

**Exercise 5.1 Min-entropy and Shannon entropy as Rényi entropies [EE5139]**

Both the min-entropy and the Shannon entropy are limiting cases of the following family of Rényi entropies:

$$H_\alpha(X) = \frac{1}{1-\alpha} \log \sum_x P(x)^\alpha, \quad \alpha \in (0, 1) \cup (1, +\infty). \quad (1)$$

- a.) To verify this, compute the limit of the above quantities for  $\alpha \rightarrow \{0_+, 1, +\infty\}$ . (Here, by saying  $\alpha \rightarrow 0_+$ , we mean  $\alpha$  “approaching 0 from right-hand side”.)
- b.) Plot the Rényi entropy as a function of  $\alpha$  for the random variable  $X$  distributed as

$x$	0	1	2
$P(x)$	1/2	1/4	1/4

- c.) Show that, for any random variable  $X \in \mathcal{X}$  and any pmf  $P(x)$ , the Rényi entropy is monotonically non-increasing in the parameter  $\alpha$ . Argue how this yields an alternative proof of the fact that  $H_{\min}(X) \leq H(X) \leq \log |\mathcal{X}|$ .
- d.) Compute the min-entropy  $H_{\min}(X|Y)$  of the joint random variables  $(X, Y)$  distributed as

		$X$		
		0	1	2
$Y$	0	1/6	1/12	1/12
	1	1/12	1/6	1/12
	2	1/12	1/12	1/6

**Exercise 5.2 Distributions with a large entropy gap [all]**

It is possible to construct distributions that have a large gap between min-entropy and Shannon entropy. This shows that controlling the Shannon entropy or the mutual information is not sufficient for most cryptographic tasks.

- a.) Given  $\epsilon \in (0, 1)$ , construct a sequence of random variables  $(X_2, X_3, \dots, X_n, \dots)$  where  $X_n \in \{0, 1, \dots, n-1\}$ , such that

$$\left. \begin{aligned} H(X_n) &\geq (1-\epsilon) \log n \\ H_{\min}(X_n) &= C, \end{aligned} \right\} \forall n \geq N$$

for some  $N \in \mathbb{N}$  and some constant  $C > 0$ .

- b.) Given  $\epsilon \in (0, 1)$ , construct a sequence of random variables  $((X_2, Y_2), (X_3, Y_3), \dots, (X_n, Y_n), \dots)$ , where  $X_n, Y_n \in \{0, 1, \dots, n-1\}$ , such that

$$\left. \begin{aligned} H(X_n) &= H_{\min}(X_n) = \log n \\ H(X_n|Y_n) &\geq (1-\epsilon) \log n \\ H_{\min}(X_n|Y_n) &= C \end{aligned} \right\} \forall n \geq N$$

for some  $N \in \mathbb{N}$  and some constant  $C > 0$ .

### Exercise 5.3 Typical sets [all]

Consider a DMS with a two symbol alphabet  $\{a, b\}$  where  $p_X(a) = 2/3$  and  $p_X(b) = 1/3$ . Let  $X^n = (X_1, \dots, X_n)$  be a string of symbols emitted by the source with  $n = 100,000$ . Let  $W(X_j)$  be the surprisal for the  $j$ -th source output, i.e.,  $W(X_j) = -\log 2/3$  for  $X_j = a$  and  $-\log 1/3$  for  $X_j = b$ . Define  $W(X^n) = \sum_{j=1}^n W(X_j)$ .

- Find the variance of  $W(X_j)$ . For  $\epsilon = 0.01$ , evaluate a bound on the probability of the typical set  $\mathcal{A}_\epsilon^{(n)}$  using Chebyshev's inequality.
- Let  $N_a$  be the number of  $a$ 's in the string  $X^n = (X_1, \dots, X_n)$ . The random variable (rv)  $N_a$  is the sum of  $n$  iid rv's. Show what these rv's are.
- Express the rv  $W(X^n)$  as a function of the rv  $N_a$ . Note how this depends on  $n$ .
- Express the typical set in terms of bounds on  $N_a$ . Use Chebyshev's inequality to derive bounds on the probability of the typical set, using properties of  $N_a$  instead of  $W(X_j)$ .  
**Hint:** You may write  $\mathcal{A}_\epsilon^{(n)} = \{x^n : \alpha < N_a < \beta\}$  and calculate  $\alpha$  and  $\beta$ .
- Find  $\Pr(N_a = i)$  for  $i = 0, 1, 2$ . Find the probability of each individual string  $x^n$  for those values of  $i$ . Find the particular string  $x^n$  that has maximum probability over all sample values of  $X^n$ . What are the next most probable  $n$ -strings. Give a brief discussion of why the most probable  $n$ -strings are not regarded as typical strings.

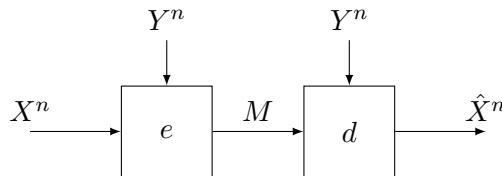
### Exercise 5.4 Source coding with side information [EE5139]

Consider a memoryless source  $(\mathbf{X}, \mathbf{Y})$  that produces in each iteration two random variables,  $X_i$  and  $Y_i$ , where  $X_i$  is private information and  $Y_i$  is public information. The pairs  $(X_i, Y_i)$  follow a joint distribution  $P_{XY}$  and are i.i.d.. We are looking for a fixed-length block code that compresses the private information  $X^n = (X_1, X_2, \dots, X_n)$  using the public information  $Y^n = (Y_1, Y_2, \dots, Y_n)$  such that the code can be decoded asymptotically error-free with help of the public information.

An  $(n, 2^L)$ -code for such a source is given by an encoder,  $e : (X^n, Y^n) \rightarrow M$ , and decoder,  $d : (M, Y^n) \rightarrow \hat{X}^n$ , as illustrated in the figure below. The codeword  $M \in \{0, 1\}^L$  is a binary string of length  $L$ . We define  $R^*(\mathbf{X}|\mathbf{Y})$  as the infimum over all rates  $R$  such that there exists a sequence of  $(n, 2^{nR})$ -codes satisfying

$$\lim_{n \rightarrow \infty} \Pr[X^n \neq \hat{X}^n] = 0, \quad \text{where} \quad \hat{X}^n = d_n(e_n(X^n, Y^n), Y^n) \quad (2)$$

is a function of both  $X^n$  and  $Y^n$ . We want to establish that  $R^*(\mathbf{X}|\mathbf{Y}) = H(X|Y)$ .



- Determine  $R^*(\mathbf{X}|\mathbf{Y})$ , by intuitive or formal arguments, for the simple cases where
  - $X$  and  $Y$  are independent,
  - $X = Y$ ,
- By explicitly constructing a code for the source  $(X, Y)$  using codes for the sources  $Y$  and  $X$  (with side information  $Y$ ), show that  $R^*(\mathbf{X}, \mathbf{Y}) \leq R^*(\mathbf{X}|\mathbf{Y}) + R^*(\mathbf{Y})$ .

- c.) Show that the converse,  $R^*(\mathbf{X}|\mathbf{Y}) \geq H(X|Y)$  using Fano's inequality. **Hint:** You will also need the following sequence of inequalities, which needs to be verified.

$$H(X^n|\hat{X}^n) \geq H(X^n|Y^n M) \quad (3)$$

$$= H(X^n M|Y^n) - H(M|Y^n) \quad (4)$$

$$\geq H(X^n M|Y^n) - L \quad (5)$$

$$\geq H(X^n|Y^n) - L. \quad (6)$$

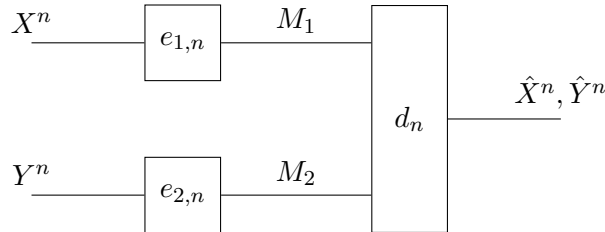
- d.) Give a formal proof or a sketch of a proof that  $R^*(\mathbf{X}|\mathbf{Y}) \leq H(X|Y)$ . **Hint:** Consider the typical set

$$\mathcal{A}_\epsilon^{(n)}(\mathbf{X}|\mathbf{Y}) := \left\{ (x^n, y^n) \in \mathcal{X}^n \times \mathcal{Y}^n : \left| \frac{1}{n} \log \frac{1}{P_{X^n|Y^n}(x^n|y^n)} - H(X|Y) \right| \leq \epsilon \right\}. \quad (7)$$

### Exercise 5.5 Achievability for the Slepian–Wolf coding problem [EE6139]

We return to the Exercise 4.4 from the last homework. Let  $X$  and  $Y$  be a pair of jointly distributed random variables. ( $X$  is distributed on finite set  $\mathcal{X}$ , and  $Y$  is distributed on finite set  $\mathcal{Y}$ .) An  $(n, 2^{nL_1}, 2^{nL_2})$ -separately-encoded-jointly-decoded source code consists of a pair of encoders  $e_1, e_2$ , and a decoder  $d$ , where

- $e_1 : \mathcal{X}^n \rightarrow \{0, 1\}^{nL_1}$ ,
- $e_2 : \mathcal{Y}^n \rightarrow \{0, 1\}^{nL_2}$ , and
- $d : \{0, 1\}^{nL_1} \times \{0, 1\}^{nL_2} \rightarrow \mathcal{X}^n \times \mathcal{Y}^n$ .



The rate pair  $(R_1, R_2)$  is said to be achievable for DMS  $(X, Y)$  if there exists a sequence of  $(n, 2^{nL_1}, 2^{nL_2})$ -codes with encoders  $e_{1,n}, e_{2,n}$  and decoder  $d_n$  such that

$$\lim_{n \rightarrow \infty} P\{(\hat{X}^n, \hat{Y}^n) \neq (X^n, Y^n)\} = 0$$

where

$$(\hat{X}^n, \hat{Y}^n) = d_n(M_1, M_2), \quad M_1 = e_{1,n}(X^n), \quad \text{and} \quad M_2 = e_{2,n}(Y^n)$$

are the reconstructed source and codewords respectively.

This time, we are interested in the achievability of the problem.

- a.) **An alternative for typical sequences** Given  $n \in \mathbb{N}$  and  $\epsilon \in (0, 1)$ , we define the set of  $Y$ -sequences as

$$\mathcal{T}_\epsilon^{(n)}(Y) \triangleq \left\{ \mathbf{y} \in \mathcal{Y}^n : \left| \frac{\sum_{i=1}^n \delta_{y, y_i}}{n} - p_Y(y) \right| < \left\lceil \sqrt{\frac{|\mathcal{Y}|}{\epsilon}} \right\rceil \sqrt{\frac{p_Y(y)(1 - p_Y(y))}{n}} \quad \forall y \in \mathcal{Y} \right\}.$$

Show that

$$\text{i.) } P[Y^n \in \mathcal{T}_\epsilon^{(n)}(Y)] \geq 1 - \epsilon$$

ii.) There exists some  $A > 0$  independent from  $n$  and  $\epsilon$  such that

$$2^{-nH(Y)-A\sqrt{n/\epsilon}} < p_{Y^n}(\mathbf{y}) < 2^{-nH(Y)+A\sqrt{n/\epsilon}}$$

for all  $\mathbf{y} \in \mathcal{T}_\epsilon^{(n)}(Y)$ .

iii.)  $\lim_{n \rightarrow \infty} \frac{1}{n} \log_2 |\mathcal{T}_\epsilon^{(n)}(Y)| = H(Y)$ .

b.) **Position-based coding** Given positive integer  $n$  and  $\epsilon > 0$ , let  $M_X \triangleq \lfloor 2^{n(H(Y|X)+\epsilon)} \rfloor$ , and let  $M$  be another positive integer. Let  $\{\mathbf{X}_{i,j}\}_{i,j}$  be a set of i.i.d. random variables on  $\mathcal{X}^n$ , where  $i \in \{1, \dots, M_X\}$ ,  $j \in \{1, \dots, M\}$ , and

$$p_{\mathbf{X}_{i,j}}(\mathbf{x}) = \prod_{k=1}^n p_X(x_k)$$

for each  $(i, j)$ .

i.) Suppose  $I(X, Y) > \frac{1}{2}\epsilon$ , and let  $M = \lfloor 2^{n(I(X,Y)-\frac{1}{2}\epsilon)} \rfloor$ . Prove that, for  $n$  large enough,

$$P[X^n \neq \mathbf{X}_{i,j} \ \forall (i, j)] < 2\epsilon.$$

ii.) Let  $A$  and  $B$  be a pair of random variable denoting the “smallest” indices  $a, b$  such that that  $X^n = \mathbf{X}_{a,b}$ . Namely,

$$p_{A,B|X^n, \{\mathbf{X}_{i,j}\}}(a, b | \mathbf{x}, \{\mathbf{x}_{i,j}\}) = \begin{cases} 1 & \text{if } \mathbf{x} = \mathbf{x}_{a,b} \ \forall i < a \\ & \mathbf{x} \neq \mathbf{x}_{i,j} \ \forall i < a \\ & \mathbf{x} \neq \mathbf{x}_{a,j} \ \forall j < b \\ 0 & \text{otherwise} \end{cases}.$$

We take the convention that  $(A, B) = (\infty, \infty)$  if  $X^n \neq \mathbf{X}_{i,j}$  for all  $i, j$ . Prove that

$$P[A < \infty, B < \infty, p_{Y^n|X^n}(Y^n | \mathbf{X}_{A,j}) \geq p_{Y^n|X^n}(Y^n | X^n) \ \exists j \neq B] < \epsilon$$

for  $n$  large enough.

c.) Based on the arguments in a.) and b.), show that, for any  $\delta_1, \delta_2 > 0$ , the following rates are achievable

$$R_1 = H(X|Y) + \delta_1, \tag{8}$$

$$R_2 = H(Y) + \delta_2. \tag{9}$$

d.) Show that any  $(R_1, R_2)$  satisfying the following inequalities are achievable

$$R_1 > H(X|Y), \tag{10}$$

$$R_2 > H(Y|X), \tag{11}$$

$$R_1 + R_2 > H(X, Y). \tag{12}$$