## Problem 1

Assume  $X_1, \ldots, X_n$  are iid Poisson distributed with parameter  $\lambda$ , i. e. the pmf for each of the variables is given by:

$$f(x|\lambda) = P(X = x) = \frac{\lambda^{x} e^{-\lambda}}{x!}, x=0,1,2,...$$

where  $\lambda \in (0,\infty)$ .

- a) Show that  $f(x|\lambda)$  is a member of an exponential family of distributions. Derive the maximum likelihood estimator for  $\lambda$ ,  $\hat{\lambda}$ , based on  $X_1, \dots, X_n$ .
- b) Derive the score statistic based on  $X_1, \ldots, X_n$ . Is  $\hat{\lambda}$  a uniform minimum variance unbiased estimator (UMVUE) for  $\lambda$ ? Explain your answer.
- c) It is of interest to estimate the probability  $P(X = 0) = e^{-\lambda}$ . What is the maximum likelihood estimator for the probability P(X = 0)? Find the moment generating function for  $Y = \sum_{i=1}^{n} X_i$  and use it to find the expected value of the maximum likelihood estimator for P(X = 0). Is it unbiased?
- d) Suppose there exist an unbiased estimator  $W(X_1,...,X_n)$  for  $\tau(\lambda) = e^{-\lambda}$ . Derive the Cramer-Rao lower bound of the variance of  $W(X_1,...,X_n)$ . Can  $W(X_1,...,X_n)$  attain this lower bound?
- e) Suppose U and V are independent Poisson distributed with parameters  $\lambda_1$  and  $\lambda_2$  respectively. Let Z = U + V. Show that the conditional distribution of U given Z = z is binomial  $\left(z, \frac{\lambda_1}{\lambda_1 + \lambda_2}\right)$ . What is the conditional distribution of  $X_1$  given that  $Y = \sum_{i=1}^n x_i$ ?

f) Let the estimator  $\hat{\tau}(X_1, \dots, X_n)$  be defined by  $\hat{\tau}(X_1, \dots, X_n) = \begin{cases} 1 & \text{if } X_1 = 0 \\ 0, & \text{otherwise} \end{cases}$ Show that  $\hat{\tau}(X_1, \dots, X_n)$  is an unbiased estimator for  $\tau(\lambda) = e^{-\lambda}$ . Explain also why the estimator  $E\left(\hat{\tau}(X_1, \dots, X_n) \middle| Y = \sum_{i=1}^n X_i\right)$  is a unique best unbiased estimator of  $\tau(\lambda) = e^{-\lambda}$ .

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g) Show that 
$$E\left(\hat{\tau}(X_1, \dots, X_n) \middle| Y = \sum_{i=1}^n X_i\right) = \left(1 - \frac{1}{n}\right)^{\sum_{i=1}^{n} X_i}$$
 and find the variance of this estimator.

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Compare the variance to its lower bound.

## Problem 2

Another way to get information about  $\lambda$  is to register the time between events in a Poisson process. Let  $T_1, T_2, \dots, T_n$  be iid and represent n such times, each with a pdf given by

$$f_{T}(t|\lambda) = \begin{cases} \lambda e^{-\lambda t}, t > 0\\ 0, \text{ otherwise} \end{cases}$$

- a) Show that  $U = \sum_{i=1}^{n} T_i$  is a sufficient statistic for  $\lambda$  and verify if it is minimal or not. What is the distribution of  $U = \sum_{i=1}^{n} T_i$ ?
- b) Show that  $\begin{bmatrix} \chi^2_{2n,1-\frac{\alpha}{2}} & \chi^2_{2n,\frac{\alpha}{2}} \\ \frac{2\sum_{i=1}^n T_i}{2\sum_{i=1}^n T_i} & \frac{2\sum_{i=1}^n T_i}{2\sum_{i=1}^n T_i} \end{bmatrix}$  is a  $1-\alpha$  confidence interval for  $\lambda$ . What is the expected length

of this interval? (Hint. You can use that if V is gamma distributed with parameters  $\alpha$  and  $\beta$ ,

then 
$$E[V^k] = \frac{\Gamma(\alpha+k)\beta^k}{\Gamma(\alpha)}, \ k > -\alpha$$
.)

c) Explain why  $\sqrt{n}\left(\overline{T_n} - \frac{1}{\lambda}\right) \xrightarrow{D} N\left(0, \frac{1}{\lambda^2}\right)$  where  $\overline{T_n} = \frac{1}{n} \sum_{i=1}^n T_i$  and  $\xrightarrow{D}$  means convergence in

distribution. Use this to construct a  $1-\alpha$  confidence interval for  $\lambda$  based on  $\overline{T_n}$ , using its asymptotic distribution. Compare the expected length of this interval to the one in b) when  $\alpha = 0.05$  and n=20. What is the asymptotic distribution of  $\overline{T_n}^{-1}$ ?