

Name:

Section:

## STAT 113, Spring 99 – Midterm 2

*Note:* On all problems, please show your work. Just the correct answer without justification and intermediate results is not acceptable.

1. Let  $(X, Y)$  have the bivariate p.d.f.  $f(x, y) = c$  in the portion of the first quadrant bounded by  $x + y = 1$ , i.e.,

$$f(x, y) = \begin{cases} c & \text{if } x \geq 0, y \geq 0 \text{ and } x + y \leq 1, \\ 0 & \text{otherwise} \end{cases}$$

- 1a. [2pts] Find  $c$ .

*Hint:* A sketch of the set  $A = \{x \geq 0, y \geq 0 \text{ and } x + y \leq 1\}$  will help (it is simply the first quadrant bounded by  $x + y = 1$ ).

$$1.0 = \int_0^1 \underbrace{\int_0^{1-x} c \, dy}_{=c(1-x)} dx = c(x - x^2/2) \Big|_0^1 = c/2 \Rightarrow c = 2.0$$

*Note:* You don't really need to do the integral. Just argue that the p.d.f. is constant and  $\{x \geq 0, y \geq 0 \text{ and } x + y \leq 1\}$  is a triangle of area  $1/2 \Rightarrow \int \int f(x, y) dx dy = c 1/2$

- 1b. [2pts] Find the marginal p.d.f.  $f(x)$ .

$$f(x) = \int_0^{1-x} 2 \, dy = 2(1 - x), \quad 0 \leq x \leq 1$$

- 1c. [2pts] Find the conditional p.d.f.  $f(y|x)$ .

$$f(y|x) = \frac{f(x, y)}{f(x)} = \begin{cases} 1/(1 - x) & \text{if } 0 \leq y \leq 1 - x \\ 0 & \text{otherwise} \end{cases}$$

- 1d. [2pts] Find  $E(X)$ .

$$E(X) = \int_0^1 x 2(1 - x) \, dx = 2(x^2/2 - x^3/3) \Big|_0^1 = 1 - 2/3 = 1/3.$$

- 1e. [2pts] Are  $X$  and  $Y$  independent? Why/why not?

No,  $X$  and  $Y$  are not independent because

$$f(x, y) = 2 \neq f(x) f(y) = 2(1 - x) 2(1 - y).$$

- 2.** A candy bar wrapper says “Net weight 1.4oz.” The bars actually vary in weight. To be reasonably confident that most bars weigh at least 1.4 oz., a manufacturer may adjust the production so that the mean weight is 1.5 oz. Assume weights are approximately normally distributed with standard deviation 0.05oz.

*Note:* In your solution, denote with  $X$  the weight of a candy bar.

- 2a.** [2pts] Find a transformation of  $X$ , such that the transformed random variable is standard normal  $N(0, 1)$ .

$$Z = (X - \mu)/\sigma \text{ with } \mu = 1.5 \text{ and } \sigma = 0.05.$$

- 2b.** [4pts] Find the proportion of bars weighting less than the advertised 1.4 oz.

The weight  $X$  of a candy bar is  $N(\mu, \sigma)$  distributed with  $\mu = 1.5$  and  $\sigma = 0.05$ . Therefore

$$P(X \leq 1.4) = P\left(\frac{X - 1.5}{0.05} \leq \frac{1.4 - 1.5}{0.05}\right) = P(Z \leq -2) = 0.5 - 0.4772 = 0.02.$$

- 2c\*.** [2pts] Find the weight  $\mu$  to which the manufacturer has to adjust the production to make sure that less than 1% of the bars weigh less than 1.4 oz.

Find  $\mu$  such that

$$P(X \leq 1.4) = P\left(\frac{X - \mu}{0.05} \leq \frac{1.4 - \mu}{0.05}\right) = P(Z \leq \underbrace{\frac{1.4 - \mu}{0.05}}_c) = 0.01.$$

From Table 4 (page 1094 in the book) we find  $c = 2.33 \Rightarrow$

$$c = -2.33 = \frac{1.4 - \mu}{0.05} \Rightarrow \mu = 1.4 - 0.05 \cdot 2.33 = \dots$$

- 3.** A pharmaceutical company is developing a new drug for stroke patients. Recovery after stroke is measured in Scandinavian Stroke Score (SSS) points. Denote with  $X_i$  the SSS points for a patient treated with the new drug and let  $\mu = E(X_i)$  and  $\sigma^2 = Var(X_i)$  denote expected value and variance. The drug is considered effective if  $\mu \geq 4$ . In a clinical trial  $n$  patients are treated with the new drug. Let  $X_1, \dots, X_n$  denote the measurements for these  $n$  patients. Let  $\bar{X}$  denote the sample mean.

- 3a.** [2pts] Find  $E(\bar{X})$  and  $Var(\bar{X})$ . (Note: Since we do not know  $\mu$  and  $\sigma$ , the answer will be an expression with  $\mu$  and  $\sigma$ , not a number.)

$$E(\bar{X}) = \mu \text{ and } Var(\bar{X}) = \sigma^2/n.$$

- 3b.** [1pt] Assuming that the  $X_i$  are normally distributed, find the sampling distribution  $f(\bar{x})$  for  $\bar{X}$ .

$$\bar{X} \text{ is normally distributed, } \bar{X} \sim N(\mu, \sigma^2/n).$$

- 3c.** [3pts] Assume  $\sigma = 10$  and  $\mu = 4$ . For  $n = 100$ , find  $P(\bar{X} > 1.5)$ .

$$\text{Note that } E(\bar{X}) = \mu = 4 \text{ and } Var(\bar{X}) = \sigma^2/n = 1.0.$$

$$P(\bar{X} > 1.5) = P\left(\frac{X - 4}{1} > \frac{1.5 - 4}{1}\right) = P(Z > -2.5) = 0.5 - P(0 \leq Z \leq 2.5) = 0.5 - 0.4938 = 0.9938$$

- 3d.** [2pts] For your answer to **3b** and **3c**, did you need to assume that the  $X_i$  be normally distributed? Why/why not?

No. For large  $n$  the sampling distribution of  $\bar{X}$  is normal even if the probability distribution of  $X_i$  is not normal (Central Limit Theorem).

- 3e\*.** [2pts] Again assume  $\sigma = 10$  and  $\mu = 4$ .

Find the minimum sample size  $n$  such that  $P(\bar{X} > 1.5) > 0.95$ .

$$\text{Note that } E(\bar{X}) = \mu = 4 \text{ and } Var(\bar{X}) = \sigma^2/n = 100/n.$$

$$P(\bar{X} > 1.5) = P\left(\frac{X - 4}{10/\sqrt{n}} > \frac{1.5 - 4}{10/\sqrt{n}}\right) = P(Z > \underbrace{\frac{1.5 - 4}{10/\sqrt{n}}}_c) = 0.95.$$

From Table 4 (p1094) we find that  $c = -1.65 \Rightarrow$

$$\frac{1.5 - 4}{10/\sqrt{n}} = -1.65 \Rightarrow n \geq (10 \cdot 1.65/2.5)^2 = 43.56$$

i.e.,  $n \geq 44$ .

4. Consider again the SSS measurements  $X_i$ ,  $i = 1, \dots, n$ , for the  $n = 100$  stroke patients discussed in problem 3. Assume  $X_i \sim N(\mu, \sigma^2)$ . Both parameters,  $\mu$  and  $\sigma$ , are unknown.

*Note:* You do not need to have solved problem 3 to work on this problem.

- 4a. [3pts] Assuming that  $\sigma^2 = 100$ , find

$$Pr(s^2 \geq 121)$$

Note that  $\chi^2 = (n-1)s^2/\sigma^2 \sim \chi_{n-1}^2$ . Using Table 8 (p1100) we find:

$$P(s^2 \geq 121) = P\left(\frac{(n-1)s^2}{\sigma^2} \geq \frac{(n-1)121}{\sigma^2}\right) = P(\chi^2 > 119.79) \approx 10\%$$

- 4b. [2pts] Find the sampling distribution  $f(y)$  of  $y = \frac{\bar{x} - \mu}{s/\sqrt{n}}$

(state the name of the distribution and values of the relevant parameters).

$t = \bar{x} - \mu/s/\sqrt{n}$  is t-distributed with  $\nu = n - 1$  degrees of freedom.

- 4c. [3pts] For  $n = 100$  patients we observe sample mean  $\bar{x} = 1.7$  and sample variance  $s^2 = 121.0$ .

Find a cutoff  $a$  such that

$$P(\mu \geq a) = 0.90$$

(Remember that  $\sigma^2$  is *unknown*).

From Table 7 (p1099) we find  $P(t \leq 1.29) = 0.90$  (there is no line for  $\nu = 99$  – use the cutoff for  $\nu = 120$  degrees of freedom). Therefore

$$P\left(\frac{\bar{x} - \mu}{s/\sqrt{n}} \leq 1.29\right) = 0.9$$

and

$$P(\mu \geq \bar{x} - 1.29s/\sqrt{n}) = P(\mu \geq 1.7 - 1.29 \cdot 1.1) = 0.9$$

- 4d\*. [2pts] For  $n = 100$  patients we observe sample mean  $\bar{x} = 1.7$  and sample variance  $s^2 = 121.0$ .

Find cutoffs  $c$  and  $d$  such that

$$Pr(c \leq \sigma^2 \leq d) = 0.95$$

and  $Pr(\sigma^2 \geq c) = 0.025$ .

For  $\chi^2 \sim \chi_{99}^2$  we find from Table 8 (p1100)

$$P(74.22 \leq \chi^2 \leq 129.6) \approx 0.95 \Rightarrow \dots \Rightarrow P((n-1)s^2/129.6 \leq \sigma^2 \leq (n-1)s^2/74.22) \approx 0.95$$

i.e.,  $c = 99 \cdot 121/129.6$  and  $d = 99 \cdot 121/74.22$ .