

## Tutorial Week 8

STAT3023: Statistical Inference

Semester 2, 2023

1. Suppose  $X$  has a gamma distribution with known shape  $\gamma_0 > 0$  and unknown scale parameter  $\theta$  (see the computer exercise). Then  $X$  has PDF

$$f(x; \gamma_0, \theta) = \frac{x^{\gamma_0-1} e^{-x/\theta}}{\theta^{\gamma_0} \Gamma(\gamma_0)}$$

for  $x > 0$ , and 0 otherwise. Since this is an exponential family with sufficient statistic  $X$  (and  $X$  has a continuous distribution, so no randomisation required), the UMPU test of  $H_0: \theta = \theta_0$  against  $H_1: \theta \neq \theta_0$  is of the form

$$\delta(X) = \begin{cases} 1 & \text{for } X < c \text{ or } X > d \\ 0 & \text{otherwise} \end{cases} \quad (1)$$

where the constants  $c$  and  $d$  ( $c < d$ ) are chosen so that both the following equalities hold:

$$E_{\theta_0} [\delta(X)] = \alpha; \quad (2)$$

$$E_{\theta_0} [X\delta(X)] = \alpha E_{\theta_0}(X). \quad (3)$$

- (a) Write down a formula for  $E_{\theta}(X)$ .  
(b) Show that the conditions (2) and (3) above imply

$$\int_c^d f(x; \gamma_0, \theta_0) dx = \int_c^d f(x; \gamma_0 + 1, \theta_0) dx. \quad (4)$$

- (c) By integrating the right-hand side of (4) by parts, show that  $c$  and  $d$  in (1) satisfy

$$c^{\gamma_0} e^{-c/\theta_0} = d^{\gamma_0} e^{-d/\theta_0}. \quad (5)$$

- (d) Explain why the UMPU test of  $H_0: \theta = \theta_0$  against  $H_1: \theta \neq \theta_0$  rejects for large values of

$$\frac{X}{\mu_0} - \log \left\{ \frac{X}{\mu_0} \right\}$$

where  $\mu_0 = \gamma_0 \theta_0 = E_{\theta_0}(X)$ , the expected value of  $X$  under  $H_0$ .

2. Suppose  $X_1, \dots, X_n$  (for  $n \geq 2$ ) are iid  $N(\mu, \sigma^2)$  and we are interesting in testing

$$H_0: \sigma^2 = \sigma_0^2 \text{ against } H_1: \sigma^2 \neq \sigma_0^2 \quad (6)$$

- (a) The statistic  $Y = \frac{1}{2} \sum_{i=1}^n (X_i - \bar{X})^2 \sim \frac{\sigma^2}{2} \chi_{n-1}^2$  which is the same as a gamma random variable with (known) shape  $\frac{n-1}{2}$  and (unknown) scale parameter  $\sigma^2$ . It turns out that the UMPU test of (6) is the same as the test from question 1 applied to the statistic  $Y$  (which makes sense since  $Y$  is sufficient for the scale parameter  $\sigma^2$ ). Show that this test rejects for large values of

$$\frac{S^2}{\sigma_0^2} - \log \left( \frac{S^2}{\sigma_0^2} \right) \quad (7)$$

where  $S^2 = \frac{2Y}{n-1} = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$  is the sample variance.

- (b) Is the GLRT of the hypotheses (6) above equivalent to the test rejecting for large values of (7)?

3. Suppose  $X_1, \dots, X_5$  are iid  $\text{Poisson}(\theta)$  random variables. Determine the UMP test of  $H_0: \theta = 1$  against  $H_1: \theta < 1$  at level 0.05. **Hint:** you will need to calculate a few Poisson probabilities!
4. Suppose that  $X_1, \dots, X_n$  are iid  $U(0, \theta)$ , with common density

$$f_\theta(x) = \frac{1 \{0 \leq x \leq \theta\}}{\theta}.$$

- (a) Show that the sample maximum  $X_{(n)}$  is a sufficient statistic for  $\theta$ .
- (b) Derive the maximum likelihood estimator  $\hat{\theta}_{\text{ML}}$  and show that it is biased.
- (c) For what value of the multiplier  $c_n$  is the estimator  $c_n \hat{\theta}_{\text{ML}}$ 
  - (i) unbiased;
  - (ii) have the smallest possible MSE, i.e. for which  $E_\theta \left[ \left( c_n \hat{\theta}_{\text{ML}} - \theta \right)^2 \right]$  is minimised?
- (d) Determine the form of the GLRT for testing  $H_0: \theta = \theta_0$  against  $H_1: \theta \neq \theta_0$  at level  $\alpha$ .