## ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

School of Computer and Communication Sciences

Handout 31	Information Theory and Coding
Homework 13	Dec. 18, 2023

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.,  $\mathcal{U} = \mathcal{V} = \{0, \ldots, K-1\}$ , with modulo K addition.) Suppose the distortion measure d(u, v) depends only on the difference between u and v and is given by  $g(u \oplus v)$ . 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

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

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

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

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

(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

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

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

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

[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 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  $\Pr\left((x^n, Y^n) \notin T(n, p_{XY}, \epsilon')\right)$  approaches 0 as 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  $\Pr\left((u^n, x^n, Y^n) \in T(n, p_{UXY}, \epsilon')\right)$ ?

PROBLEM 5. Consider a two-way communication system where two parties communicate via a *common* output they both can observe and influence. Denote the common output by Y, and the signals emitted by the two parties by  $x_1$  and  $x_2$  respectively. Let  $p(y|x_1, x_2)$ model the memoryless channel through which the two parties influence the output.

We will consider feedback-free block codes, i.e., we will use encoding and decoding functions of the form

enc<sub>1</sub>: 
$$\{1, ..., 2^{nR_1}\} \to \mathcal{X}_1^n$$
 dec<sub>1</sub>:  $\mathcal{Y}^n \times \{1, ..., 2^{nR_1}\} \to \{1, ..., 2^{nR_2}\}$   
enc<sub>2</sub>:  $\{1, ..., 2^{nR_2}\} \to \mathcal{X}_2^n$  dec<sub>2</sub>:  $\mathcal{Y}^n \times \{1, ..., 2^{nR_2}\} \to \{1, ..., 2^{nR_1}\}$ 

with which the parties encode their own message and decode the other party's messages. (Note that when a party is decoding the other party's message, it can make use of the knowledge of its own message).

We will say that the rate pair  $(R_1, R_2)$  is achievable, if for any  $\epsilon > 0$ , there exist encoders and decoders with the above form for which the average error probability is less than  $\epsilon$ .

Consider the following 'random coding' method to construct the encoders:

- (i) Choose probability distributions  $p_j$  on  $\mathcal{X}_j$ , j = 1, 2.
- (ii) Choose  $\{ enc_1(m_1)_i : m_1 = 1, \ldots, 2^{nR_1}, i = 1, \ldots, n \}$  i.i.d., each having distribution as  $p_1$ . Similarly, choose  $\{ enc_2(m_2)_i : m_2 = 1, \ldots, 2^{nR_2}, i = 1, \ldots, n \}$  i.i.d., each having distribution as  $p_2$ , independently of the choices for  $enc_1$ .

For the decoders we will use typicality decoders:

- (i) Set  $p(x_1, x_2, y) = p_1(x_1)p_2(x_2)p(y|x_1, x_2)$ . Choose a small  $\epsilon > 0$  and consider the set T of  $\epsilon$ -typical  $(x_1^n, x_2^n, y^n)$ 's with respect to p.
- (ii) For decoder 1: given  $y^n$  and the correct  $m_1$ , dec<sub>1</sub> will declare  $\hat{m}_2$  if it is the unique  $m_2$  for which  $(\text{enc}_1(m_1), \text{enc}_2(m_2), y^n) \in T$ . If there is no such  $m_2$ , dec<sub>1</sub> outputs 0. (Similar description applies to Decoder 2.)

- (a) Given that  $m_1$  and  $m_2$  are the transmitted messages, show that  $(\text{enc}_1(m_1), \text{enc}_2(m_2), Y^n) \in T$  with high probability.
- (b) Given that  $m_1$  and  $m_2$  are the transmitted messages, and  $\tilde{m}_1 \neq m_1$  what is the probability distribution of  $(\text{enc}_1(\tilde{m}_1), \text{enc}(m_2), Y^n)$ ?
- (c) Under the assumptions in (b) show that the

 $\Pr\{(\mathrm{enc}_1(\tilde{m}_1), \mathrm{enc}_2(m_2), Y^n) \in T\} \doteq 2^{-nI(X_1; X_2Y)}.$ 

(d) Show that all rate pairs satisfying

$$R_1 \le I(X_1; YX_2), \quad R_2 \le I(X_2; YX_1)$$

for some  $p(x_1, x_2) = p(x_1)p(x_2)$  are achievable.

(e) For the case when  $X_1$ ,  $X_2$ , Y are all binary and Y is the product of  $X_1$  and  $X_2$ , show that the achievable region is strictly larger than what we can obtain by 'half duplex communication' (i.e., the set of rates that satisfy  $R_1 + R_2 \leq 1$ .)