

# ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

School of Computer and Communication Sciences

**Handout 23**  
Homework 10

Information Theory and Coding  
Nov. 25, 2024

PROBLEM 1. Suppose  $U$  is  $\{0, 1\}$  valued with  $\mathbb{P}(U = 0) = \mathbb{P}(U = 1) = 1/2$ . Suppose we have a distortion measure  $d$  given by

$$d(u, v) = \begin{cases} 0, & u = v \\ 1, & (u, v) = (1, 0) \\ \infty, & (u, v) = (0, 1) \end{cases}$$

i.e., we never want to represent a 0 with a 1. Find  $R(D)$ .

PROBLEM 2. Suppose  $\mathcal{U} = \mathcal{V}$  are additive groups with group operation  $\oplus$ . (E.g.,  $\mathcal{U} = \mathcal{V} = \{0, \dots, K-1\}$ , with modulo  $K$  addition.) Suppose the distortion measure  $d(u, v)$  depends only on the difference between  $u$  and  $v$  and is given by  $g(u \ominus v)$ . Let  $\phi(D)$  denote  $\max\{H(Z) : E[g(Z)] \leq D\}$ .

a) Show that  $\phi(D)$  is concave.

b) Let  $(U, V)$  be such that  $E[d(U, V)] \leq D$ . Show that  $I(U; V) \geq H(U) - \phi(D)$  by justifying

$$I(U; V) = H(U) - H(U|V) = H(U) - H(U \ominus V|V) \geq H(U) - H(U \ominus V) \geq H(U) - \phi(D).$$

c) Show that  $R(D) \geq H(U) - \phi(D)$ .

d) Assume now that  $U$  is uniform on  $\mathcal{U}$ . Show that  $R(D) = H(U) - \phi(D)$ .

PROBLEM 3. Suppose  $\mathcal{U} = \mathcal{V} = \mathbb{R}$ , the set of real numbers, and  $d(u, v) = (u - v)^2$ .

(a) Show that for any  $U$  with variance  $\sigma^2$ ,  $R(D)$  satisfies

$$h(U) - \frac{1}{2} \log(2\pi e D) \leq R(D).$$

(b) Show that  $R(D)$  does not depend on the mean of  $U$ .

Now, assume without loss of generality that  $U$  is zero-mean. Suppose we have access to a noisy observation  $V$  of  $U$  through the channel  $U + Z = V$ , where  $Z \sim \mathcal{N}(0, \sigma_Z^2)$  and independent of  $U$ . We reconstruct  $U$  via a linear estimator  $\hat{U} = aV + b$ .

(c) Show that  $a = \frac{\sigma^2}{\sigma^2 + \sigma_Z^2}$  and  $b = 0$  minimizes  $E[(U - \hat{U})^2]$  and for such choice of  $a, b$ ,  

$$E[(U - \hat{U})^2] = \sigma^2 \frac{\sigma_Z^2}{\sigma^2 + \sigma_Z^2}.$$

(d) For the channel above, show that

$$I(U; V) \leq \frac{1}{2} \log\left(1 + \frac{\sigma^2}{\sigma_Z^2}\right)$$

(e) Show that for  $D \leq \sigma^2$

$$R(D) \leq \frac{1}{2} \log(\sigma^2/D).$$

[Hint: Use the channel above for a candidate  $p_{V|U}$ .]

PROBLEM 4. Given finite alphabets  $\mathcal{X}$  and  $\mathcal{Y}$ , a distribution  $p_{XY}$ ,  $0 < \epsilon < \epsilon'$ , and a sequence  $x^n \in T(n, p_X, \epsilon)$ , consider a random vector  $Y^n$  with independent components with  $\Pr(Y_i = y) = p_{Y|X}(y|x_i)$ .

For  $x \in \mathcal{X}$ , let  $J(x) = \{i : x_i = x\}$ . For an  $x \in \mathcal{X}$  and  $y \in \mathcal{Y}$ , let  $N(x, y) = \sum_i \mathbb{1}\{(x_i, Y_i) = (x, y)\} = \sum_{i \in J(x)} \mathbb{1}\{Y_i = y\}$ .

- (a) Show that for each  $x$  and  $y$ ,  $np(x, y)(1 - \epsilon) \leq E[N(x, y)] \leq np(x, y)(1 + \epsilon)$ , and  $\text{Var}(N(x, y))$  is at most  $n$ . [Hint: don't forget that  $x^n$  is in  $T(n, p_X, \epsilon)$ .]
- (b) Show that for each  $x$  and  $y$ , both  $\Pr(N(x, y) < np(x, y)(1 - \epsilon'))$  and  $\Pr(N(x, y) > np(x, y)(1 + \epsilon'))$  approach to zero as  $n$  gets large. Would this be true if we had not assumed  $\epsilon < \epsilon'$ ?
- (c) Using (a) and (b) show that  $\Pr((x^n, Y^n) \notin T(n, p_{XY}, \epsilon'))$  approaches 0 as  $n$  gets large.
- (d) Suppose now we have a distribution  $p(u, x, y)$  where  $p(y|ux) = p(y|x)$ . [In other words,  $U, X, Y$  form a Markov chain.] Suppose  $(u^n, x^n)$  is in  $T(n, p_{UX}, \epsilon)$ , and  $Y^n$  has independent components as above. What can we say about  $\Pr((u^n, x^n, Y^n) \in T(n, p_{UXY}, \epsilon'))$ ?