## RWTH Aachen Graduierten-Kolleg Statistical Methods for Particle Physics Problem Sheet 4

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

PART MARKS

- 1. An experiment yields n time values,  $t_1, ..., t_n$ , 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

$$\hat{\lambda} = y. ag{10}$$

(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 2×2 matrix

$$A = \left( \begin{array}{cc} a & b \\ c & d \end{array} \right)$$

is

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

PART MARKS

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

$$v = v_a + v_b$$
,

$$\alpha = \frac{v_{a} - v_{b}}{v_{a} + v_{b}}.$$

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

(a) Write down the likelihood function in terms of v 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 v and  $\alpha$ , and suppose we take the prior pdf for the parameters to be

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

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

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

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

$$p(\alpha \mid n_{a}, n_{b}) \propto (1 + \alpha)^{n_{a}} (1 - \alpha)^{n_{b}}$$
 [14]

(d) Find the posterior modes for v and  $\alpha$ . Comment on how these relate to the corresponding ML estimators.

[6]