

# Statistical Methods for Particle Physics

## Problems on parameter estimation

The two problems below are from an a 2012 University of London exam on Statistical Data Analysis

**PART  
MARKS**

1. An experiment yields  $n$  time values,  $t_1, \dots, t_n$ , and a calibration value  $y$ , all of which are independent. The time measurements are all exponentially distributed with a mean of  $\tau + \lambda$  and the calibration measurement,  $y$ , follows a Gaussian distribution with a mean  $\lambda$  and standard deviation  $\sigma$ . Suppose that  $\sigma$  is known and we want to estimate  $\tau$  and  $\lambda$ .

- (a) Write down the likelihood function for  $\tau$  and  $\lambda$ , and show that the Maximum Likelihood (ML) estimators for these parameters are

$$\hat{\tau} = \frac{1}{n} \sum_{i=1}^n t_i - y$$

and

$$\hat{\lambda} = y. \quad [10]$$

- (b) Find the variances of  $\hat{\tau}$  and  $\hat{\lambda}$ , and the covariance  $\text{cov}[\hat{\tau}, \hat{\lambda}]$ . Use the fact that the variance of an exponentially distributed variable is equal to the square of its mean. [10]
- (c) Show using a sketch how a contour of constant log-likelihood can be used to determine the standard deviations of  $\hat{\tau}$  and  $\hat{\lambda}$ .

Explain qualitatively how you would expect the variance of  $\hat{\tau}$  to be different if the parameter  $\lambda$  were to be known exactly. [10]

- (d) Show that the (co)variances of  $\hat{\tau}$  and  $\hat{\lambda}$  obtained from the matrix of second derivatives of the log-likelihood are the same as those found in (c). Use the fact that the inverse of a  $2 \times 2$  matrix

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

is

$$A^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}. \quad [10]$$

**PART  
MARKS**

2. A measurement yields two values,  $n_a$  and  $n_b$ , which are independent and Poisson distributed with mean values  $\nu_a$  and  $\nu_b$ , respectively. From these parameters we define the total mean,  $\nu$ , and asymmetry parameter,  $\alpha$ , as

$$\nu = \nu_a + \nu_b ,$$

$$\alpha = \frac{\nu_a - \nu_b}{\nu_a + \nu_b} .$$

Recall that the Poisson distribution is  $P(n; \nu) = \nu^n e^{-\nu} / n!$ .

- (a) Write down the likelihood function in terms of  $\nu$  and  $\alpha$  and find the Maximum Likelihood (ML) estimators for these parameters. **[10]**
- (b) Estimate the variance of  $\hat{\alpha}$  as a function of  $\nu$  and  $\alpha$  using error propagation. **[10]**
- (c) Consider the Bayesian approach to inference about  $\nu$  and  $\alpha$ , and suppose we take the prior pdf for the parameters to be

$$\pi(\nu, \alpha) \propto 1 / \sqrt{\nu} .$$

Find as a proportionality the joint posterior pdf for  $\nu$  and  $\alpha$  given  $n_a$  and  $n_b$ .

Show that  $\nu$  and  $\alpha$  are independent, and that the marginal posterior pdfs follow

$$p(\nu | n_a, n_b) \propto \nu^{(n_a + n_b - 1/2)} e^{-\nu} ,$$

$$p(\alpha | n_a, n_b) \propto (1 + \alpha)^{n_a} (1 - \alpha)^{n_b} .$$

- (d) Find the posterior modes for  $\nu$  and  $\alpha$ . Comment on how these relate to the corresponding ML estimators. **[6]**