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M378K Introduction to Mathematical Statistics

Fall 2024

University of Texas at Austin

**In-Term Exam II**

Instructor: Milica Čudina

**Notes:** This is a closed book and closed notes exam. The maximal score on the real exam will be 100 points.

**There are many ways in which any single problem can be solved. The solutions herein are just one possible way to tackle the given problems.**

**Time:** 50 minutes

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All written work handed in by the student is considered to be  
**their own work, prepared without unauthorized assistance.**

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**Signature:**

**2.1. Formulas.** If  $Y$  has the binomial distribution with parameters  $n$  and  $p$ , then  $p_Y(k) = \mathbb{P}[Y = k] = \binom{n}{k} p^k (1-p)^{n-k}$ , for  $k = 0, \dots, n$ ,  $\mathbb{E}[Y] = np$ ,  $\text{Var}[Y] = np(1-p)$ . The binomial coefficients are defined as follows for integers  $0 \leq k \leq n$ ,  $\binom{n}{k} = \frac{n!}{k!(n-k)!}$ . The moment generating function of  $Y$  is given by  $m_Y(t) = (pe^t + q)^n$ .

If  $Y$  has a geometric distribution with parameter  $p$ , then  $p_Y(k) = p(1-p)^k$  for  $k = 0, 1, \dots$ ,  $\mathbb{E}[Y] = \frac{1-p}{p}$ ,  $\text{Var}[Y] = \frac{1-p}{p^2}$ . Its mgf is  $m_Y(t) = \frac{p}{1-qe^t}$  for  $t$  such that  $qe^t < 1$ .

If  $Y$  has a Poisson distribution with parameter  $\lambda$ , then  $p_Y(k) = e^{-\lambda} \frac{\lambda^k}{k!}$  for  $k = 0, 1, \dots$ ,  $\mathbb{E}[Y] = \text{Var}[Y] = \lambda$ . Its mgf is  $m_Y(t) = e^{\lambda(e^t-1)}$ .

If  $Y$  has a uniform distribution on  $[l, r]$ , its density is

$$f_Y(y) = \frac{1}{r-l} \mathbf{1}_{(l,r)}(y),$$

its mean is  $\frac{l+r}{2}$ , and its variance is  $\frac{(r-l)^2}{12}$ . Let  $U \sim U(0, 1)$ . The mgf of  $U$  is  $m_U(t) = \frac{1}{t}(e^t - 1)$ .

If  $Y$  has the standard normal distribution, then its mean is zero, its variance is one, and its density equals

$$\varphi(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} \quad x \in \mathbb{R}.$$

Its mgf is  $m_Y(t) = e^{\frac{t^2}{2}}$ .

If  $Y$  has the exponential distribution with parameter  $\tau$ , then its cumulative distribution function is  $F_Y(y) = 1 - e^{-\frac{y}{\tau}}$  for  $y \geq 0$ , its probability density function is  $f_Y(y) = \frac{1}{\tau} e^{-y/\tau}$  for  $y \geq 0$ . Also,  $\mathbb{E}[Y] = \text{SD}[Y] = \tau$ . Its mgf is  $m_Y(t) = \frac{1}{1-\tau t}$ .

The mgf of  $Y \sim \Gamma(k, \tau)$  is

$$m_Y(t) = \frac{1}{(1-\tau t)^k} \text{ for } t < 1/\tau.$$

Its expectation is  $k\tau$  and its variance is  $k\tau^2$ . The  $\chi^2$ -distribution with  $n$  degrees of freedom is the special case  $\Gamma(\frac{n}{2}, 2)$

## 2.2. DEFINITIONS.

**Problem 2.1.** (10 points) Write down the definition of the **moment generating function** of a random variable  $Y$ .

**Solution:**

$$m_Y(t) = \mathbb{E}[e^{tY}]$$

for all  $t \in \mathbb{R}$  such that the above expectation exists. In fact, we say that the moment-generating function **exists** if there exists a positive number  $b$  such that  $m_Y(t)$  is finite for all  $t$  such that  $|t| \leq b$ .

**Problem 2.2.** (10 points) Write down the definition of the **random sample** of size  $n$  from a distribution  $D$ .

**Solution:** A *random sample* of size  $n$  from a distribution  $D$  is a random vector

$$(Y_1, Y_2, \dots, Y_n)$$

such that:

1.  $Y_1, Y_2, \dots, Y_n$  are **independent**, and
  2.  $Y_i$  has the distribution  $D$  for every  $i = 1, 2, \dots, n$ .
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### 2.3. TRUE/FALSE QUESTIONS.

**Problem 2.3.** (5 points) Let  $Y_1, Y_2, \dots, Y_n$  be a random sample. Then,

$$\frac{1}{n} \sum_{i=1}^n \mathbf{1}_{(0,\infty)}(Y_i)$$

is a well-defined statistic. *True or false?*

**Solution: TRUE**

**Problem 2.4.** (5 points) Let  $Y_1, Y_2, \dots, Y_n$  be a random sample from  $N(\mu, \sigma)$  with  $\mu$  **known** and  $\sigma$  **unknown**. Then,

$$\frac{1}{n} \sum_{i=1}^n (Y_i - \mu)^2$$

is a well-defined estimator for  $\sigma^2$ . *True or false?*

**Solution: TRUE**

### 2.4. Free-response problems.

Please, explain carefully all your statements and assumptions. Numerical results or single-word answers without an explanation (even if they're correct) are worth 0 points.

**Problem 2.5.** (15 points) Let  $(Y_1, Y_2)$  be a random vector with the joint pdf

$$f_{Y_1, Y_2}(y_1, y_2) = \frac{1}{4} \mathbf{1}_{\{-1 \leq y_1 \leq 1\}} \mathbf{1}_{\{-1 \leq y_2 \leq 1\}}.$$

Find  $\mathbb{P}[|Y_1| + |Y_2| \leq 1/2]$ .

**Solution:** The pair  $(Y_1, Y_2)$  is uniformly distributed over the square  $[-1, 1] \times [-1, 1]$ , while the region  $\{(y_1, y_2) \in [-1, 1]^2 : y_1^2 \leq y_2^2\}$  corresponds to the square with vertices  $(\frac{1}{2}, 0)$ ,  $(0, \frac{1}{2})$ ,  $(-\frac{1}{2}, 0)$  and  $(0, -\frac{1}{2})$ . The side length of this square is  $1/\sqrt{2}$ , so its total area is  $\frac{1}{2}$ . The total area of the square  $[-1, 1]$  is 4, and, since we are dealing with a geometric-probability problem, the answer is  $\frac{1}{2}/4 = \frac{1}{8}$ .

**Problem 2.6.** (10 points) Let  $Y \sim U(l, r)$  What is the moment generating function of  $Y$ ?

**Solution:** See **Example 6.1.5** from the lecture notes.

**Problem 2.7.** (20 points) In Croatia, if you go to the chocolate-factory store, you can buy broken off chunks of rice-puff chocolate. From past experience, we know that the weight of the individual chunks has mean of 40 grams and standard deviation of 5 grams. Assume that the weights of individual pieces of chocolate are independent.

You buy 400 chocolate chunks. What is the probability that the total weight exceeds 16128 grams?

**Solution:** Let  $n = 400$  denote the total number of chocolate chunks. Let  $Y_i, i = 1, \dots, n$  be the random variables which stand for the weights of individual chunks. Then, their total weight can be expressed as

$$S = Y_1 + \dots + Y_n$$

We can use the Central Limit Theorem (CLT) here since  $n = 400$ . We have that  $S$  is approximately normal with mean  $40(400) = 16000$  and standard deviation  $5\sqrt{400} = 100$ .

The probability we are asked to calculate is

$$\mathbb{P}[S > 16128] = \mathbb{P}\left[\frac{S - 16000}{100} > \frac{16128 - 16000}{100}\right] \approx \mathbb{P}[Z > 1.28]$$

where  $Z \sim N(0, 1)$ . We get

$$\mathbb{P}[S > 16128] \approx 1 - \Phi(1.28).$$

If we consult the standard normal tables, we get our answer as  $1 - 0.8997 = 0.1003$ .

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## 2.5. MULTIPLE CHOICE QUESTIONS.

**Problem 2.8.** (5 points) Let  $Y$  be a uniform random variable on  $[0, 1]$ , and let  $W = Y^2$ . The pdf of  $W$  is

(a)  $\frac{1}{2\sqrt{|w|}} \mathbf{1}_{\{-1 < w < 1\}}$

(b)  $\frac{1}{\sqrt{w}} \mathbf{1}_{\{0 < w < 1\}}$

(c)  $\frac{1}{2\sqrt{w}} \mathbf{1}_{\{0 < w < 1\}}$

(d)  $2w \mathbf{1}_{\{0 < w < 1\}}$

(e) none of the above

**Solution:** The correct answer is **(c)**.

We use the  $h$ -method.  $g(y) = y^2$  and  $h(w) = \sqrt{w}$ . Therefore

$$f_W(w) = f_Y(h(w))h'(w) = \frac{1}{2\sqrt{w}} \mathbf{1}_{\{0 < w \leq 1\}}.$$

**Problem 2.9.** (5 points) Let  $Y_1, Y_2, \dots, Y_n$  be independent, identically distributed normal random variables with mean  $\mu$  and standard deviation  $\sigma$ . What is the distribution of the random variable  $Y$  defined as

$$Y = \left( \frac{Y_1 - \mu}{\sigma} \right)^2 + \left( \frac{Y_2 - \mu}{\sigma} \right)^2 + \dots + \left( \frac{Y_n - \mu}{\sigma} \right)^2 ?$$

- (a)  $N(0, \sqrt{n})$
- (b)  $\chi^2(n)$
- (c)  $\chi^2(n-1)$
- (d)  $N(0, n^2)$
- (e) **None of the above.**

(Note: In our notation  $N(\mu, \sigma)$  means normal with mean  $\mu$  and *standard deviation*  $\sigma$ .)

**Solution:** The correct answer is **(b)**.

**Problem 2.10.** (5 points) You are monitoring a cash register at a store. A seemingly endless queue of customers is waiting. The times it takes for Cassie the Cashier to check out a single customer is exponential with mean 5 minutes for each customer. Moreover, the customer service times are independent. Cassie the cashier can go on a break after every batch of 10 customers leave. She just came in from a break. What is the distribution of her waiting time until the next break?

- (a)  $\Gamma(10, 5)$ , i.e.,  $k = 10, \tau = 5$
- (b)  $\Gamma(50, 1)$ , i.e.,  $k = 50, \tau = 1$
- (c)  $\chi^2(25)$
- (d)  $E(50)$
- (e)  $E(1/50)$

**Solution:** The correct answer is **(a)**.

We need the distribution of the sum of 10 independent exponential random variables  $E(\tau)$  with parameter  $\tau = 5$ . Each  $E(5)$  is a special case of the gamma distribution with parameters  $k = 1$  and  $\tau = 5$ . Gammas are additive in the shape parameter, so the result is  $\Gamma(10, 5)$ .



**Problem 2.11.** (5 points) A math graduate student basically survives on espresso and chocolate. Their daily chocolate consumption is normally distributed with mean 16 oz and standard deviation 4 oz. Their daily espresso consumption is normally distributed with mean 20 oz and standard deviation 3 oz. Assume that espresso consumption and chocolate consumption are independent.

What is the probability that chocolate consumption exceeds coffee consumption in a single day?

- (a) About 0.2119
- (b) About 0.2839
- (c) About 0.4207
- (d) About 0.4432
- (e) **None of the above.**

**Solution:** The correct answer is **(a)**.

Let  $Y_1$  be the chocolate consumption and let  $Y_2$  be the espresso consumption. We are given that  $Y_1$  and  $Y_2$  are independent. Also,

$$Y_1 \sim N(\mu_1 = 16, \sigma_1 = 4) \quad \text{and} \quad Y_2 \sim N(\mu_2 = 20, \sigma_2 = 3).$$

We need to calculate  $\mathbb{P}[Y_1 > Y_2] = \mathbb{P}[Y_1 - Y_2 > 0]$ . From the given information, we can conclude that

$$Y_1 - Y_2 \sim N(\mu = -4, \sigma = \sqrt{3^2 + 4^2} = 5).$$

So,

$$\mathbb{P}[Y_1 - Y_2 > 0] = \mathbb{P}\left[\frac{Y_1 - Y_2 - \mu}{\sigma} > \frac{0 - \mu}{\sigma}\right] = \mathbb{P}[Z > 0.8]$$

where  $Z \sim N(0, 1)$ . From the standard normal tables, we get  $1 - 0.7881 = 0.2119$ .

**Problem 2.12.** (5 points) Let  $Y_1, \dots, Y_{100}$  be independent random variables with the Bernoulli  $B(p)$  distribution, with  $p = 0.2$ . The best approximation to  $\bar{Y} = \frac{1}{n}(Y_1 + \dots + Y_n)$  (among the offered answers) is

- (a)  $N(0, 1)$
- (b)  $N(100, 20)$
- (c)  $N(0.2, 0.04)$
- (d)  $N(20, 4)$
- (e)  $N(20, 20)$

(Note: In our notation  $N(\mu, \sigma)$  means normal with mean  $\mu$  and standard deviation  $\sigma$ .)

**Solution:** The correct answer is (c).

The sum  $W = Y_1 + \cdots + Y_n$  is binomially distributed with mean  $np = 20$  and variance  $np(1-p) = 16$ , i.e., standard deviation 4. It is well approximated by a normal  $N(20, 4)$ . Since  $\bar{Y} = \frac{1}{n}W$ , its best approximation will be a normal with mean  $\frac{1}{100}20 = 0.2$  and standard deviation  $\sigma = \frac{1}{100}4 = 0.04$ .

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