

## Probability and Statistics Preliminary Examination: August 2024

### 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:

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

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

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

1. Suppose that  $X$  has a pdf  $f_X(x) = 1$  with  $x \in (0, 1)$ , and  $Y$  has a pdf  $f_Y(y) = e^{-y}$  with  $y > 0$ . Moreover,  $X$  and  $Y$  are independent. Let  $W = \min(X, Y)$ . Find  $\mathbb{P}(W < t | W = X)$  for all  $t \in \mathbb{R}$ .

2. Let  $X_1, \dots, X_n$  be independent random variables with a common pmf  $f(x|N) = 1/N$  with  $x = 1, \dots, N$ . Here  $N \in \{1, 2, \dots\}$  is the parameter.
- Find a minimal sufficient statistic for  $N$ .
  - Find the UMP test for  $H_0 : N = N_0$  v.s.  $H_1 : N > N_0$ . Here  $N_0 \geq 1$  is a known integer.

3. Let  $X_1, \dots, X_n$  be independent random variables from  $Bin(1, \theta)$  with  $\theta \in (0, 1)$ . Let  $\tau(\theta) = \theta(1 - \theta)$  and  $\bar{X}$  be the sample average.
- (a) Show that  $W = \frac{n}{n-1} \bar{X}(1 - \bar{X})$  is the UMVUE FOR  $\tau(\theta)$ .
- (b) When  $\theta \neq 1/2$ , find the limiting distribution of  $\sqrt{n}(W - \tau(\theta)) \xrightarrow{D} ?$

4. Let  $X_1, \dots, X_n$  be independent random variables from the common pdf  $f(x|\theta) = \theta^3 x^2 e^{-\theta x}/2$  for  $0 < x < \infty, 0 < \theta < \infty$ . Let  $\hat{\theta} = 3n/\sum_{i=1}^n X_i$ .
- (a) Find  $\mathbb{E}(\hat{\theta})$ .
  - (b) Find the limiting distribution  $\sqrt{n}(\hat{\theta} - \theta) \xrightarrow{D} ?$
  - (c) Based on the result in (b), find an asymptotic  $1 - \alpha$  confidence interval for  $\theta$ .

5. Let  $X_1, \dots, X_n$  be i.i.d observations from the pdf  $f(x|\theta) = \frac{1}{2}I(\theta - 1 \leq x \leq \theta + 1)$ .
- (a) Show that  $X_{(n)} - \theta$  is a pivotal quantity.
  - (b) Find  $a$  and  $b$  such that the interval  $[X_{(n)} - a, X_{(n)} - b]$  is the shortest  $1 - \alpha$  confidence interval.