Probability and Statistics Preliminary Examination: May 2017

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.
 - E(X): expectation of random variable X.
 - Var(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

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

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

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

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

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

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

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

Order Statistics: Let $X_{(1)} \leq \cdots \leq X_{(n)}$ denote the order statistics from a random sample X_1, \ldots, 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 $(X_{(1)}, \ldots, X_{(n)})$, are given by:

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

$$f_{X_{(i)},X_{(j)}}(x_i,x_j) = \frac{n!}{(i-1)!(j-1-i)!(n-j)!} f(x_i)f(x_j)[F(x_i)]^{i-1}[F(x_j)-F(x_i)]^{j-1-i}[1-F(x_j)]^{n-j} \\ \times I(-\infty < x_i \le x_j < \infty) \\ f_{X_{(1)},\dots,X_{(n)}}(x_1,\dots,x_n) = n!f(x_1)\cdots f(x_n)I(-\infty < x_1 \le \dots \le x_n < \infty)$$

- 1. An urn contains N balls numbered 1, 2, ..., N. From these balls n balls are drawn at random without replacement. Let X_i denote the number that appears on the *i*th sampled ball and $T = \sum_{i=1}^{n} X_i$. Find E(T) and Var(T).
- 2. Let the random variables $X_1, X_2 \stackrel{iid}{\sim} Exponential(1)$. Define $U = X_1 + X_2$ and $V = X_1 X_2$.
 - (a) Find the marginal distributions of U and V.
 - (b) Obtain P(U > 1, V > 1)
 - (c) Find the correlation between U and V.
 - (d) Are the random variables U and V indpendent? Provide your reasoning.
- 3. Suppose X is a non-negative random variable for which, for some positive integer, r, $E(X^r)$ does not exist. Obtain the necessary and sufficient condition for the moment generating function of X to exist.
- 4. Consider the model $Y_i = x_i\beta + \epsilon_i$ i = 1, 2, ..., n, where β is an unknown parameter to be estimated. ϵ_i are iid random variables from a distribution whose moments all exist, in particular $E(\epsilon_i) = 0$, $\forall i$ and $Var(\epsilon_i) = \sigma^2 \quad \forall i$. Then an estimate of β is given by

$$\hat{\beta_n} = \frac{\sum_{i=1}^n x_i^2}{\sum_{i=1}^n x_i y_i}$$

If $\lim_{n\to\infty} \frac{1}{n} \sum_{i=1}^{n} x_i^2 = S_{xx}$ (a positive constant), then obtain the limiting distribution of $\hat{\beta}_n$.

- 5. Consider the following Poisson-Gamma model: $X_1, X_2, ..., X_n | \lambda \stackrel{iid}{\sim} Poisson(\lambda)$. The prior, $\pi(\lambda) \propto \lambda^{a-1}e^{-b\lambda}$, where a and b are known positive constants.
 - (a) Obtain the mode¹ of the posterior distribution of λ .
 - (b) Using the posterior mode obtained in part (a), show that, as $n \to \infty$, the posterior distribution of λ can be approximated by a *Normal* distribution. Clearly state the assumptions you need to make.
 - (c) Obtain the mean and variance of the Normal distribution you obtained in part (b).
- 6. Let $X_1, X_2, ..., X_n \stackrel{iid}{\sim} Uniform(0, \theta), \theta > 0$. We wish to test $H_0: \theta \leq \theta_0$ Vs. $H_1: \theta > \theta_0$.
 - (a) Find an UMP level α test and derive the power function for this.
 - (b) Consider another decision rule given by

Reject H_0 with probability 1 if $x_{(n)} > \theta_0$ Reject H_0 with probability α if $x_{(n)} < \theta_0$

Would you prefer this test to the UMP test obtained in part (a)? Why, why not?

¹for a continuous r.v, X, with pdf $f_X(x)$, mode is defined as argmax $f_X(x)$