}(x|y) = 3x^2(1-x)/(1/4) = 12x^2(1-x)$  for  $0 \leq x < 1$ , 0 otherwise.

## Statistics

6) a)  $\hat{a} = \min(X_1, \dots, X_n)$   $P(\hat{a} \leq x) = 1 - P(\hat{a} > x) = 1 - [e^{-2(x-a)}]^n = 1 - e^{-2n(x-a)}$   
 for  $x \geq a$

b)  $\mu = a + 1/2$ , so the MOM Est. of  $a$  is  $\bar{X} - 1/2$ .

c)  $E(\hat{a}) = a + (1/2)/n$ , so the bias is  $1/(2n)$ .

7) Let  $D_i = (\text{Before Value} - \text{After Value})$  for patient  $i$ ,  $i = 1, \dots, 10$ .

Suppose the  $D_i$  are a random sample from the  $N(\mu_D, \sigma_D^2)$  distribution.

Test  $H_0: \mu_D \leq 0$  vs  $H_a: \mu_D > 0$ .

The  $D_i$  are : 2.7, 2.1, 2.2, 2.6, 2.4, -0.5, -1.0, 1.3, -1.8, -2.0.

We observe  $\bar{D} = 1.2$ ,  $S^2 = 2.76$ ,  $t = 2.28$ . The 0.99 quantile of the  $t$ -distribution with 9 df is 2.82, so we should not reject  $H_0$  at the  $\alpha = 0.01$  level.

b) Either the sign test or the Wilcoxon signed rank test could be used. The  $p$ -value for the sign test is  $P(X \leq 4) = 386/2^{10} = 0.37695$ . Do not reject.

The Wilcoxon signed rank statistic is  $W_+ = 55 - (\text{sum of ranks of negative values}) = 55 - (1+2+4) = 48$ .  $P(W_+ \geq 48) = P(W_- \leq 7) = 1 + 7 + 5 + 3 + 1 + 2)/2^{10} = 20/1024 > 0.01$ , so again we do not reject at the  $\alpha = 0.01$  level.

8) a)  $X_1 \sim \text{Binomial}(n_1 = 1000, p_1)$ ,  $X_2 \sim \text{Binomial}(n_2 = 800, p_2)$ , independent.

Let  $\hat{p}_1 = X_1 / n_1$ ,  $\hat{p}_2 = X_2 / n_2$ .  $\hat{\Delta} = \hat{p}_1 - \hat{p}_2$ . 99% confidence interval on  $\Delta$  is

$$[\hat{\Delta} \pm 2.58 \sqrt{\hat{p}_1 (1 - \hat{p}_1)/n_1 + \hat{p}_2 (1 - \hat{p}_2)/n_2}].$$

b)  $[-0.07 - 0.0456, -0.07 + .0456]$

c) These samples were taken and the corresponding interval determined in such a way that repeated sampling would determine intervals which contain the true population difference in population proportions (first proportion minus the second proportion) for 99% of all repetitions.

d) 8295, using the “worst case”, for which  $p_1 = p_2 = 0.5$ .

9) a) Reject  $H_0$  for Level of Significance =  $P(T > 25.5 \mid \mu = 2)$

$$= 1 - \Phi((25.5 - 18)/\sqrt{9}) = 1 - \Phi(2.5) = 0.0062$$

b) Use the Neyman-Pearson Lemma, with the likelihood for  $\mu = \mu_0 > 2$  in the numerator, and the likelihood for  $\mu = 0$  in the denominator. The test reduces to critical region  $T \geq k$  for some  $k$ , equivalently  $\bar{X} > k^*$  for some constant  $k^*$ .

$$\text{c) Power} = 1 - \Phi((25.5 - 27)/\sqrt{9}) = 1 - \Phi(-0.5) = \Phi(0.5) = 0.691$$

10) a) Let  $Q(\beta) = \sum (Y_i - \beta/x_i)^2$ , so  $\frac{\delta}{\delta\beta} Q(\beta) = \sum (-1/x_i) (Y_i - \beta/x_i)$ . Setting this equal to zero and solving for  $\beta$ , we get  $\hat{\beta}$  as given.

$$\text{b) } \hat{\beta} = (\sum Y_i / x_i) / (\sum 1/x_i^2) = \beta + \sum (\epsilon_i/x_i) / (\sum 1/x_i^2). \text{ Since } E(\epsilon_i) = 0 \text{ for each } i, \\ E(\hat{\beta}) = \beta.$$

$$\text{c) } \text{Var}(\hat{\beta}) = \text{Var}(\sum (\epsilon_i/x_i) / (\sum 1/x_i^2)) = (1 / (\sum 1/x_i^2))^2 \sum (1/x_i^2) \sigma^2 = \sigma^2 / (\sum 1/x_i^2).$$

$$\text{d) } \hat{\beta} = 0.8, S^2 = 48.6/3 = 16.2 \quad 90\% \text{ CI on } \beta: 0.8 \pm 4.89.$$

11) Let  $W = (\# \text{ women on the jury})$  Under random sampling  $W$  has a hypergeometric distribution.

$$P(W \leq 1) = P(W = 0) + P(W = 1) = \frac{\binom{12}{7} \binom{6}{0}}{\binom{18}{7}} + \frac{\binom{12}{6} \binom{6}{1}}{\binom{18}{7}}.$$

12)  $W = 11$ ,  $P(W \leq 11 | H_0) = 2/\binom{9}{4} = 1/42$ , so the p-value for a 2 sided test is  $1/21$ .