Statistics 2 - Problem Sheet 5

Dom Hutchinson

November 3, 2019

Question - 1.

Suppose that $\mathbf{X} \stackrel{\text{iid}}{\sim} \text{Poisson}(\theta^*)$. Consider the observed Fisher Infromation evaluated at $\hat{\theta}_n$ and show that $J_n(\hat{\theta}_n) = 1/\bar{x}$. Find an asymptotically exact $1 - \alpha$ confidence interval for θ^* .

Let $\mathbf{X} \stackrel{\text{iid}}{\sim} \operatorname{Poisson}(\theta^*)$ and consider $\hat{\theta}_n = \hat{\theta}_n(\mathbf{X}) := \frac{1}{n} \sum_{i=1}^n X_i$. Then

$$\ell'(\theta; \mathbf{x}) = -n + \frac{1}{\theta} \sum_{i=1}^{n} x_{i}$$

$$\implies ell''(\theta; \mathbf{x}) = -\frac{1}{\theta^{2}} \sum_{i=1}^{n} x_{i}$$
We have
$$I(\theta) = -\mathbb{E}(\ell''(\hat{\theta}; X); \hat{\theta})$$

$$= -\mathbb{E}\left(-\frac{1}{\hat{\theta}^{2}} \sum_{i=1}^{n} X_{i}; \hat{\theta}\right)$$

$$= \frac{1}{\hat{\theta}^{2}} \sum_{i=1}^{n} \mathbb{E}(X_{i}; \hat{\theta})$$

$$= \frac{1}{\hat{\theta}^{2}} n \hat{\theta}$$

$$= \frac{n}{\hat{\theta}}$$

$$= \frac{n}{\hat{\theta}}$$
and
$$J_{n}(\hat{\theta}_{n}) := -\frac{1}{n} \sum_{i=1}^{n} \ell''(\hat{\theta}_{n}; X_{i})$$

$$= -\frac{1}{n} \sum_{i=1}^{n} \left(-\frac{1}{\hat{\theta}^{2}_{n}}; X_{i}\right)$$

$$= \frac{1}{n} \hat{\theta}^{2} \sum_{i=1}^{n} X_{i}$$

$$= \frac{\hat{\theta}^{2}}{\hat{\theta}^{2}}$$

$$= \frac{1}{\bar{x}}$$
notes we have that $[L(\mathbf{X}), U(\mathbf{X})]$ is an asymptoticall are

From Theorem 12.2 in notes we have that $[L(\mathbf{X}), U(\mathbf{X})]$ is an asymptotically exact $1 - \alpha$ confidence interval of θ^* where

$$L(\mathbf{x}) := \hat{\theta}_n - \frac{z_{\alpha/2}}{\sqrt{nJ_n(\hat{\theta}_n)}} = \bar{x} - z_{\alpha/2}\sqrt{\frac{\bar{x}}{n}} \text{ and } U(\mathbf{x}) := \hat{\theta}_n + \frac{z_{\alpha/2}}{\sqrt{nJ_n(\hat{\theta}_n)}} = \bar{x} + z_{\alpha/2}\sqrt{\frac{\bar{x}}{n}}$$

Question - 2.

Suppose you are given the observations

$$\mathbf{x} = (11, 23, 20, 11, 15, 29, 20, 16, 15, 14)$$

presumed to come from some independent, identically distributed random sample X.

Question 2.1

Compute an approximate 95% observed confidence interval for the expectation of μ of X, assumed to be such that $|\mathbb{E}(X)| < \infty$ and $\sigma^2 = \text{Var}(X) < \infty$, explaining all of your reasoning carfully. Note that we are not assuming any model for the observations, except for the independence, and that neither μ nor σ^2 are known.

Hint - Start with the cetnral limit theorem for $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$.

Answer 2.1

Let X be an independent, indentically distributed sample of an unknown model with its mean & variance unknown but finite.

Consider the maximum likelihood estimate $\hat{\mu} := \bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$.

By the Central Limit Theorem we have

$$\frac{\hat{\mu} - \mu^*}{\sqrt{n/\sigma^2}} \to_{\mathcal{D}} Z \sim \text{Normal}(0, 1)$$

Thus

$$\mathbb{P}\left(z_{-\alpha/2} \leq \frac{\hat{\mu}_n - \mu^*}{\sqrt{n/\sigma^2}} \leq z_{\alpha/2}; \mu\right) = 1 - \alpha$$

$$\implies \mathbb{P}\left(\hat{\mu}_n - z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \leq \mu^* \leq \hat{\mu}_n + z_{\alpha/2} \frac{\sigma}{\sqrt{n}}\right) = 1 - \alpha \text{ by rearrangement}$$

Since σ^2 is unknown we consider a consistent sequence of estimators $\{\hat{\sigma}_n^2\}_{n\in\mathbb{N}}$ where

$$\sigma_n^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \hat{\mu}_n)^2$$

From the data we have n = 10, $\hat{\mu}_{10} = \frac{174}{10} \& \hat{\sigma}_{10}^2 = \frac{1}{9} \times \frac{1432}{5} = \frac{1432}{45}$. Since we want a 95% confidence interval we set $1 - \alpha = 0.95 \implies \alpha = 0.05$.

Thus $z_{0.975} = 1.960$ we have

$$\mathbb{P}\left(\frac{174}{10} - 1.96\sqrt{\frac{1432/45}{10}} \le \mu^* \le \frac{174}{10} + 1.960\sqrt{\frac{1432/45}{10}}\right) = 0.95$$

$$\mathbb{P}\left(\mu^* \in [13.9, 20.9]\right) = 0.95$$

Question 2.2

Now suppose you are also told that $\mathbf{X} \stackrel{\text{iid}}{\sim} \text{Poisson}(\theta)$. Compute another approximate 95% confidence interval for the expectation of X, using the Wald approach.

Answer 2.2

Suppose $\mathbf{X} \stackrel{\text{iid}}{\sim} \text{Poisson}(\theta)$.

By Wald's Approach we have that $\mathcal{I}(\mu^*) := [L(\mathbf{X}), U(\mathbf{X})]$ is a $1 - \alpha$ confidence interval of μ where

$$L(\mathbf{x}) := \hat{\mu}_n - \frac{z_{\alpha/2}}{\sqrt{nI(\mu^*)}}$$
 and $U(\mathbf{x}) := \hat{\mu}_n + \frac{z_{\alpha/2}}{\sqrt{nI(\mu^*)}}$

From question 1 we have $I(\mu^*) = n/\bar{x}$ for a poisson iid random sample. Thus, using data from previous part, we have

$$\mathcal{I}(\mu^*) = \left[\hat{\mu}_n - z_{\alpha/2} \sqrt{\frac{\bar{x}}{n^2}}, \hat{\mu}_n + z_{\alpha/2} \sqrt{\frac{\bar{x}}{n^2}} \right] \\
= \left[\frac{174}{10} - 1.96 \sqrt{\frac{174/10}{100}}, \frac{174}{10} + 1.96 \sqrt{\frac{174/10}{100}} \right] \\
= [16.6, 18.2]$$

Question 2.3 - Which of the two confidence intervals do you prefer?

Answer 2.3 - The second since it is a tighter bound on μ^* and encodes information about the possible model of the data, rather than just the data.