

ENGINEERING TRIPOS PART IIB

Wednesday 6 May 2009 9 to 10.30

Module 4F6

SIGNAL DETECTION AND ESTIMATION

*Answer not more than **three** questions.*

All questions carry the same number of marks.

*The **approximate** percentage of marks allocated to each part of a question is indicated in the right margin.*

There are no attachments

STATIONERY REQUIREMENTS

Single-sided script paper

SPECIAL REQUIREMENTS

Engineering Data Book

CUED approved calculator allowed.

**You may not start to read the questions
printed on the subsequent pages of this
question paper until instructed that you
may do so by the Invigilator**

- 1 a) Define *Information* and *Entropy* of a probability distribution and describe, in detail, how *Maximum Entropy* methods may be used to assign probability distributions. [30%]
- b) Given the first moment of a distribution (from experimental measurements, for example) show, using Lagrange multipliers, that the distribution having *Maximum Entropy* is an Exponential distribution. [40%]
- c) Derive an expression for the entropy of the Exponential distribution in terms of its standard deviation. [30%]

- 2 a) Define *Fisher Information* and *Cramer-Rao Lower Bound* and show that for an unbiased estimator, $\hat{\theta}(x)$, of a parameter θ , the variance associated with $\hat{\theta}(x)$ satisfies

$$\text{var}(\hat{\theta}(x)) \geq I_{\theta}^{-1}$$

where I_{θ} is the *Fisher Information* for the scalar parameter θ . [40%]

- b) Derive the following condition for an efficient unbiased estimator:

$$\frac{\partial \ln p(x|\theta)}{\partial \theta} = I_{\theta}(\hat{\theta}(x) - \theta) \quad (1)$$

where $p(x|\theta)$ is the likelihood function and $\hat{\theta}(x)$ is an estimator for θ .

- [30%]
- c) Describe how equation (1) leads to the *Neyman-Fisher* factorization theorem. [10%]
- d) Using Bayesian reasoning, derive the *Neyman-Fisher* factorization theorem. [20%]

3 a) Define the terms *Error of the first kind* and *Error of the second kind* explaining the role they play in Detection Theory and explain the role played by the *log-likelihood ratio*. [40%]

b) In an M-ary digital transmission system the source signal can take one of M possible levels during a symbol period. In each symbol period the detector makes N measurements $\mathbf{y} = [y_1 y_2 \dots y_N]^T$ of the channel output.

i) Show that the Maximum A-Posteriori (MAP) decision rule for the detector may be expressed as: [30%]

$$\text{Choose } H_i \text{ if } \max_{H_j} \{p(\mathbf{y}|H_j)P(H_j)\} = p(\mathbf{y}|H_i)P(H_i).$$

ii) Show that the average error probability P_e for the detector is given by: [30%]

$$P_e = 1 - \sum_{i=1}^M P(D_i|H_i)P(H_i)$$

where:

$p(\mathbf{y}|H_i)$ is the probability density of the observation vector \mathbf{y} conditional on hypothesis H_i being in force;

$P(H_i)$ is the *a-priori* probability of hypothesis H_i ;

$P(D_i|H_i)$ is the probability of deciding in favour of hypothesis H_i when H_i is in force.

4 a) Describe, in detail, the Neyman-Pearson decision rule applied to detection theory and discuss the advantages and disadvantages of this decision rule over the MAP and Bayes criteria. [30%]

b) Define the *Receiver Operating Characteristic* of a detector and describe its role in both detection theory and data classification. [30%]

c) Show that the value of the threshold for the Neyman-Pearson test for a single observation is given by the slope of the receiver operating characteristic (ROC) at the required false alarm probability. [40%]

END OF PAPER