

# Information Theory, Exam 1 (02 Feb, 2023)

Serial Number: \_\_\_\_\_

Roll Number: \_\_\_\_\_

## Important

- Any malpractice will lead to instant fail grade.
- For Part A: provide only the final answers in this question paper itself. For Part B: provide complete solutions in the answer booklets.
- **Write your roll number and “serial number” both in this question paper and in the answer sheet.**
- No breaks during the exam. No books, notes, laptops, calculators, mobile devices etc. are allowed.

## Part A (7 questions $\times$ 1 mark)

**Instructions:** Provide your **answers in this question paper** only. Write only the final answers in the space provided here.

1. If  $T$  is the number of fair coin tosses needed by an optimal algorithm to generate a discrete random variable  $X$ , give upper and lower bounds on  $\mathbb{E}(T)$ .

**Answer:**

2. Write the expression for the log-sum inequality.

**Answer:**

3. Write the expression for data processing inequality if  $X \rightarrow Y \rightarrow Z$ .

**Answer:**

4.  $H(X_1, \dots, X_n) = \sum_{i=1}^n H(X_i)$  if and only if

**Answer:**

5. If  $X$  and  $Y$  are Bernoulli random variables and  $\alpha = P[X \neq Y]$ . Then  $Y$  contains the least information about  $X$  for what value of  $\alpha$ ?

**Answer:**

6. Give an upper bound on  $P[p_X(X) \leq a]$ , where  $a < 1$  and  $X$  is a discrete random variable. The upper bound must be in terms of  $H(X)$  and  $a$ .

**Answer:**

7. If  $X_1, \dots, X_n$  are i.i.d with probability mass function  $p_X$ , what is the limit, in probability, of  $(p(X_1, \dots, X_n))^{\frac{1}{n}}$ ?

**Answer:**

## Part B (3 questions $\times$ 5 marks)

**Instructions:** Provide **mathematically precise answers, with complete proofs**, in the answer booklet. Any result proved during the lectures can be stated and used in your answer without giving a proof.

8. Let  $X$  be a discrete random variable with probability mass function  $p(x)$ ,  $x \in \mathcal{X}$ . Let  $q(x)$ ,  $x \in \mathcal{X}$  be any probability mass function defined on  $\mathcal{X}$ .

Show that  $\sum_{x \in \mathcal{X}} p(x) \log \frac{1}{q(x)} \geq H(X)$ . When is equality attained in this inequality?

9. Let  $\mathcal{N}(\mu, \sigma^2)$  denote the normal (Gaussian) distribution with mean  $\mu$  and variance  $\sigma^2$ . Let the parameter  $\theta \in \{0, 1\}$ , i.e., it assumes only two possible values. Let  $X = (X_1, X_2)$  and

if  $\theta = 0$  :  $X_1 \sim \mathcal{N}(1, 1)$ ,  $X_2 \sim \mathcal{N}(0, 1)$ , and

if  $\theta = 1$  :  $X_1 \sim \mathcal{N}(0, 1)$ ,  $X_2 \sim \mathcal{N}(1, 1)$ .

In both cases  $X_1, X_2$  are independent.

Is  $T(X) = X_1 - X_2$  a sufficient statistic of  $X$  for  $\theta$ ? Support your answer with a proof.

10. *Scenario:*  $M = (X_1, \dots, X_n) \in \mathcal{X}^n$  where  $X_i$  are i.i.d discrete random variables each with probability mass function  $p_X$ . We now describe two (deterministic) functions:

(i)  $f$  maps  $M$  into a string of  $\ell$  binary digits  $C = f(M) \in \{0, 1\}^\ell$ , i.e.,  $f((x_1, \dots, x_n)) \in \{0, 1\}^\ell$ .

(ii)  $g$  is a function used for decoding  $M$  from  $C$ , i.e.,  $g(C) = g(f(M)) \in \mathcal{X}^n$  is an estimate of  $M$ .

Let  $P_e = P[g(C) \neq M]$  denote the probability of detection error.

*Question:* Assuming that  $P_e \leq \epsilon$  where  $0 < \epsilon < 0.5$ , obtain a good lower bound on the ratio  $\frac{\ell}{n}$  for all sufficiently large  $n$ . Your lower bound must be independent of the functions  $f$  and  $g$ .

*Bonus: (3 marks)* If  $n$  is allowed grow, and  $\ell$  also grows with  $n$ , and  $P_e \leq \epsilon$  for all  $n$ , find a lower bound on  $\lim_{\epsilon \rightarrow 0} (\lim_{n \rightarrow \infty} \frac{\ell}{n})$ , assuming that this limit exists. Can you identify functions  $f$  and  $g$  that attain this limit?