



Read the questions carefully and make your own assumptions if needed.

**1 Bernoulli**

Let the data  $x = (x_1, \dots, x_n)$  be a random sample from the Bernoulli  $B(p)$ .

- a) Calculate the Fisher information in a single Bernoulli variable, in the sample, and in  $y = n\bar{x}$ . Hint: Use independence and sufficiency.
- b) Determine the UMVU estimators of the parameters  $\text{Var } X_1$  and  $p^2$ . Compare the variances with the corresponding Cramer-Rao lower bounds. Calculate the Fisher information in the estimators and the resulting Cramer-Rao lower bounds.
- c) Let  $T$  be a statistic. Show that  $\tau = \text{E}T$  is a polynomial in  $p$ , and that any  $n$ 'th order polynomial can be obtained like this. Can you find UMVU estimators for  $\tau$ ,  $1/p$ , and  $\eta = \ln p - \ln(1 - p)$ ? What about UMRU estimators using the loss  $l = (\hat{\eta} - \eta)^2$ ?
- d) When is it possible for a statistic  $T$  to obtain the lower bound in the Cramer-Rao inequality? Hint: Equality in the Cauchy-Schwarz inequality.
- e) Let  $f$  be the  $B(p)$  density. Explain that  $\sqrt{f}$  is a unit vector in the plane and illustrate with a drawing. Identify  $p$  with the corresponding point in the plane. Find a formula for the direct distance between  $\hat{p}$  and  $p$ , and also the distance along the unit circle. Explain that this gives two different metrics on the space of Bernoulli densities. Is there a relation between the resulting two metric spaces?
- f) What is the Jeffreys' prior for  $p$ ? Show that the Haldane prior  $1/[p(1-p)]$  gives a uniform prior on  $\eta = \ln p - \ln(1 - p)$ . Show that both priors determines a metric by

$$d(\hat{p}, p) = \left| \int_{\hat{p}}^p \pi(p) dp \right| \quad (1)$$

and calculate explicit formulas for the distances. Find a formula for the distance given by the Fisher information metric.

- g) Find a formula for the Kullback-Leibler divergence  $D(\hat{p} \| p)$ . Is this given by a prior density? Decide if the previous distances and divergence define convex loss functions.
- h) How would you calculate Bayes estimators for  $p$  using the previous priors and with loss given by absolute distance or distance squared?

## 2 Fisher information for the multinomial

Let  $f$  be the density a random point  $X$  with values  $R(X) = \{x_1, \dots, x_m\}$ .

- a) Explain that  $\sqrt{f}$  is a unit vector in the vector space  $\mathbb{R}^m$ . Assume  $P_X$  is known when the time  $t$  is known. Explain that this defines a statistical model, and also a path on the unit sphere in  $\mathbb{R}^m$ . Let  $\iota(t) = E S^2$  with  $S = \partial_t \ln f(X)$ . Prove that the length of a part of the assumed smooth path is

$$d(t_1, t_2) = \left| \int_{t_0}^{t_1} \frac{\sqrt{\iota(t)}}{2} dt \right| \quad (2)$$

Hint: Distance is given by speed and time.

- b) Let  $\tau = ET$ . Prove that the Cauchy inequality  $|E[(T - \tau)S]| \leq \|T - \tau\| \|S\|$  implies the Cramer-Rao inequality  $\text{Var } T \geq \dot{\tau}^2 / \iota$ .
- c) Assume  $P_X$  is known when  $\theta$  is known and assume  $\theta$  is known when  $t$  is known. Prove  $\iota(t) = \dot{\theta} \iota \dot{\theta}^T$  where  $\iota$  is the Fisher information. Explain that  $\iota$  is a differential metric on the model space  $\Omega_\Theta$ . The length element and the volume element  $\sqrt{|\det(\iota)|} d\theta$  are both coordinate independent. What does this mean? Hint: Equation (2) and the chain rule for differentiation.
- d) Prove the multivariate Cramer-Rao inequality. Hint: Reduce to a component of  $T$  and use the chain rule on the one-dimensional Cramer-Rao inequality.
- e) Are the previous arguments valid if it is only assumed that  $f$  is a density? Find  $d(\beta_1, \beta_2)$  for exponential distributions with scales  $\beta_1$  and  $\beta_2$ .

## 3 Standard uncertainty and Cramer-Rao

Let  $X \sim N(\theta, \Sigma)$  where the variance matrix  $\text{Var } X = \Sigma$  is known.

- a) Let  $\tau = A + B\theta$  where  $A$  and  $B$  are matrices. Find an UMVU  $T$  for  $\tau$  and determine its distribution. Compare  $\text{Var } T$  with the Cramer-Rao lower bound.
- b) Let  $\tau$  be a parameter. Determine the MLE of  $\tau$ . Justify that the MLE has a distribution close to the normal distribution if  $\Sigma$  is small. Compare the variance of this normal distribution with the Cramer-Rao lower bound.
- c) Explain that the previous gives a recipe for calculating a standard uncertainty when estimating  $\tau$ , but that the actual uncertainty is larger. Explain how you could use simulation on a computer to find an improved standard uncertainty. Explain how you could use a finite difference approximation to calculate a standard uncertainty if calculation of  $\tau$  is costly.
- d) The length  $\lambda$  of a pendulum with period  $\tau$  is

$$\lambda = \left( \frac{\tau}{2\pi} \right)^2 g \quad (3)$$

Assume that  $T$  estimates  $\tau$  and that  $S$  estimates  $g$ . Find an estimator of  $\lambda$ . Assume that  $T$  and  $S$  are unbiased with normal distribution and known variances. Find an UMVU of  $\lambda$ . Determine standard uncertainties for the two competing estimates, and compare with the resulting Cramer-Rao lower bounds.