## Math 340 / 640 Fall 2024 Midterm Examination Two

## Professor Adam Kapelner November 14, 2024

## Instructions

Full Name

This exam is 110 minutes (variable time per question) and closed-book. You are allowed **one** page (front and back) of a "cheat sheet", blank scrap paper (provided by the proctor) and a graphing calculator (which is not your smartphone). Please read the questions carefully. Within each problem, I recommend considering the questions that are easy first and then circling back to evaluate the harder ones. No food is allowed, only drinks.

signature

date

Problem 1 Let  $X_1, X_2, \ldots \stackrel{iid}{\sim}$  some continuous random variable with  $0 < \mu < \infty$  and  $0 < \sigma^2 < \infty$ . We also use the standard notation

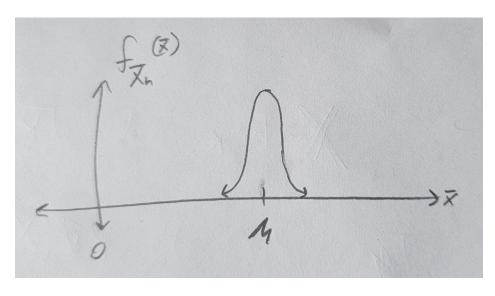
$$\bar{X}_n := \frac{1}{n} \sum_{i=1}^n X_i$$
 and  $S_n^2 := \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2$ .

(a) [6 pt / 6 pts] Use Chebyshev's inequality to prove the WLLN, i.e., that  $\bar{X}_n \xrightarrow{p} \mu$ .

$$\forall \epsilon > 0, \quad \lim_{n \to \infty} \mathbb{P}\left(|\bar{X}_n - \mu| \ge \epsilon\right) \le \lim_{n \to \infty} \frac{\mathbb{V}\mathrm{ar}\left[\bar{X}_n\right]}{\epsilon} = \lim_{n \to \infty} \frac{\sigma^2/n}{\epsilon} = \frac{\sigma^2}{\epsilon} \lim_{n \to \infty} \frac{1}{n} = 0 \quad \checkmark$$

(b) [6 pt / 12 pts] Assume n is large. Plot the approximate PDF of  $\bar{X}_n$  denoted  $f_{\bar{X}_n}(\bar{x})$ . Label the horizontal and vertical axes. Indicate critical value(s) on the axes.

By the WLLN,  $\bar{X}_n \xrightarrow{p} \mu$ , which means that as n gets larger, all the probability "piles up" near  $\mu$  and since we assumed  $\mu > 0$ , we should get something like this:



(c) [8 pt / 20 pts] Using the fact that  $S_n^2 \stackrel{p}{\to} \sigma^2$  and the theorems we learned about in class, show the following fact about the ratio Student was investigating at the turn of the century:

$$\frac{\bar{X}_n - \mu}{\frac{S}{\sqrt{n}}} \stackrel{d}{\to} \mathcal{N}(0, 1)$$

Using the fact given above,  $S \xrightarrow{p} \sigma$  by the CMT where  $g(t) = \sqrt{t}$ . Then,  $\frac{\sigma}{S} \xrightarrow{p} 1$  by the CMT where  $g(t) = t/\sigma$ .

$$\frac{\bar{X}_n - \mu}{\frac{S}{\sqrt{n}}} = \underbrace{\frac{\sigma}{S}}_{\text{see above}} \underbrace{\frac{\bar{X}_n - \mu}{\frac{\sigma}{\sqrt{n}}}}_{\text{by the CLT this term } \stackrel{d}{\to} \mathcal{N}(0, 1)}_{\text{by the CLT this term } \stackrel{d}{\to} \mathcal{N}(0, 1)}$$

For the remainder of the questions in this problem, assume that

$$X_1, \ldots, X_n \stackrel{iid}{\sim} \mathcal{N}\left(\mu, \sigma^2\right)$$

(d) [3 pt / 23 pts] Let  $Z_i := (X_i - \mu)/\sigma$ . How are following distributed?

$$Z_1, \ldots, Z_n \stackrel{iid}{\sim} \mathcal{N}(0, 1)$$

(e) [4 pt / 27 pts] The following has a brand name distribution. Find this distribution and its parameter(s).

The ratio of two independent standard normals was proved in class to be a Cauchy(0,1). Below is just a linear transformation yielding

$$17 + 37 \frac{Z_7}{Z_{11}} \sim \text{Cauchy} (17, 37)$$

(f) [4 pt / 31 pts] The following has a brand name distribution. Find this distribution and its parameter(s).

$$\frac{\bar{X}_n - \mu}{\frac{\frac{\sigma}{\sqrt{n}}}{\sqrt{\frac{\frac{n-1}{\sigma^2}S_n^2}{n-1}}}} = \frac{\bar{X}_n - \mu}{\frac{S_n}{\sqrt{n}}} \sim T_{n-1}$$

Where the equality follows from algebraic simplification.

(g) [7 pt / 38 pts] The following has a brand name distribution. Find this distribution and its parameter(s).

$$\frac{\frac{n-1}{\sigma^2}S_n^2}{\left(\frac{\bar{X}_n - \mu}{\frac{\sigma}{\sqrt{n}}}\right)^2} =: R = \frac{U}{V^2}$$

We know from class that  $U \sim \chi_{n-1}^2 = \operatorname{Gamma}\left(\frac{n-1}{2}, \frac{1}{2}\right)$  and  $V \sim \mathcal{N}\left(0, 1\right)$  which imples  $V^2 \sim \chi_1^2 = \operatorname{Gamma}\left(\frac{1}{2}, \frac{1}{2}\right)$ . By Cochran's theorem, U and V are independent. Thus, R is a ratio of two independent gamma rv's that share their  $\beta$  parameter value. We proved in class that such a ratio is beta prime with first parameter given by the first parameter of the denominator,

$$R \sim \text{BetaPrime}\left(\frac{n-1}{2}, \frac{1}{2}\right)$$

(h) [7 pt / 45 pts] The following expression is distributed as a brand name variable. Find the brand name variable and its parameter values:

$$n\bar{X}_n^2 - 2n\mu\bar{X}_n + n\mu^2 + nS_n^2 - S_n^2 = n\left(\bar{X}_n - \mu\right)^2 + (n-1)S_n^2$$

$$= \frac{\sigma^2}{\sigma^2} \left(n\left(\bar{X}_n - \mu\right)^2 + (n-1)S_n^2\right)$$

$$= \sigma^2 \left(\left(\frac{\bar{X}_n - \mu}{\frac{\sigma}{\sqrt{n}}}\right)^2 + \frac{n-1}{\sigma^2}S_n^2\right)$$

$$\sim \operatorname{Gamma}\left(\frac{n}{2}, \frac{1}{2\sigma^2}\right)$$

Let's examine the term inside the parentheses on the third line. By Cochran's theorem, the first term is  $\chi_1^2$  and the second term is  $\chi_{n-1}^2$  and both terms are independent. Thus the entire parenthesis is distributed as  $\chi_n^2 = \operatorname{Gamma}\left(\frac{n}{2}, \frac{1}{2}\right)$ . Hence we are scaling a gamma by  $\sigma^2$  which yields another gamma distribution with its  $\beta$  parameter divided by the scaling value.

## Problem 2 Some theoretical problems.

(a) [7 pt / 52 pts] Let  $X \sim \text{Weibull}(k, \lambda)$ . Find the most succinct expression as possible for  $\mathbb{E}[X]$ . I have started you off below. Hint: the solution contains a gamma function.

$$\mathbb{E}\left[X\right] = \int_{\mathbb{R}} x f_X(x) dx = \int_{\mathbb{R}} x \left(k\lambda(\lambda x)^{k-1} e^{-(\lambda x)^k}\right) \mathbb{1}_{x>0} dx = k\lambda^k \int_0^\infty x^k e^{-\lambda^k x^k} dx$$

Let  $u = x^k$  thus  $x = u^{1/k}$ ,  $du/dx = kx^{k-1}$ ,  $dx = 1/(kx^{k-1})du$  and the bounds of the integration do not change. Using this substitution,

$$\mathbb{E}\left[X\right] = k\lambda^k \int_0^\infty x^k e^{-\lambda^k u} 1/(kx^{k-1}) du = \lambda^k \int_0^\infty x e^{-\lambda^k u} du = \lambda^k \int_0^\infty u^{1/k} e^{-\lambda^k u} du$$

Since  $u^{1/k} = u^{1/k+1-1}$ , we have a gamma-like integral with  $\alpha = 1/k + 1$  and  $c = \lambda^k$ :

$$\mathbb{E}\left[X\right] = \lambda^k \int_0^\infty u^{\alpha - 1} e^{-cu} du = \lambda^k \frac{\Gamma(\alpha)}{c^{\alpha}} = \lambda^k \frac{\Gamma(1/k + 1)}{(\lambda^k)^{1/k + 1}} = \lambda^k \frac{\Gamma(1/k + 1)}{\lambda^{1 + k}} = \frac{1}{\lambda} \Gamma(1/k + 1)$$

(b) [7 pt / 59 pts] Let  $X \sim \operatorname{ParetoI}(k, \lambda)$ . Find the distribution of  $Y = X \mid X > c$  where c > k. If it's a brand-name rv, mark it as its brand name and find its parameters.

The trick here is to use the survival function of the ParetoI,  $S_X(x) := 1 - F_X(x) = \left(\frac{k}{x}\right)^{\lambda}$ 

$$S_{Y}(y) := \mathbb{P}(Y > y) = \mathbb{P}(X > y \mid X > c) = \frac{\mathbb{P}(X > y, X > c)}{\mathbb{P}(X > c)} = \frac{\mathbb{P}(X > y)}{\mathbb{P}(X > c)}$$
$$= \frac{S_{X}(y)}{S_{X}(c)} = \frac{\left(\frac{k}{y}\right)^{\lambda}}{\left(\frac{k}{c}\right)^{\lambda}} = \left(\frac{c}{y}\right)^{\lambda} \Rightarrow Y \sim \text{ParetoI}(c, \lambda)$$

Survival functions characterize a distribution (just like CDF's, PDF's, PMF's, chf's).

(c) [3 pt / 62 pts] Let  $X \sim \text{Mult}_3\left(9, \frac{1}{3}\mathbf{1}_3\right)$ . Find the distribution of  $Y = \mathbf{1}_k^{\top}X$  where k is the appropriate dimension to make the vector operation legal.

$$Y = \mathbf{1}_3^{\top} \mathbf{X} = X_1 + X_2 + X_3 \sim \text{Deg}(9)$$

(d) [5 pt / 67 pts] Let  $X \sim \text{Mult}_3(9, \frac{1}{3}\mathbf{1}_3)$ . Find  $\mathbb{V}\text{ar}[X]$ . Simplify your answer so that it is a function of  $I_3$ , the identity matrix and  $J_3$ , the matrix of all ones.

Note:  $p = \frac{1}{3}\mathbf{1}_3$  which means  $p_1 = p_2 = p_3 = \frac{1}{3}$  thus we just plug in these values to the formula from class and simplify:

$$\mathbb{V}\mathrm{ar}\left[\boldsymbol{X}\right] = n \begin{bmatrix} p_1(1-p_1) & -p_1p_2 & -p_1p_3 \\ -p_2p_1 & p_2(1-p_2) & -p_2p_3 \\ -p_3p_1 & -p_3p_2 & p_3(1-p_3) \end{bmatrix} = \frac{9}{9} \begin{bmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{bmatrix} = 3\boldsymbol{I}_3 - \boldsymbol{J}_3$$

(e) [7 pt / 74 pts] If  $X \sim \text{Laplace}(0,1)$  and  $Y = e^X$ . Find  $f_Y(y)$  in simplest form.

$$X \sim \text{Laplace}(0,1) := \frac{1}{2} e^{-|x|} \mathbb{1}_{x \in \mathbb{R}}, \quad X = \ln(Y) = g^{-1}(Y) \quad \Rightarrow \quad \frac{d}{dy} \left[ g^{-1}(y) \right] = \frac{1}{y}$$

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{d}{dy} \left[ g^{-1}(y) \right] \right| = \frac{1}{2} e^{-|(\ln(y))|} \mathbb{1}_{\ln(y) \in \mathbb{R}} \frac{1}{|y|} = \frac{1}{2y} e^{-|\ln(y)|} \mathbb{1}_{y \in (0,\infty)}$$

$$= \frac{1}{2} \begin{cases} y^{-2} & \text{if } y \geq 1 \\ 1 & \text{if } y \in (0,1) \\ 0 & \text{if } y \leq 0 \end{cases}$$

(The piecewise function notation is the simplest form but it is not needed for full credit).

(f) [4 pt / 78 pts] If  $X \sim F_{2\alpha,2\beta}$ , find  $k_X(x)$  in simplest form.

$$f_X(x) \propto x^{\frac{2\alpha}{2}-1} \left(1 + \frac{2\alpha}{2\beta}x\right)^{-\frac{2\alpha+2\beta}{2}} \mathbb{1}_{x>0} = x^{\alpha-1} \left(1 + \frac{\alpha}{\beta}x\right)^{-(\alpha+\beta)} \mathbb{1}_{x>0} = k_X(x)$$

(g) [7 pt / 85 pts] If  $X \sim F_{2\alpha,2\beta}$ , find the distribution of  $Y = \frac{\alpha}{\beta}X$ . If it is a brand-name distribution, indicate which one and the values of its parameter(s). Hint: use kernels. Because the parameter space of the F distribution is two positive parameters, we know  $\alpha > 0$  and  $\beta > 0$ . We make use of  $k_X(x)$  from the previous problem

$$f_{Y}(y) = \frac{1}{\frac{\alpha}{\beta}} f_{X} \left( \frac{1}{\frac{\alpha}{\beta}} y \right) \propto f_{X} \left( \frac{\beta}{\alpha} y \right) \propto k_{X} \left( \frac{\beta}{\alpha} y \right) = \left( \frac{\beta}{\alpha} y \right)^{\alpha - 1} \left( 1 + \frac{\alpha}{\beta} \left( \frac{\beta}{\alpha} y \right) \right)^{-(\alpha + \beta)} \mathbb{1}_{\left( \frac{\beta}{\alpha} y \right) > 0}$$

$$\propto y^{\alpha - 1} (1 + y)^{-(\alpha + \beta)} \mathbb{1}_{y > 0} = \frac{y^{\alpha - 1}}{(1 + y)^{\alpha + \beta}} \mathbb{1}_{y > 0} \propto \text{BetaPrime}(\alpha, \beta)$$

(h) [5 pt / 90 pts] If  $X \sim \text{Bernoulli}(p) := p^x (1-p)^{1-x} \mathbb{1}_{x \in \{0,1\}}$ , find the PMF of  $Y = e^X$  in simplest form. Your answer should contain a  $\mathbb{1}_{y \in \mathbb{S}_Y}$  term where  $\mathbb{S}_Y$  is an explicit set.

$$\begin{array}{rcl} X & = & \ln{(Y)} = g^{-1}(Y) \\ p_Y(y) & = & p_X(g^{-1}(y)) \\ & = & p^{\ln(y)}(1-p)^{1-\ln(y)}\mathbbm{1}_{\ln(y) \in \{0,1\}} \\ & = & p^{\ln(y)}(1-p)^{1-\ln(y)}\mathbbm{1}_{y \in \{1,e\}} \end{array}$$

- (i) [3 pt / 93 pts] If  $X \sim \text{ExtNegBin}(k, p)$  where k > 0 and  $p \in (0, 1)$ , what is  $S_X$ ?  $N_0$
- (j) [7 pt / 100 pts] Each of the following distributions will be either "waiting time distributions", "error distributions" or neither (but not both). Underline those that are waiting time distributions. Draw a rectangle box around all the that are error distributions. Do not mark those that are neither waiting nor nor error distributions.