

## Probability and Statistics Preliminary Examination: May 2023

### Instructions:

- Work all 5 Problems. Neither calculators nor electronic devices of any kind are allowed. Clearly state any theorem or fact that you use. Each of the 5 Problems is equally weighted (but the parts within a Problem may not be).
- Abbreviations/Acronyms.
  - pmf (probability mass function); pdf (probability density function); cdf (cumulative distribution function); mgf (moment generating function); iid (independent and identically distributed).
  - MSE (mean squared error), MOME (method of moments estimator); MLE (maximum likelihood estimator); PBE (posterior Bayes estimator); UMVUE (uniform minimum variance unbiased estimator); UMP (uniformly most powerful); LRT (likelihood ratio test).
- Notation.
  - $I(x \in A)$  or  $I_A(x)$ : indicator function for set  $A$ ; takes on the value 1 if  $x \in A$  and 0 otherwise.
  - $\mathbb{E}(X)$ : expectation of random variable  $X$ .
  - $\mathbb{V}(X)$ : variance of random variable  $X$ .
  - $X \sim N(a, b)$ :  $X$  has a normal distribution with mean  $a$  and variance  $b$ .
- Common distributions and other results.

**Bin**( $n, p$ ):  $\mathbb{E}(X) = np$ ,  $\mathbb{V}(X) = np(1 - p)$ , and pmf and mgf given respectively by

$$f(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x} I(x \in \{0, 1, \dots, n\}), \quad M(t) = (1-p + pe^t)^n$$

**Poisson**( $\lambda$ ):  $\mathbb{E}(X) = \lambda$ ,  $\mathbb{V}(X) = \lambda$ , and pmf and mgf given respectively by

$$f(x) = \frac{e^{-\lambda} \lambda^x}{x!} I(x \in \{0, 1, \dots\}), \quad M(t) = \exp\{\lambda(e^t - 1)\}$$

**Geometric**( $p$ ):  $\mathbb{E}(X) = 1/p$ ,  $\mathbb{V}(X) = (1-p)/p^2$ , and pmf and mgf given respectively by

$$f(x) = p(1-p)^{x-1} I(x \in \{1, 2, \dots\}), \quad M(t) = \frac{pe^t}{1 - (1-p)e^t}, \quad t < -\log(1-p)$$

**Negative-Binomial**( $r, p$ ):  $\mathbb{E}(X) = r(1-p)/p$ ,  $\mathbb{V}(X) = r(1-p)/p^2$ , and pmf and mgf, respectively:

$$f(x) = \binom{r+x-1}{x} p^r (1-p)^x I(x \in \{0, 1, \dots\}), \quad M(t) = \left( \frac{p}{1 - (1-p)e^t} \right)^r, \quad t < -\log(1-p)$$

**Gamma**( $\alpha, \beta$ ):  $\mathbb{E}(X) = \alpha\beta$ ,  $\mathbb{V}(X) = \alpha\beta^2$ , and pdf and mgf given respectively by

$$f(x) = \frac{1}{\Gamma(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta} I(x > 0), \quad M(t) = \left( \frac{1}{1 - \beta t} \right)^\alpha, \quad t < 1/\beta$$

**Exp**( $\beta$ ): Gamma(1,  $\beta$ ).

**Beta**( $\alpha, \beta$ ):  $\mathbb{E}(X) = \frac{\alpha}{\alpha + \beta}$ ,  $\mathbb{V}(X) = \frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}$ , and pdf given by

$$f(x) = \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} I(0 < x < 1)$$

**Order Statistics:** Let  $X_{(1)} \leq \dots \leq X_{(n)}$  denote the order statistics from a random sample  $X_1, \dots, X_n$ . If  $X_1$  is continuous with pdf  $f(x)$  and cdf  $F(x)$ , the pdf of  $X_{(j)}$ ,  $(X_{(i)}, X_{(j)})$ , and  $(X_{(1)}, \dots, X_{(n)})$ , are given by:

$$f_{X_{(j)}}(x) = \frac{n!}{(j-1)!(n-j)!} [F(x)]^{j-1} [1-F(x)]^{n-j} f(x) I(-\infty < x < \infty)$$

$$f_{X_{(i)}, X_{(j)}}(x_i, x_j) = \frac{n!}{(i-1)!(j-1-i)!(n-j)!} f(x_i) f(x_j) [F(x_i)]^{i-1} [F(x_j) - F(x_i)]^{j-1-i} [1-F(x_j)]^{n-j} \\ \times I(-\infty < x_i \leq x_j < \infty)$$

$$f_{X_{(1)}, \dots, X_{(n)}}(x_1, \dots, x_n) = n! f(x_1) \cdots f(x_n) I(-\infty < x_1 \leq \dots \leq x_n < \infty)$$

## Problems

1. Let the joint pdf of  $X$  and  $Y$  be

$$f(x, y) = \begin{cases} 2 & \text{if } x > 0, y > 0, x + y < 1, \\ 0 & \text{elsewhere.} \end{cases}$$

Find the pdf of  $X/Y$ .

2. Suppose that  $X_1, \dots, X_m$  has a multinomial distribution with pmf given by

$$f(x_1, \dots, x_m | p_1, \dots, p_m) = \frac{n!}{x_1! \dots x_m!} p_1^{x_1} \dots p_m^{x_m}.$$

Here  $0 \leq x_i \leq n$ ,  $\sum_{i=1}^m x_i = n$ , and  $\sum_{i=1}^m p_i = 1$ .

- (a) Find a likelihood ratio test (LRT) statistic for the hypotheses  $H_0 : p_1 = p_2 = \dots = p_m = \frac{1}{m}$  V.S.  $H_1 : H_0$  is false.
- (b) Find the asymptotic size  $\alpha$  rejection region based on the LRT statistic in (a).

3. Suppose  $X_1, \dots, X_n$  are i.i.d with uniform distribution  $Unif(-\theta, \theta)$  with  $\theta > 0$ .
- (a) Find the MLE  $\hat{\theta}$  of  $\theta$ .
  - (b) Find the distribution of  $\hat{\theta}/\theta$ .
  - (c) Using the exact distribution in (2), find the value of  $m$  such that  $[\hat{\theta}, m\hat{\theta}]$  is a  $1 - \alpha$  confidence interval for  $\theta$ .

4. Let  $X_1, \dots, X_n$  be independent random variables from  $Bin(1, \theta)$  with  $\theta \in (0, 1)$ . Let  $T = \sum_{i=1}^n X_i$ .
- Find the conditional pmf of  $X_i$  given  $T = t$ .
  - Find the UMVUE of  $\theta(1 - \theta)$ .

5. Let  $X_1, \dots, X_n$  be i.i.d observations from  $Pois(\lambda)$  with  $\lambda > 0$ . Let us define

$$T = \sum_{i=1}^n X_i, \quad W = (1 - 1/n)^T, \quad \tau(\lambda) = e^{-\lambda}.$$

- (a) Show that  $W$  is an unbiased estimator of  $\tau(\lambda)$ .
- (b) Calculate the variance of  $W$ .
- (c) Is  $W$  a UMVUE? Prove or disprove.