

Problems in Parameter Estimation
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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 λ and standard deviation σ . Suppose that σ is known and we want to estimate τ and λ .

(a) Write down the likelihood function for τ and λ , 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

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

(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.

(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 λ were to be known exactly.

(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

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

is

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

2. A measurement yields two values, n_a and n_b , which are independent and Poisson distributed with mean values ν_a and ν_b , respectively. From these parameters we define the total mean, ν , and asymmetry parameter, α , as

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

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

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

- Write down the likelihood function in terms of ν and α and find the Maximum Likelihood (ML) estimators for these parameters.
- Estimate the variance of $\hat{\alpha}$ as a function of ν and α using error propagation.
- Consider the Bayesian approach to inference about ν and α , 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 ν and α given n_a and n_b .

Show that ν and α are independent, and that the marginal posterior pdfs follow

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

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

- Find the posterior modes for ν and α . Comment on how these relate to the corresponding ML estimators.