## Statistics and Probability Prelim Exam

Tuesday, August 21, 2018 12pm - 4 pm

Note: To pass this exam, you need to pass both Statistics and Probability parts Statistics Part

## **Statistics Part**

1. Suppose that  $\{X_1, ..., X_n\}$  is a random sample from an exponential distribution with an unknown parameter  $\beta > 0$  and pdf given by

$$f(x|\beta) = \frac{1}{\beta} \exp\{-x/\beta\}; \ x > 0,$$

and that  $\{Y_1, ..., Y_n\}$  is random sample from an exponential distribution with an unknown parameter  $\delta > 0$ . Assume that the two samples are independent. Let  $\theta = P(X_1 < Y_1)$ . Find the uniformly minimum variance unbiased estimator of  $\theta$  based on the random samples given above when n = 2.

2. Suppose  $X_1, ..., X_n$  are iid from a distribution with pdf

$$f(x|\theta) = \begin{cases} e^{-(x-\theta)} & \text{if } x > \theta\\ 0 & else, \end{cases}$$

where  $\theta > 0$  is unknown. We want to test the null hypothesis  $H_0$ :  $\theta \leq \theta_0$  versus the alternative hypothesis  $H_1: \theta > \theta_0$  for a known value  $\theta_0 > 0$ . Is there a uniformly most powerful (UMP) test? If there is one, find the UMP test and give the critical region of size  $\alpha = 0.05$ . If there is none, explain why.

- 3. Let  $X_1, ..., X_n$  be iid Bernoulli(p),  $0 . Let <math>T_n$  be the uniformly minimum variance unbiased estimator of  $p^2$ . Find  $T_n$  and determine if it is asymptotically normally distributed in the sense that  $\sqrt{n}(T_n \mu)$  converges in distribution to a normal distribution, for some constant  $\mu$ .
- 4. Let  $x_1, ..., x_n$  be a random sample from  $Bernoulli(\theta)$  distribution, where  $0 < \theta < 1$ , and is unknown. Assume a U(0, 1) prior for  $\theta$ . Consider the loss function for estimating  $\theta$  given by

$$L(\theta, a) = \frac{(\theta - a)^2}{(1 - \theta)}$$

Find the Bayes rule with respect to the above loss function.

## **Probability Part**

5. Let X be a non-negative random variable. Prove that

$$\mathbb{E}e^X = 1 + \int_0^\infty e^t \mathbb{P}(X > t) dt.$$

6. Consider a sequence of independent random variables  $\{X_n\}_{n\in\mathbb{N}}$ , each  $X_n$  has the following distribution:

$$\mathbb{P}(X_n = k) = \frac{1}{(n+5)^{k\gamma}}, k = 1, 2, 3,$$

and  $\mathbb{P}(X_n = 0) = 1 - \mathbb{P}(X_n \in \{1, 2, 3\})$ , for some parameter  $\gamma > 0$ . So, each realization of  $\{X_n\}_{n \in \mathbb{N}}$  is an infinite sequence of 0, 1, 2, 3. Find the range of  $\gamma$  so that *both* of the following conditions are satisfied at the same time.

- (a) The number 1 occurs infinitely often with probability one.
- (b) The number 3 occurs only a finite number of times with probability one.

Justify your answer.

7. Suppose that  $X_1, X_2, \ldots$  are independent random variables, and  $X_n$  is uniformly distributed over [0, n] (i.e.,  $\mathbb{P}(X_n \leq x) = x/n, x \in [0, n], n \in \mathbb{N}$ ). Find appropriate sequences  $\{a_n\}_{n \in \mathbb{N}}$  and  $\{b_n\}_{n \in \mathbb{N}}$  so that

$$\frac{\sum_{k=1}^{n} X_k - b_n}{a_n} \Rightarrow \mathcal{N}(0, 1),$$

as  $n \to \infty$ , where  $\mathcal{N}(0, 1)$  is the standard normal distribution. Justify your answer. *Hints:* You may use the following estimates

$$\sum_{k=1}^{n} k = \frac{n(n+1)}{2} \quad \text{and} \quad \sum_{k=1}^{n} k^2 = \frac{n(n+1)(2n+1)}{6}.$$

8. Suppose that for each  $n \in \mathbb{N}$ , random variable  $X_n$  has density  $f_n(x) = 1 + \cos(2\pi nx)$  on [0, 1]. Prove that  $X_n$  converges in distribution to some random variable X as  $n \to \infty$ , and determine the law of X.