

# MATH 563 HOMEWORK 1

## SOLUTIONS

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**GRADING SCALE: 5 points per problem. About half the points given for correct mathematical result, and the other half for reasoning / explanation of the solution.**

1. (Problem 5.1.)

Let  $X = \#$  color blind people in a sample of size  $n$ . Then  $X \sim \text{binomial}(n, p)$ , where  $p = .01$ . The probability that a sample contains a color blind person is  $P(X > 0) = 1 - P(X = 0)$ , where  $P(X = 0) = \binom{n}{0}(.01)^0(.99)^n = .99^n$ . Thus,

$$P(X > 0) = 1 - .99^n > .95 \Leftrightarrow n > \log(.05)/\log(.99) \approx 299.$$

2. (Exercise 5.15.)

a.

$$\bar{X}_{n+1} = \frac{\sum_{i=1}^{n+1} X_i}{n+1} = \frac{X_{n+1} + \sum_{i=1}^n X_i}{n+1} = \frac{X_{n+1} + n\bar{X}_n}{n+1}.$$

b.

$$\begin{aligned} nS_{n+1}^2 &= \frac{n}{(n+1)-1} \sum_{i=1}^{n+1} (X_i - \bar{X}_{n+1})^2 \\ &= \sum_{i=1}^{n+1} \left( X_i - \frac{X_{n+1} + n\bar{X}_n}{n+1} \right)^2 && \text{(use (a))} \\ &= \sum_{i=1}^{n+1} \left( X_i - \frac{X_{n+1}}{n+1} - \frac{n\bar{X}_n}{n+1} \right)^2 \\ &= \sum_{i=1}^{n+1} \left[ (X_i - \bar{X}_n) - \left( \frac{X_{n+1}}{n+1} - \frac{\bar{X}_n}{n+1} \right) \right]^2 && (\pm \bar{X}_n) \\ &= \sum_{i=1}^{n+1} \left[ (X_i - \bar{X}_n)^2 - 2(X_i - \bar{X}_n) \left( \frac{X_{n+1} - \bar{X}_n}{n+1} \right) + \frac{1}{(n+1)^2} (X_{n+1} - \bar{X}_n)^2 \right] \\ &= \sum_{i=1}^n (X_i - \bar{X}_n)^2 + (X_{n+1} - \bar{X}_n)^2 - 2 \frac{(X_{n+1} - \bar{X}_n)^2}{n+1} + \frac{n+1}{(n+1)^2} (X_{n+1} - \bar{X}_n)^2 \\ &\qquad\qquad\qquad \left( \text{since } \sum_{i=1}^n (X_i - \bar{X}_n) = 0 \right) \\ &= (n-1)S_n^2 + \frac{n}{n+1} (X_{n+1} - \bar{X}_n)^2. \end{aligned}$$

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**This material is *not* to be shared outside Math563 Spring 2021 at Illinois Tech.**

3. (Exercise 5.4.) Let  $X_i|P \sim_{iid} \text{Bernoulli}(P)$ , for  $i = 1, \dots, n$ , and let  $P \sim \text{Uniform}(0, 1)$ .  
 a) By the definition of Bernoulli distribution,  $\Pr(X_i = x_i|P = p) = p^{x_i}(1-p)^{1-x_i}$ . Thus by the independence of  $X_i$ s we have

$$\begin{aligned} & \Pr(X_1 = x_1, X_2 = x_2, \dots, X_k = x_k|P = p) \\ &= \prod_{i=1}^k p^{x_i}(1-p)^{1-x_i} = p^{\sum_{i=1}^k x_i} \cdot (1-p)^{\sum_{i=1}^k (1-x_i)} = p^t \cdot (1-p)^{(k-t)} \end{aligned}$$

Here  $t = \sum_{i=1}^k x_i$ .

We know that for probability density functions,  $f(x) = \int_{-\infty}^{\infty} f(x, y)dy = \int_{-\infty}^{\infty} f(x|y)f(y)dy$ , also  $P \sim U(0, 1)$ , thus we have

$$\begin{aligned} & \Pr(X_1 = x_1, X_2 = x_2, \dots, X_k = x_k) \\ &= \int_{-\infty}^{\infty} \Pr(X_1 = x_1, X_2 = x_2, \dots, X_k = x_k|P = p)f_P(p)dp \\ &= \int_0^1 \Pr(X_1 = x_1, X_2 = x_2, \dots, X_k = x_k|P = p) \cdot 1dp \\ &= \int_0^1 p^t \cdot (1-p)^{(k-t)}dp = \frac{\Gamma(t+1)\Gamma(k-t+1)}{\Gamma(k+2)} = \frac{t!(k-t)!}{(k+1)!} \end{aligned}$$

- b) Because usually  $\int_a^b f(x)dx \cdot \int_a^b g(x)dx \neq \int_a^b f(x)g(x)dx$ , thus  
 $\prod_{i=1}^n \Pr(X_i = x_i) = \prod_{i=1}^n \int_0^1 p^{x_i}(1-p)^{(1-x_i)}dp$   
 $\neq \int_0^1 \prod_{i=1}^n p^{x_i}(1-p)^{(1-x_i)}dp = \Pr(X_1 = x_1, \dots, X_n = x_n)$ .  
 4. (either Exercise 5.13 or Exercise 5.16)

5.13

$$\begin{aligned} \mathbb{E}(c\sqrt{S^2}) &= c\sqrt{\frac{\sigma^2}{n-1}}\mathbb{E}\left(\sqrt{\frac{S^2(n-1)}{\sigma^2}}\right) \\ &= c\sqrt{\frac{\sigma^2}{n-1}}\int_0^\infty \sqrt{q}\frac{1}{\Gamma(\frac{n-1}{2})2^{(n-1)/2}}q^{(\frac{n-1}{2})-1}e^{-q/2}dq, \end{aligned}$$

Since  $\sqrt{S^2(n-1)/\sigma^2}$  is the square root of a  $\chi^2$  random variable. Now adjust the integrand to be another  $\chi^2$  pdf and get

$$\mathbb{E}(c\sqrt{S^2}) = c\sqrt{\frac{\sigma^2}{n-1}} \cdot \frac{\Gamma(n/2)2^{n/2}}{\Gamma((n-1)/2)2^{((n-1)/2)}} \underbrace{\int_0^\infty \frac{1}{\Gamma(n/2)2^{n/2}}q^{(n-1)/2} - \frac{1}{2}e^{-q/2}dq}_{=1 \text{ since } \chi_n^2 \text{ pdf}}$$

So  $c = \frac{\Gamma(\frac{n-1}{2})\sqrt{n-1}}{\sqrt{2}\Gamma(\frac{n}{2})}$  gives  $\mathbb{E}(cS) = \sigma$ .

5.16

Since  $X_i \sim N(i, i^2)$ ,  $i = 1, 2, 3$  are independent, we denote  $Z_i = \frac{X_i - i}{i}$ , then  $Z_i \sim N(0, 1)$ , i.i.d. We will use  $Z_1, Z_2, Z_3$  to construct the required random variables.

(a) From Lemma 5.3.2.a,  $Z_i^2 \sim \chi^2(1)$ ,  $i = 1, 2, 3$ , then from Lemma 5.3.2.b,  $Z_1^2 + Z_2^2 + Z_3^2 \sim \chi^2(1 + 1 + 1) = \chi^2(3)$ , which is the desired random variable.

(b) There are two ways to construct this random variable. Firstly, a  $t(2)$  distributed variable can be constructed as the ratio  $\frac{U}{\sqrt{V/2}}$ , where  $U \sim N(0, 1)$  and  $V \sim \chi^2(2)$ . We take  $U = Z_1, V = Z_2^2 + Z_3^2$  to obtain  $\frac{Z_1}{\sqrt{(Z_2^2 + Z_3^2)/2}} \sim t(2)$ .

Alternatively, one can also directly use definition 5.3.4 to construct  $\frac{\bar{Z}}{S/\sqrt{3}} \sim t(3-1) = t(2)$ , where  $\bar{Z} = \frac{1}{3}(Z_1 + Z_2 + Z_3)$  and  $S = \sqrt{\frac{Z_1^2 + Z_2^2 + Z_3^2 - 3\bar{Z}^2}{2}}$ .

**Remark: Please note that  $S^2 \neq \frac{1}{3}(Z_1^2 + Z_2^2 + Z_3^2)$ . There are some confusions in your homeworks.**

(c) Since  $Z_1^2 \sim \chi^2(1), Z_2^2 + Z_3^2 \sim \chi^2(2)$ , we have  $\frac{Z_1^2}{(Z_2^2 + Z_3^2)/2} \sim F(1, 2)$ .