

# Mathematical Tripos: Part IB Solutions

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## 1 Question 2

In a genetics experiment, a sample of  $n$  individuals was found to include  $a, b, c$  of the three possible genotypes GG, Gg, gg respectively. The population frequency of a gene of type G is  $\theta/(\theta+1)$ , where  $\theta$  is unknown, and it is assumed that the individuals are unrelated and that two genes in a single individual are independent. Show what  $\theta$  is proportional to and calculate the MLE.

### Solution

The likelihood function is given as:  $\frac{\theta}{\theta+1}^{2a} \cdot \frac{2\theta}{(\theta+1)^2}^b \cdot \frac{1}{\theta+1}^{2c}$

The log likelihood function is given as:

$$(2a + b) \log \theta - (2a + 2b + 2c) \log (\theta + 1) = 0$$

Solving this, we get:

$$\theta = \frac{2a+b}{b+2c}$$

## 2 Question 3

For some  $n \geq 2$ , suppose that  $X_1, \dots, X_n$  are iid random variables uniformly distributed on  $[\theta, 2\theta]$  for some  $\theta > 0$ . Show that  $\tilde{\theta} = \frac{2}{3}X_1$  is an unbiased estimator of  $\theta$ . Show that  $T(\mathbf{X}) = (\min_i X_i, \max_i X_i)$  is a minimal sufficient statistic for  $\theta$ . Use the Rao-Blackwell theorem to find an unbiased estimator  $\hat{\theta}$  of  $\theta$  which is a function of  $T$  and whose variance is strictly smaller than the variance of  $\tilde{\theta}$  for all  $\theta > 0$ .

### Solution

The likelihood function is given by:

$$L(\theta | \mathbf{x}) = \prod_{i=1}^n \frac{1}{\theta} 1[\min(X_i) > \theta, \max(X_i) < 2\theta]$$

Hence, a sufficient statistic is  $T(\mathbf{X}) = (\min_i X_i, \max_i X_i)$

$\tilde{\theta} = \frac{2}{3}X_1$  is unbiased because  $E[X_1] = \frac{3\theta}{2}$

An unbiased estimator for  $\theta$  is:

$$\begin{aligned} \hat{\theta}^* &= E\left(\frac{2}{3}X_1 \mid \min_i X_i = a, \max_i X_i = b\right) \\ &= \frac{2a}{3n} + \frac{2b}{3n} + \frac{n-2}{n} \frac{2a+b}{3} = \frac{a+b}{3} \end{aligned}$$

## 3 Question 7

(a) Let  $X_1, \dots, X_n$  be iid with  $X_i \sim U[0, \theta]$ . Find the maximum likelihood estimator  $\hat{\theta}$  of  $\theta$ . Show that the distribution of  $R(\mathbf{X}, \theta) = \hat{\theta}/\theta$  does not depend on  $\theta$ , and use  $R(\mathbf{X}, \theta)$  to find a  $100(1 - \alpha)\%$  confidence interval for  $\theta$  for  $0 < \alpha < 1$ .

(a) **Solution** Please refer to the section 7. As we know,  $\hat{\theta} = \max(X_i)$ .

Substituting the MLE into the CDF, we get:

$$\begin{aligned} P(x_n \leq x) &= P(x_1 \leq x, x_2 \leq x, \dots, x_n \leq x) \\ &= (F(x))^n \end{aligned}$$

Differentiating the above, we get the pdf, which is expressed as

$$f_{\hat{\theta}} = n(F(x))^{n-1}f(x)$$

We know that the CDF of a uniform distribution is given as  $F(x) = \frac{1}{\theta}x$ , the pdf is given as  $f(x) = \frac{1}{\theta}$ , substituting those in, we get:

$$f_{\hat{\theta}/\theta} = f_{\hat{\theta}}(\theta x) \cdot \theta = nx^{n-1}, \quad (0 < x < 1)$$

, this does not depend on  $\theta$ .

In order to find the confidence interval, we can set (for the upper bound):

$$\int_0^u nx^{n-1} = (1 - \alpha/2)$$

In order to find the confidence interval, we can set (for the lower bound):

$$\int_0^l nx^{n-1} = (\alpha/2)$$

(b) The lengths (in minutes) of calls to a call centre may be modelled as iid exponentially distributed random variables, and  $n$  such call lengths are observed. The original sample is lost, but the data manager has noted down  $n$  and  $t$  where  $t$  is the total length of the  $n$  calls in minutes. Derive a 95 percent confidence interval for the probability that a call is longer than 2 minutes if  $n = 50$  and  $t = 105.3$ .

(b) **Solution** First of all, we can try to write out the cdf of the exponential distribution.

$$\begin{aligned} f_X(x) &= \int_2^\infty \lambda e^{-\lambda x} dx \\ &= [1 - e^{-\lambda x}]_2^\infty = e^{-2\lambda} \end{aligned}$$

We don't actually know what is the distribution of the  $\lambda$  in the expression.

However, what we know is that the the distribution of the calls is modelled by  $Gamma(50, \lambda)$ , hence, the distribution of  $\lambda t$  may be expressed as  $Gamma(50, 1)$ . Then, we can just read off from the  $Gamma(50, 1)$  table for what the confidence interval is.

## 4 Question 8

Suppose that  $X_1 \sim N(\theta_1, 1)$  and  $X_2 \sim N(\theta_2, 1)$  independently, where  $\theta_1$  and  $\theta_2$  are unknown. Show that  $(\theta_1 - X_1)^2 + (\theta_2 - X_2)^2$  has a  $\chi^2_2$  distribution and that this is the same as Exponential  $(\frac{1}{2})$ , i.e., the exponential distribution with mean 2. Show that both the square  $S$  and circle  $C$  in  $R^2$ , given by

$$\begin{aligned} S &= \{(\theta_1, \theta_2) : |\theta_1 - X_1| \leq 2.236; |\theta_2 - X_2| \leq 2.236\} \\ C &= \left\{(\theta_1, \theta_2) : (\theta_1 - X_1)^2 + (\theta_2 - X_2)^2 \leq 5.991\right\} \end{aligned}$$

are 95% confidence regions for  $(\theta_1, \theta_2)$ . Hint:  $\Phi(2.236) = (1 + \sqrt{.95})/2$ , where  $\Phi$  is the distribution function of  $N(0, 1)$ . What might be a sensible criterion for choosing between  $S$  and  $C$ ?

We know that, if  $Z_1, \dots, Z_k$  are independent, standard normal random variables, then the sum of their squares,

$$Q = \sum_{i=1}^k Z_i^2$$

is distributed according to the chi-squared distribution with  $k$  degrees of freedom. This is usually denoted as  $Q \sim \chi^2(k)$  or  $Q \sim \chi^2_k$

## 5 Question 9

Suppose that the number of defects on a roll of magnetic recording tape is modelled with a Poisson distribution for which the mean  $\lambda$  is known to be either 1 or 1.5. Suppose the prior mass function for  $\lambda$  is

$$\pi_{\lambda}(1) = 0.4, \quad \pi_{\lambda}(1.5) = 0.6.$$

A random sample of five rolls of tape has  $\mathbf{x} = (3, 1, 4, 6, 2)$  defects respectively. Show that the posterior distribution for  $\lambda$  given  $\mathbf{x}$  is

$$\pi_{\lambda|\mathbf{X}}(1 | \mathbf{x}) = 0.012, \quad \pi_{\lambda|\mathbf{X}}(1.5 | \mathbf{x}) = 0.988$$

## 6 Question 10

Suppose  $X_1, \dots, X_n$  are iid with (conditional) probability density function  $f(x | \theta) = \theta x^{\theta-1}$  for  $0 < x < 1$  (and is zero otherwise), for some  $\theta > 0$ . Suppose that the prior for  $\theta$  is  $\text{Gamma}(\alpha, \beta)$ ,  $\alpha > 0, \beta > 0$ . Find the posterior distribution of  $\theta$  given  $\mathbf{X} = (X_1, \dots, X_n)$  and the Bayesian estimator of  $\theta$  under quadratic loss. +11 For some  $n \geq 3$ , let  $\epsilon_1, \dots, \epsilon_n$  be iid with  $\epsilon_i \sim N(0, 1)$ . Set  $X_1 = \epsilon_1$  and  $X_i = \theta X_{i-1} + (1 - \theta^2)^{1/2} \epsilon_i$  for  $i = 2, \dots, n$  and some  $\theta \in (-1, 1)$ . Find a sufficient statistic for  $\theta$  that takes values in a subset of  $R^3$ .

## 7 MLE for different distributions

MLE for different distributions

Exponential distribution

$$f_X(x) = \lambda e^{-\lambda x}$$

likelihood function is given as:

$$L(x_i) = \lambda^n e^{-\lambda \sum_{i=1}^n x_i}$$

log-likelihood function is given as:

$$n \ln(\lambda) - \lambda \sum_{i=1}^n x_i \Rightarrow \frac{d \log(L(x_i))}{d \lambda} = \frac{n}{\lambda} - \sum x_i$$

Setting derivative to zero:

$$\lambda = \frac{\sum x_i}{n}$$

Normal distribution

$$f_X(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{x-\mu}{\sigma} \right)^2}$$

likelihood function is given as:

$$L(x_i) = \left( \frac{1}{\sigma \sqrt{2\pi}} \right)^n e^{-\frac{1}{2} \sum_{i=1}^n \left( \frac{x_i - \mu}{\sigma} \right)^2}$$

log-likelihood function is given as:

$$n \frac{1}{2} \log(2\pi) + n \frac{1}{2} \log(\sigma^2) + \left[ + \frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2 \right]$$

taking partial derivative wrt.  $\sigma^2$ :

$$\frac{d \log(L(x_i))}{d(\sigma^2)} = \frac{n}{2\sigma^2} - \frac{1}{2\sigma^4} \sum_{i=1}^n (x_i - \mu)^2 \rightarrow \hat{\sigma}^2 = \frac{\sum (x_i - \mu)^2}{n}$$

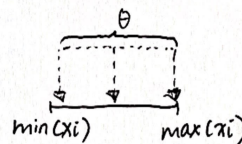
$$\frac{d \log(L(x_i))}{d(\mu)} = \frac{n\mu}{\sigma^2} - \frac{1}{\sigma^2} \sum_{i=1}^n x_i \rightarrow \hat{\mu} = \bar{x}$$

Uniform Distribution

$$\text{likelihood} \Rightarrow \prod_{i=1}^n \frac{1}{\theta} (1_{(0 < x_i < \theta)}) = \frac{1}{\theta^n} (1_{(0 < x_i < \theta)})$$

$$\text{log-likelihood function} \Rightarrow -n \log(\theta) (1_{(0 < x_i < \theta)})$$

$$= -n \log(\theta) (1_{[\min(x_i) > 0; \max(x_i) < \theta]}) \Rightarrow \text{setting } \hat{\theta} = \max(x_i)$$



maximizes MLE