

# ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

School of Computer and Communication Sciences

**Handout 15**  
Midterm exam

Information Theory and Coding  
Oct. 29, 2024

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3 problems, 43 points, 180 minutes.  
1 sheet (2 pages) of notes allowed.

Good Luck!

PLEASE WRITE YOUR NAME ON EACH SHEET OF YOUR ANSWERS.

PLEASE WRITE THE SOLUTION OF EACH PROBLEM ON A SEPARATE SHEET.

(All logarithms are taken to the base 2.)

PROBLEM 1. (16 points)

Suppose  $f : [0, \infty) \rightarrow \mathbb{R} \cup \{\pm\infty\}$  is a decreasing, convex function, and  $p$  and  $q$  are probability distributions on an alphabet  $\mathcal{U}$  (i.e.,  $p(u) \geq 0$  and  $\sum_{u \in \mathcal{U}} p(u) = 1$ , similarly for  $q$ ). Define

$$K_f(p, q) = \sum_{u: p(u) > 0} p(u) f\left(\frac{q(u)}{p(u)}\right).$$

- (a) (2 pts) Show that  $K_f(p, q) \geq f(1)$ , and equality happens if  $q = p$ .

*Hint:* Make sure to use convexity.

Suppose  $U$  is a random variable with distribution  $p$ . A “prediction” about  $U$  is a probability distribution  $q$  on  $\mathcal{U}$  — basically saying “I believe we will see the value  $u$  with probability  $q(u)$ ”. A prediction  $q$  is assigned a score via  $\text{score}(q) = \mathbb{E}\left[\frac{1}{q(U)}\right] = \sum_u \frac{p(u)}{q(u)}$ .

- (b) (3 pts) Let  $p_{1/2}(u) = \frac{p(u)^{1/2}}{A}$ , where  $A = \sum_u p(u)^{1/2}$  to ensure that  $p_{1/2}$  is a probability distribution. Show that for any probability distribution  $q$ ,  $\text{score}(q) \geq A^2$ , with equality if  $q = p_{1/2}$ .

*Hint:* First show that with  $f(x) = 1/x$ ,  $\text{score}(q) = A^2 K_f(p_{1/2}, q)$ .

- (c) (3 pts) Suppose  $c : \mathcal{U} \rightarrow \{0, 1\}^*$  is a uniquely decodable code. Show that  $\mathbb{E}[2^{\text{length}(c(U))}] \geq A^2$ .

Fix  $\alpha > 0$ .

- (d) (3 pts) Replace the score function above with  $\text{score}_\alpha(q) = \mathbb{E}[q(U)^{-\alpha}] = \sum_u \frac{p(u)}{q(u)^\alpha}$ . Show that for any  $q$ ,  $\text{score}_\alpha(q) \geq (A_{1/1+\alpha})^{1+\alpha}$ , with equality if  $q = p_{1/1+\alpha}$ , where we define  $p_s(u) = p(u)^s / A_s$  where  $A_s = \sum_u p(u)^s$ .

*Hint:* Choose  $f$  appropriately and express  $\text{score}_\alpha(q)$  in terms of  $K_f(p_s, q)$  for some  $s$ .

- (e) (2 pts) Show that for any uniquely decodable code  $c : \mathcal{U} \rightarrow \{0, 1\}^*$ ,

$$\mathbb{E}[2^{\alpha \text{length}(c(U))}] \geq (A_{1/1+\alpha})^{1+\alpha}.$$

- (f) (3 pts) Show that there exists a prefix-free code  $c : \mathcal{U} \rightarrow \{0, 1\}^*$  such that

$$\mathbb{E}[2^{\alpha \text{length}(c(U))}] \leq 2^\alpha (A_{1/1+\alpha})^{1+\alpha}.$$

PROBLEM 2. (12 points)

For this problem, we define the following notation different from that used in the lectures. Fix a natural number  $n$ . Let  $(X_1, \dots, X_n)$  be a vector of binary random variables, with each  $X_i$  taking values in  $\{0, 1\}$ . For  $i, j = 1, \dots, n$ , let  $X_i^j = (X_i, \dots, X_j)$  if  $i \leq j$  and empty if  $i > j$ . Let  $X_{\neq i}$  denote the vector  $X_1^n$  without the  $i^{\text{th}}$  element, i.e.,  $X_{\neq i} = (X_1^{i-1}, X_{i+1}^n)$ . Also let  $X_{\bar{i}}$  denote the vector  $X_1^n$  with its  $i^{\text{th}}$  element flipped, i.e.,  $X_{\bar{i}} = (X_1^{i-1}, 1 - X_i, X_{i+1}^n)$ .

- (a) (3 pts) Show that  $\sum_{i=1}^n H(X_i | X_{\neq i}) \leq H(X_1^n)$ .

Let  $A$  be a subset of  $\{0, 1\}^n$ , i.e.,  $A$  consists of binary vectors of length  $n$ . Denote by  $E(A)$  the set of pairs of vectors in  $A$  that differ at *exactly* one position, i.e.,

$$\begin{aligned} E(A) &= \{(x_1^n, \tilde{x}_1^n) \in A \times A \text{ such that } \tilde{x}_i \neq x_i \text{ for exactly one } i\} \\ &= \{(x_1^n, \tilde{x}_1^n) \in A \times A \text{ such that } \tilde{x}_1^n = x_{\bar{i}} \text{ for some } i\}. \end{aligned}$$

Let  $(X_1, \dots, X_n)$  be randomly and uniformly chosen from  $A$ .

- (b) (3 pts) Fix  $x_1^n \in A$ . Compute  $H(X_i | X_{\neq i} = x_{\neq i})$ .

*Hint:* Consider two cases:  $x_{\bar{i}} \in A$  and  $x_{\bar{i}} \notin A$ .

- (c) (3 pts) Show that  $H(X_i | X_{\neq i}) = \frac{1}{|A|} \sum_{x_1^n \in A} \mathbb{1}\{x_{\bar{i}} \in A\}$ .

- (d) (3 pts) Show that  $\sum_{i=1}^n H(X_i | X_{\neq i}) = \frac{|E(A)|}{|A|}$  and conclude that  $|E(A)| \leq |A| \log |A|$ .

*Hint:* Use (a).

PROBLEM 3. (15 points)

- (a) (2 pts) Suppose  $p$  is a probability distribution on  $\mathcal{U}$ . Show that for any probability distribution  $q$  on  $\mathcal{U}$ ,  $\max_{u \in \mathcal{U}} \log \frac{p(u)}{q(u)} \geq 0$ . Additionally, show that  $\min_q \max_{u \in \mathcal{U}} \log \frac{p(u)}{q(u)} = 0$ , where the minimization is over all probability distributions  $q$  on  $\mathcal{U}$ .
- (b) (2 pts) Show that  $\min_q \max_{u \in \mathcal{U}} \log \frac{f(u)}{q(u)} = \log K$ , where  $K = \sum_u f(u)$  for a nonnegative function  $f$ .  
*Hint:* Use (a).

Suppose from now on that for every  $\theta$  in some parameter set  $\Theta$ , we have a probability distribution  $p_\theta$  on  $\mathcal{U}$ .

- (c) (2 pts) Show that  $\min_q \max_{u \in \mathcal{U}, \theta \in \Theta} \log \frac{p_\theta(u)}{q(u)} = S$ , where  $S = \log \sum_{u \in \mathcal{U}} \max_{\theta \in \Theta} p_\theta(u)$ .  
*Hint:* Use (b).
- (d) (3 pts) Suppose we do not know the probability distribution of a random variable  $U$ , except that the distribution is one of the  $p_\theta$  above. Show that there is a prefix-free code  $c : \mathcal{U} \rightarrow \{0, 1\}^*$  such that, for every  $\theta \in \Theta$  and every  $u \in \mathcal{U}$ ,  $\text{length } c(u) \leq \log \frac{1}{p_\theta(u)} + S + 1$ , where  $S$  is as in part (c) above.

Suppose we know that  $U_1, U_2, \dots$ , are i.i.d. Bernoulli( $\theta$ ) random variables, but we do not know the value of  $\theta \in [0, 1]$ . For  $u^n \in \{0, 1\}^n$ , define  $p_\theta(u^n) = \theta^{k(u^n)}(1 - \theta)^{n - k(u^n)}$ , where  $k(u^n)$  is the number of 1's in the sequence  $(u_1, \dots, u_n)$ . With this definition,  $\Pr(U^n = u^n) = p_\theta(u^n)$ .

- (e) (3 pts) Show that for any  $u^n$ , we have  $\max_{\theta \in [0, 1]} p_\theta(u^n) = \left(\frac{k}{n}\right)^k \left(1 - \frac{k}{n}\right)^{n-k}$  with  $k = k(u^n)$ , and conclude that

$$\sum_{u^n \in \{0, 1\}^n} \max_{\theta \in [0, 1]} p_\theta(u^n) = \sum_{i=0}^n \binom{n}{i} \left(\frac{i}{n}\right)^i \left(1 - \frac{i}{n}\right)^{n-i}.$$

*Hint:* Differentiate  $\log p_\theta(u^n)$  with respect to  $\theta$ .

- (f) (3 pts) Show that for each  $n$ , there is a prefix-free code  $c_n : \{0, 1\}^n \rightarrow \{0, 1\}^*$  such that, for every  $\theta \in [0, 1]$  and every  $u^n \in \{0, 1\}^n$ ,

$$\text{length } c_n(u^n) \leq \log \frac{1}{p_\theta(u^n)} + \log(1 + n) + 1.$$

*Hint:* Use (d) and (e).