

# MATH562 – Fall 2025

## Problem Set: Week 11

1. Suppose we hope to eliminate to nuisance parameters  $\lambda$  from a likelihood  $L(\psi, \lambda)$  by using an integrated likelihood

$$\int L(\psi, \lambda) \, d\lambda.$$

- (a) Criticise this approach by computing the integrated likelihoods when the likelihood is based on two independent exponential variables with parameters (i)  $\lambda$  and  $\lambda\psi$ , (ii)  $1/\lambda$  and  $\psi/\lambda$ , where  $\lambda, \psi > 0$ .
- (b) Now suppose that in (i) the parameters are given a density  $\pi(\psi, \lambda)$  and we compute the resulting marginal density for  $\psi$ . Show that if the corresponding prior density is used in the parametrization in (ii), the problems in (a) do not arise.

- \*2. A random sample  $y_1, \dots, y_n$  of distinct observations,  $y_i \neq y_j$  for  $i \neq j$ , has arisen from an unknown distribution function  $G$ . Consider a multinomial distribution in which  $p_j = \Pr(Y = y_j)$ , for  $j = 1, \dots, n$ .

- (a) Use Lagrange multipliers to show that the empirical distribution function with  $p_j \equiv 1/n$  maximises the likelihood  $\sum_j \log p_j$  of the observed data subject to the constraints  $p_j \geq 0$  and  $\sum_{j=1}^n p_j \leq 1$ .

*Hint:* first argue that its sufficient to consider the case where  $p_j > 0$  for all  $j$  and  $\sum_{j=1}^n p_j = 1$ .

- (b) Now add the constraint that  $\sum_j p_j c_j(\theta) = 0$ , where  $c_j(\theta) \equiv c(y_j; \theta)$  is a  $d \times 1$  function of  $y_j$  and  $\theta$ ; this is the empirical version of the constraint  $E\{c(Y; \theta)\} = 0$ , with expectation taken over  $Y \sim G$ . Show that in this case the log-likelihood for a specific  $\theta$ , maximizing over the  $p_j$  given the constraints, is the *empirical log-likelihood*

$$\ell_E(\theta) = -n \log n - \sum_{j=1}^n \log\{1 + \lambda_\theta^\top c_j(\theta)\}, \quad \text{where } \lambda_\theta \text{ satisfies } 0 = \sum_{j=1}^n \frac{c_j(\theta)}{1 + \lambda_\theta^\top c_j(\theta)}.$$

- (c) If  $c_j(\theta) = y_j - \theta$ , show that the maximum empirical likelihood estimate is  $\hat{\theta}_E = \bar{y}$ , with  $\lambda_\theta = 0$ , and that for any valid  $\theta$  we have  $\sum y_j p_j = \theta$  and  $\min_j y_j < \theta < \max_j y_j$ .

3. In testing  $\alpha \neq 1$  when  $Y_1, \dots, Y_n$  is a random sample from the gamma  $(\alpha, \lambda)$  distribution, with  $\lambda$  unknown (Problem 1(c) of Week 10),

- (a) show that the distribution of the test statistic is invariant to  $\lambda$
- (b) describe how you would simulate the distribution of the test statistic in (a) to obtain  $p$ -values without relying on the asymptotic distribution for  $n \rightarrow \infty$ .
- (c) show that  $\lambda$  may be eliminated by appropriate conditioning.

4. (a) If  $P \sim U(0, 1)$ , show that  $Q = 2 \min(P, 1 - P) \sim U(0, 1)$ .

- (b) What are the achievable significance levels for testing that a single Poisson variable has mean  $\mu_0 = 2$ , with alternative mean (i) greater than 2 and (ii) less than 2? Does this matter for computing confidence intervals?