

## Problem Set 3

You are strongly encouraged to solve the following exercises before next week's tutorial:

On page 317 onwards (end of section 8), exercises 19 (a)-(c), 53 (a)-(d), 54, 55 (a)-(b), 57 (c), 60 (a)-(e) (you might consider trying the additional exercise first).

### Additional problem:

- (a) Let  $X_1, \dots, X_n \stackrel{\text{i.i.d}}{\sim} \text{Exp}(\lambda)$ . Show that  $\sum_{i=1}^n X_i \sim \text{Gamma}(n, \lambda)$ . (Hint: use the moment generating function)
- (b) Show that if  $U \sim \text{Gamma}(n, \lambda)$  then  $2\lambda U \sim \chi_{2n}^2$ .  
**know pdf of U find pdf of a function of U by transformation method; see if it matches that of chi squared**
- (c) Use part (b) to find a  $(1 - \alpha)100\%$  confidence interval for  $\lambda$ .
- (d) Suppose that we collected 30 observations from the  $\text{Exp}(\lambda)$  distribution, and the sample mean turned out to be 0.9. Compare the resultant 95% confidence interval of part (c) with the approximate (large sample) 95% confidence interval for  $\lambda$  derived in class.

### Solution:

- (a) The moment generating function of  $X \sim \text{Exp}(\lambda)$  is  $M_X(t) = \frac{\lambda}{\lambda - t}$ , and since the  $X_i$ 's are independent

$$M_{\sum_{i=1}^n X_i}(t) = \prod_{i=1}^n M_{X_i}(t) = \prod_{i=1}^n \frac{\lambda}{\lambda - t} = \left( \frac{\lambda}{\lambda - t} \right)^n.$$

This is the moment generating function of a  $\text{Gamma}(n, \lambda)$ , and the uniqueness of the mgf seals the deal.

- (b) The pdf of  $U \sim \text{Gamma}(n, \lambda)$  is  $f_U(u) = \frac{\lambda^n}{(n-1)!} e^{-\lambda u} u^{n-1}$ ,  $u \geq 0$ . If we now define

$V = 2\lambda U$ , then its cdf is given by

$$F_V(v) = \mathbb{P}(V \leq v) = \mathbb{P}\left(U \leq \frac{v}{2\lambda}\right) = F_U\left(\frac{v}{2\lambda}\right),$$

and thus the pdf is

$$\begin{aligned} f_V(v) &= \frac{dF_V(v)}{dv} = \overset{\text{by transformation method}}{f_U\left(\frac{v}{2\lambda}\right) \cdot \frac{1}{2\lambda}} = \frac{\lambda^n}{(n-1)!} e^{-\lambda\left(\frac{v}{2\lambda}\right)} \left(\frac{v}{2\lambda}\right)^{n-1} \cdot \frac{1}{2\lambda} \\ &= \frac{2^{-n}}{(n-1)!} e^{-v/2} v^{n-1}. \end{aligned}$$

Incidentally, this is the pdf of a  $\text{Gamma}(n, 1/2) = \chi_{2n}^2$  random variable.

(c)

$$1 - \alpha = \mathbb{P}\left(\chi_{2n, \alpha/2}^2 \leq 2\lambda \sum_{i=1}^n X_i \leq \chi_{2n, 1-\alpha/2}^2\right) = \mathbb{P}\left(\frac{\chi_{2n, \alpha/2}^2}{2 \sum_{i=1}^n X_i} \leq \lambda \leq \frac{\chi_{2n, 1-\alpha/2}^2}{2 \sum_{i=1}^n X_i}\right),$$

hence  $\left[\frac{\chi_{2n, \alpha/2}^2}{2 \sum_{i=1}^n X_i}, \frac{\chi_{2n, 1-\alpha/2}^2}{2 \sum_{i=1}^n X_i}\right]$  is a  $(1 - \alpha)100\%$  confidence interval for  $\lambda$ .

(d) Substituting  $\alpha = 0.05$ ,  $n = 30$ ,  $\bar{X} = 0.9$ ,  $\sum_{i=1}^{30} X_i = 27$ ,  $\chi_{60, 0.025}^2 = 40.48$  and  $\chi_{60, 0.975}^2 = 83.3$  in the expression of part (c) yields an exact 95% confidence interval for  $\lambda$  –

$$\left[\frac{40.48}{2 \times 27}, \frac{83.3}{2 \times 27}\right] = [0.75, 1.54].$$

In class we found an approximate  $(1 - \alpha)100\%$  confidence interval for  $\lambda$  of the form –

$$\left[1/\bar{X} - \frac{z_{1-\alpha/2}}{\bar{X}\sqrt{n}}, 1/\bar{X} + \frac{z_{1-\alpha/2}}{\bar{X}\sqrt{n}}\right].$$

Recalling that  $z_{0.975} = 1.96$ , we learn that

$$\left[ \frac{1}{0.9} - \frac{1.96}{0.9 \times \sqrt{30}}, \frac{1}{0.9} + \frac{1.96}{0.9 \times \sqrt{30}} \right] = [0.71, 1.51].$$

Overall the two CIs in this case are very similar and of an almost identical length.