

DEPARTMENT OF ELECTRICAL AND ELECTRONIC ENGINEERING
EXAMINATIONS 2019

MSc and EEE/EIE PART IV: MEng and ACGI

Corrected copy

INFORMATION THEORY

Correction 10:55
2c

Thursday, 9 May 10:00 am

Time allowed: 3:00 hours

There are FOUR questions on this paper.

Answer ALL questions.

All questions carry equal marks

Any special instructions for invigilators and information for candidates are on page 1.

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Information for students

Notation:

- (a) Random variables are shown in Tahoma font. x , \mathbf{x} , \mathbf{X} denote a random scalar, vector and matrix respectively.
- (b) The size of a set A is denoted by $|A|$.
- (c) By default, the logarithm is to the base 2.
- (d) \oplus denotes the exclusive-or operation, or modulo-2 addition.
- (e) “i.i.d.” means “independent identically distributed”.
- (f) $H(\cdot)$ is the entropy function.
- (g) $C(x) = \frac{1}{2} \log_2(1+x)$ is the capacity function for the Gaussian channel in bits/channel use.

The Questions

1. Basics of information theory.

- a) Suppose x_1 and x_2 are i.i.d. Bernoulli random variables taking values of 0 and 1 with equal probabilities ($p = 0.5$). Let $y = \max(x_1, x_2)$. Compute the following entropy or mutual information:

- i) $H(y)$
- ii) $I(x_1; y)$
- iii) $I(x_{1:2}; y)$

[9]

- b) For $\mathbf{p} = [\frac{1}{3}, \frac{2}{3}]$ and $\mathbf{q} = [\frac{1}{2}, \frac{1}{2}]$, compute the relative entropy terms $D(\mathbf{p}||\mathbf{q})$ and $D(\mathbf{q}||\mathbf{p})$.

[6]

- c) Data processing theorem for relative entropy. Suppose after going through a channel with transitional probabilities $W(y|x)$, distribution \mathbf{p} becomes \mathbf{p}' , while \mathbf{q} becomes \mathbf{q}' . Show that

$$D(\mathbf{p}'||\mathbf{q}') \leq D(\mathbf{p}||\mathbf{q})$$

Hint: use the so-called log-sum inequality (which can be derived from Jensen's inequality)

$$\sum_{i=1}^n a_i \log \frac{a_i}{b_i} \geq \left(\sum_{i=1}^n a_i \right) \log \frac{\sum_{i=1}^n a_i}{\sum_{i=1}^n b_i}$$

where a_i and b_i 's are nonnegative numbers.

[10]

2. Fano's inequality.

Consider the Markov chain shown in Fig. 2.1, where X and Y are discrete random variables, and \hat{X} is the estimate of X .



Fig. 2.1. Markov chain arising in Fano's inequality.

- a) Define a random variable $e = (\hat{X} \neq X) \in \{0,1\}$. Justify each step of the following derivations.

$$\begin{aligned}
 H(e, X | Y) &\stackrel{(1)}{=} H(X | Y) + H(e | X, Y) \stackrel{(2)}{=} H(e | Y) + H(X | e, Y) \\
 &\stackrel{(3)}{\Rightarrow} H(X | Y) + 0 \leq H(e) + H(X | e, Y) \\
 &\stackrel{(4)}{=} H(e) + H(X | Y, e=0)(1-p_e) + H(X | Y, e=1)p_e \\
 &\stackrel{(5)}{\leq} H(p_e) + 0 \times (1-p_e) + \log(|X|-1)p_e \\
 &\stackrel{(6)}{\Rightarrow} p_e \geq \frac{(H(X | Y) - H(p_e))^{(7)}}{\log(|X|-1)} \geq \frac{(H(X | Y) - 1)}{\log(|X|-1)}
 \end{aligned}$$

[8]

- b) Given the following joint distribution

$X \backslash Y$	a	b
1	1/3	1/12
2	1/12	1/3
3	1/12	1/12

Find the minimum error probability corresponding to the optimum estimator of X given Y , and compare it with Fano's inequality for this problem.

[10]

- c) Suppose X is a finite-entropy random variable defined on an infinite set X , e.g., the integers $\{\dots, -2, -1, 0, 1, 2, \dots\}$. How would you modify Fano's inequality in this case?

[7]

3. Gaussian sources and channels.

- a) Justify each step in the following derivation of the entropy of multivariate Gaussian sources.

Given mean, \mathbf{m} , and symmetric positive definite covariance matrix \mathbf{K} , we have the probability density function

$$X_{1:n} \sim \mathbf{N}(\mathbf{m}, \mathbf{K}) \Leftrightarrow f(\mathbf{x}) \stackrel{(1)}{=} |2\pi\mathbf{K}|^{-n/2} \exp\left(-\frac{1}{2}(\mathbf{x} - \mathbf{m})^T \mathbf{K}^{-1}(\mathbf{x} - \mathbf{m})\right)$$

Then,

$$\begin{aligned} h(f) &\stackrel{(2)}{=} -(\log e) \int f(\mathbf{x}) \times \left(-\frac{1}{2}(\mathbf{x} - \mathbf{m})^T \mathbf{K}^{-1}(\mathbf{x} - \mathbf{m}) - \frac{1}{2} \ln |2\pi\mathbf{K}| \right) d\mathbf{x} \\ &\stackrel{(3)}{=} \frac{1}{2} \log(e) \times \left(\ln |2\pi\mathbf{K}| + E\left((\mathbf{x} - \mathbf{m})^T \mathbf{K}^{-1}(\mathbf{x} - \mathbf{m})\right) \right) \\ &\stackrel{(4)}{=} \frac{1}{2} \log(e) \times \left(\ln |2\pi\mathbf{K}| + E \operatorname{tr}\left((\mathbf{x} - \mathbf{m})(\mathbf{x} - \mathbf{m})^T \mathbf{K}^{-1}\right) \right) \\ &\stackrel{(5)}{=} \frac{1}{2} \log(e) \times \left(\ln |2\pi\mathbf{K}| + \operatorname{tr}\left(E(\mathbf{x} - \mathbf{m})(\mathbf{x} - \mathbf{m})^T \mathbf{K}^{-1}\right) \right) \\ &\stackrel{(6)}{=} \frac{1}{2} \log(e) \times \left(\ln |2\pi\mathbf{K}| + \operatorname{tr}(\mathbf{K}\mathbf{K}^{-1}) \right) \\ &\stackrel{(7)}{=} \frac{1}{2} \log(e) \times \left(\ln |2\pi\mathbf{K}| + n \right) \stackrel{(8)}{=} \frac{1}{2} \log(e^n) + \frac{1}{2} \log(|2\pi\mathbf{K}|) \\ &\stackrel{(9)}{=} \frac{1}{2} \log(|2\pi e\mathbf{K}|) \stackrel{(10)}{=} \frac{1}{2} \log((2\pi e)^n |\mathbf{K}|) \quad \text{bits} \end{aligned}$$

[10]

- b) Consider an additive-noise fading channel

$$Y = V X + Z$$

where Z is additive noise, V is a random variable representing the fading coefficient, and Z and V are independent of each other and of X . Show that the knowledge of fading coefficient V improves the capacity:

$$I(X; Y | V) \geq I(X; Y).$$

[8]

- c) Consider the additive Gaussian noise channel

$$Y_i = X_i + Z_i$$

where Z_i 's are independent with mean μ_i and variance N_i . X_i 's have power constraint P . Find the capacity if

- $\mu_i = 0$, for all i .
- μ_i is unknown to the transmitter or receiver, but i.i.d. Gaussian with mean 0 and variance N_i for all i .

[7]

4. Network information theory.

a) Multi-access channel.

- i) With the help of a diagram, explain the term multi-access channel. Give an example.
- ii) Write down the capacity region of the two-user bandlimited multi-access channel with additive white Gaussian noise, where the users have equal powers P , bandwidth W , and the noise power is N . Draw the region and explain how to achieve the two corner points.
- iii) Find the capacity region for this two-user multiple access channel with infinite bandwidth W . Show that both senders can send at their individual capacities, i.e., infinite bandwidth eliminates interference.

[15]

b) Parallel transmission. Consider a scenario where the transmitted signal X of power P is received by two antennas:

$$Y_1 = \alpha X + Z_1$$

$$Y_2 = (1 - \alpha)X + Z_2$$

where $0 < \alpha < 1$, Z_1 and Z_2 are independent Gaussian noises of zero mean and power N .

- i) Assuming that both signals Y_1 and Y_2 are available at a common decoder $\mathbf{Y} = (Y_1; Y_2)$; what is the capacity of the channel from the sender to the common decoder?
- ii) If instead the two receivers Y_1 and Y_2 each independently decode their signals, this becomes a broadcast channel. Let R_1 be the rate to base station 1 and R_2 be the rate to base station 2. Find the capacity region of this channel.

[10]

