

Probability and Statistics Preliminary Examination: May 2015

Instructions:

- Work all 6 Problems. Neither calculators nor electronic devices of any kind are allowed. Clearly state any theorem or fact that you use. Each of the 6 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.

Poisson(λ): $\mathbb{E}(X) = \lambda$, $\mathbb{V}(X) = \lambda$, 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(α, β): $\mathbb{E}(X) = \alpha\beta$, $\mathbb{V}(X) = \alpha\beta^2$, and pdf and mgf given respectively by

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

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. Recall that there are 52 cards in a standard deck, with 4 cards for each denomination. Suppose that you flip over the top 4 cards. This constitutes the first draw from the deck. Discard those cards that are not a 2 and redraw new cards from the remaining deck so that you again have 4 cards flipped over. Continue doing this until all 4 cards are a 2. Let X denote the number of draws made and Y_j denote the number of cards drawn on draw j .

- (a) Calculate $P(X = 3)$.
- (b) Derive the pmf of Y_2 .
- (c) Derive the pmf of $Y_2|X \geq 3$.

2. Let X be a random variable from a *contaminated normal* distribution. That is, let $B \sim \text{Bernoulli}(p)$. Then $X|B = 0 \sim N(\mu, \tau^2)$ and $X|B = 1 \sim N(\mu, \sigma^2)$.

- (a) Derive the pdf for X
- (b) Calculate $\mathbb{E}(X)$ and $\mathbb{V}(X)$.
- (c) Derive an expression for $P(B = 1|X = x)$.
- (d) Calculate $\text{Cov}(X, B)$. Are X and B independent? Justify your answer.

3. Let X_1, \dots, X_n ($n > 4$) be independent random variables such that $X_i \sim N(i, i)$ for $i = 1, \dots, n$. Let $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$ be the sample mean.

- (a) Calculate $\mathbb{E}(\bar{X})$ and $\mathbb{V}(\bar{X})$. Describe the behavior of both of these quantities as $n \rightarrow \infty$.
- (b) Determine the distribution of

$$\frac{2(X_1 + X_2 - X_3)}{\sqrt{6}(X_4 - 4)}$$

- (c) Calculate the probability that X_n is not the largest observation in the sample. This probability should be expressed in terms of $\Phi(\cdot)$, the cdf of the standard normal distribution.

4. Let $(X_1, Y_1), \dots, (X_n, Y_n)$ be iid random vectors from the pdf:

$$f_{X,Y}(x, y) = \frac{1}{2\pi\sqrt{1-\rho^2}} \exp\left(\frac{-1}{2(1-\rho^2)}(x^2 - 2\rho xy + y^2)\right)$$

- (a) Find a minimal sufficient statistic for ρ .
 (b) Show that the MLE may be found as a solution to the following equation:

$$0 = \rho^3 - \rho^2 \frac{\sum_{i=1}^n x_i y_i}{n} - \rho \frac{n - \sum_{i=1}^n x_i^2 - \sum_{i=1}^n y_i^2}{n} - \frac{\sum_{i=1}^n x_i y_i}{n}$$

- (c) Is $(\sum_{i=1}^n X_i, \sum_{i=1}^n Y_i)$ a sufficient statistic for ρ ? Why or why not?
 (d) Are \bar{X} and \bar{Y} (separately) ancillary statistics for ρ ? Why or why not?

5. Let X_1, \dots, X_n ($n > 2$) be a random sample from a distribution with $\mathbb{E}(X_1) = \mu < \infty$ and $\mathbb{V}(X_1) = \sigma^2 < \infty$. Define the following estimators for μ .

$$\bar{X}_t = \frac{\sum_{i=2}^{n-1} X_{(i)}}{n-2}, \quad \tilde{\mu}_n = \frac{1}{a} X_1 + \frac{c}{n-2} \sum_{i=2}^{n-1} X_i + \frac{1}{a} X_n,$$

where $a > 0$ and c are constants.

- (a) Calculate $\mathbb{E}(\bar{X}_t)$.
 (b) Under what conditions is \bar{X}_t an unbiased estimator of μ ?
 (c) Determine the value of c for which $\tilde{\mu}_n$ is an unbiased estimator of μ .
 (d) If $a = 2$, determine whether $\tilde{\mu}_n$ is a consistent estimator for μ .
 (e) Find the value of a that minimizes $\mathbb{V}(\tilde{\mu}_n)$.

6. Let X_1, X_2 , and X_3 be random variables from the following joint pmf:

$$f_{X_1, X_2, X_3}(x_1, x_2, x_3) = \frac{n!}{x_1! x_2! x_3!} p_1^{x_1} p_2^{x_2} p_3^{x_3},$$

with $0 \leq x_i \leq n$ for $i = 1, 2, 3$, $\sum_{i=1}^3 x_i = n$, and $0 \leq p_i \leq 1$ for $i = 1, 3$, and $\sum_{i=1}^3 p_i = 1$.

- (a) Find the MLEs for p_1 and p_2 .
 (b) Give the test statistic and rejection region for an asymptotic level α LRT of $H_0 : p_1 = p_2$ vs. $H_1 : p_1 \neq p_2$.
 (c) Conditional on $X_3 = x_3$, construct an asymptotic level $100(1-\alpha)\%$ confidence interval for $\frac{p_1}{p_1+p_2}$.