## STA261 S19: Test 2 Solutions

No aids. 60 minutes. Write all answers directly beneath where the question is asked. Use pages 4 and 5 for rough work.

- 1. Basic, 4 marks Let  $X_1, \ldots, X_n$  be an IID sample from a parametric family of distributions  $\{F_{\theta} : \theta \in \Theta\}$  with corresponding densities  $f_{\theta}$ .
- a) (2) Define the <u>Likelihood function</u> in <u>words</u> and <u>very briefly explain</u> how it is used for inference on  $\theta$ .

Solution. The Likelihood function evaluated at  $\theta$  is the probability under our model of observing the data given that the true parameter is  $\theta$  (1 mark). We can use it to get a region of  $\theta$  values that correspond to a high probability under our model of observing the data that we did (1 mark).

**b)** (2) Give a mathematical definition of the following in terms of  $f_{\theta}$  (.5 each):

## Likelihood function:

Solution.

$$L(\theta \mid X_1, \dots, X_n) = \prod_{i=1}^n f_{\theta}(X_i)$$
 (0.5 marks)

Score function:

Solution.

$$S(\theta \mid X_1, \dots, X_n) = \frac{\partial \log L(\theta \mid X_1, \dots, X_n)}{\partial \theta} = \frac{\partial \sum_{i=1}^n \log f_{\theta}(X_i)}{\partial \theta}$$
(0.5 marks)

Observed Information:

Solution.

$$\widehat{I}(X_1, \dots, X_n) = -\frac{\partial^2 \log L(\theta \mid X_1, \dots, X_n)}{\partial \theta^2} \bigg|_{\theta = \theta_{MLE}} = -\frac{\partial^2 \sum_{i=1}^n \log f_{\theta}(X_i)}{\partial \theta^2} \bigg|_{\theta = \theta_{MLE}} \tag{0.5 marks}$$

Expected/Fisher Information:

Solution.

$$I(\theta) = \operatorname{Var}_{\theta} \left( S(\theta \mid X_1, \dots, X_n) \right) = \operatorname{Var}_{\theta} \left( \frac{\partial \sum_{i=1}^n \log f_{\theta}(X_i)}{\partial \theta} \right)$$
 (0.5 marks)

Also acceptable: negated expected second derivative of log likelihood.

- **2.** Adept, 4 marks. Suppose  $X_1, \ldots, X_n$  is an IID sample from a Poisson( $\lambda$ ) distribution, having  $P(X = x) = \lambda^x e^{-\lambda}/x!$  for  $x = 0, 1, 2, \ldots$
- a) (1) Find the Likelihood function for  $\lambda$ .

Solution.

$$L(\lambda \mid X_1, \dots, X_n) = \prod_{i=1}^n P_{\lambda}(X_i)$$

$$= \prod_{i=1}^n \frac{\lambda^{X_i} e^{-\lambda}}{X_i!}$$

$$= \frac{\lambda^{\sum_{i=1}^n X_i} e^{-n\lambda}}{\prod_{i=1}^n X_i!}.$$
(1 mark)

**b)** (1) Show that  $T = \sum_{i=1}^{n} X_i$  is a <u>sufficient statistic</u> for  $\lambda$ . Briefly explain why this means that the sample mean  $\bar{X} = T/n$  is also sufficient.

Solution. Consider two different IID samples  $X_1, \ldots, X_n$  and  $Y_1, \ldots, Y_n$  from the same Poisson model that satisfy  $T = \sum_{i=1}^n X_i = \sum_{i=1}^n Y_i$  (0.25 marks). Then, using 2.a),

$$\frac{L(\lambda \mid X_{1}, \dots, X_{n})}{L(\lambda \mid Y_{1}, \dots, Y_{n})} = \frac{\lambda^{\sum_{i=1}^{n} X_{i}} e^{-n\lambda}}{\prod_{i=1}^{n} X_{i}!} \frac{\prod_{i=1}^{n} Y_{i}!}{\lambda^{\sum_{i=1}^{n} Y_{i}} e^{-n\lambda}}$$

$$= \frac{\lambda^{T} e^{-n\lambda}}{\lambda^{T} e^{-n\lambda}} \frac{\prod_{i=1}^{n} Y_{i}!}{\prod_{i=1}^{n} X_{i}!}$$

$$= \frac{\prod_{i=1}^{n} Y_{i}!}{\prod_{i=1}^{n} X_{i}!}.$$
(0.5 marks)

Since this value is constant with respect to  $\lambda$ , T is a sufficient statistic. If  $\bar{X} = \bar{Y}$ , then clearly  $\sum_{i=1}^{n} X_i = \sum_{i=1}^{n} Y_i$ , so the ratio of the likelihood functions is the same constant (0.25 marks). This implies T/n is also a sufficient statistic.

Alternatively, note that the parts of  $L(\lambda|X_1,\ldots,X_n)$  which contain  $\lambda$  depend on X only through T, and use the factorization theorem.

## c) (1) Find the maximum likelihood estimator for $\lambda$ .

Solution. First, notice that  $\sum_{i=1}^{n} X_i = n\bar{X}$ . Then, the log-likelihood is

$$\ell(\lambda \mid X_1, \dots, X_n) = \log L(\lambda \mid X_1, \dots, X_n)$$

$$= \log \left( \frac{\lambda^{\sum_{i=1}^n X_i} e^{-n\lambda}}{\prod_{i=1}^n X_i!} \right)$$

$$= n\bar{X} \log(\lambda) - n\lambda - \log \left( \prod_{i=1}^n X_i! \right). \tag{0.25 marks}$$

Differentiating, the score function is

$$S(\lambda \mid X_1, \dots, X_n) = \frac{\partial \ell(\lambda \mid X_1, \dots, X_n)}{\partial \lambda} = \frac{n\bar{X}}{\lambda} - n.$$
 (0.25 marks)

The MLE is the value  $\widehat{\lambda}$  such that  $S(\widehat{\lambda} \mid X_1, \dots, X_n) = 0$ . Thus,

$$\frac{n\bar{X}}{\hat{\lambda}} - n = 0 \implies \hat{\lambda} = \bar{X}. \tag{0.5 marks}$$

## d) (1) Find a random interval $(L_n, U_n)$ such that $\mathbb{P}(\lambda \in (L_n, U_n)) \approx 1 - \alpha$ for some $\alpha \in (0, 1)$ .

Solution. Recall that for an MLE under regularity conditions with  $I(\lambda)$  continuous, an approximate  $\alpha$ -confidence interval is given by

$$\widehat{\lambda} \pm \frac{z_{\alpha/2}}{\sqrt{I(\widehat{\lambda})}},$$

where  $z_{\alpha/2}$  is the  $\alpha/2$ -quantile of the standard Normal distribution. (0.5 marks) So, it remains to calculate the observed Fisher information.

$$I(\widehat{\lambda}) = -\frac{\partial^2 \ell(\lambda \mid X_1, \dots, X_n)}{\partial \lambda^2} \bigg|_{\lambda = \widehat{\lambda}} = -\frac{\partial}{\partial \lambda} \left( \frac{n\overline{X}}{\lambda} - n \right) \bigg|_{\lambda = \widehat{\lambda}} = \frac{n\overline{X}}{\widehat{\lambda}^2} = \frac{n}{\overline{X}}.$$

Thus, the random interval desired is

$$\bar{X} \pm z_{\alpha/2} \sqrt{\frac{\bar{X}}{n}}$$
. (0.5 marks)

This is great, it's also acceptable for them to use the CLT for IID sums, since the MLE happens to be the sample mean. I believe this gives the same answer, since the variance of a poisson is the mean.

**3.** Advanced, 2 marks. Suppose I flip a coin 10 times and get 7 heads. I claim that the coin is fair. Find an expression for the approximate probability of seeing data at least this extreme if my claim is true. For full marks, show all work and justify all steps using theorems/results from lecture. The density of a binomial random variable is ...

Solution. First, we need to choose a model. Let's suppose that each coin flip is an IID Bernoulli(p) trial, where p is the probability of a head (denoted by an outcome of 1) (0.5 marks). We wish to find the probability of seeing at least 7 heads when p = 1/2. By the symmetry of this choice, we must also include the probability of seeing 3 or fewer heads (0.5 marks). Let X denote the sum of the individual flips, so that X is in fact the number of heads observed, and recall that then X has a Binomial(10,1/2) distribution. That is, the probability of interest is

$$1 - \sum_{x=4}^{6} P(X = x) = 1 - \sum_{x=4}^{6} {10 \choose x} (1/2)^x (1 - 1/2)^{10-x}$$

$$= 1 - (1/2)^{10} \left[ {10 \choose 4} + {10 \choose 5} + {10 \choose 6} \right]$$

$$= 0.3438.$$
(0.5 marks)

While you are not required to use this terminology in your solution, notice that what you have calculated is a p-value.

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