

# EE276: Homework #2

Due on Friday Jan 23, 6pm - Gradescope entry code: E6VP4X

## 1. Data Processing Inequality.

The random variables  $X$ ,  $Y$  and  $Z$ , belonging to alphabets  $\mathcal{X}$ ,  $\mathcal{Y}$ , and  $\mathcal{Z}$  respectively, form a Markov triplet  $(X - Y - Z)$  if  $p(z|y) = p(z|y, x)$ , or, equivalently, if  $p(x|y) = p(x|y, z)$ . If  $X$ ,  $Y$ ,  $Z$  form a Markov triplet  $(X - Y - Z)$ , show that:

- (a)  $H(X|Y) = H(X|Y, Z)$  and  $H(Z|Y) = H(Z|X, Y)$
- (b)  $H(X|Y) \leq H(X|Z)$
- (c)  $I(X; Y) \geq I(X; Z)$  and  $I(Y; Z) \geq I(X; Z)$
- (d)  $I(X; Z) \leq \log |\mathcal{Y}|$
- (e)  $I(X; Z|Y) = 0$

where the *conditional mutual information* of random variables  $X$  and  $Y$  given  $Z$  is defined by

$$\begin{aligned} I(X; Y|Z) &= H(X|Z) - H(X|Y, Z) \\ &= \sum_{x,y,z} p(x, y, z) \log \frac{p(x, y|z)}{p(x|z)p(y|z)} \end{aligned}$$

## 2. Conditional mutual information vs. unconditional mutual information.

Give examples of joint random variables  $X$ ,  $Y$  and  $Z$  such that

- (a)  $I(X; Y | Z) < I(X; Y)$ ,
- (b)  $I(X; Y | Z) > I(X; Y)$ .

## 3. Prefix and Uniquely Decodable codes

Consider the following code:

$u$	Codeword
a	1 0
b	0 0
c	1 1
d	1 1 0

- (a) Is this a Prefix code?
- (b) Argue that this code is uniquely decodable, by describing an algorithm for the decoding.

4. **Relative entropy and the cost of miscoding.** Let the random variable  $X$  be defined on  $\{1, 2, 3, 4, 5, 6\}$  according to pmf  $p$ . Let  $p$  and another pmf  $q$  be

Symbol	$p(x)$	$q(x)$	$C_1(x)$	$C_2(x)$
1	1/2	1/2	0	0
2	1/8	1/4	100	10
3	1/8	1/16	101	1100
4	1/8	1/16	110	1101
5	1/16	1/16	1110	1110
6	1/16	1/16	1111	1111

- (a) Calculate  $H(X)$ ,  $D(p||q)$  and  $D(q||p)$ .
- (b) The last two columns above represent codes for the random variable. Verify that codes  $C_1$  and  $C_2$  are optimal under the respective distributions  $p$  and  $q$ .
- (c) Now assume that we use  $C_2$  to code  $X$ . What is the average length of the codewords? By how much does it exceed the entropy  $H(X)$ , i.e., what is the redundancy of the code?
- (d) What is the redundancy if we use code  $C_1$  for a random variable  $Y$  with pmf  $q$ ?
5. **Shannon code.** Consider the following method for generating a code for a random variable  $X$  which takes on  $m$  values  $\{1, 2, \dots, m\}$  with pmf  $p$  having probabilities  $p_1, p_2, \dots, p_m$ . Assume that the probabilities are ordered so that  $p_1 \geq p_2 \geq \dots \geq p_m$ . Define

$$F_i = \sum_{k=1}^{i-1} p_k,$$

i.e. the sum of the probabilities of all symbols less than  $i$ . Then the codeword for  $i$  is the number  $F_i \in [0, 1]$  rounded off to  $l_i$  bits, where  $l_i = \lceil \log \frac{1}{p_i} \rceil$ .

- (a) Show that the code constructed by this process is prefix-free and the average length satisfies

$$H(X) \leq L < H(X) + 1.$$

- (b) Construct the code for the probability distribution  $(0.5, 0.25, 0.125, 0.125)$ .
- (c) Now, suppose the code in (a) is used on a random variable  $\tilde{X}$  taking values in  $\{1, 2, \dots, m\}$  distributed with pmf  $q$  having probabilities  $q_1, q_2, \dots, q_m$ . Show that the average length satisfies

$$H(\tilde{X}) + D(q||p) \leq L < H(\tilde{X}) + D(q||p) + 1$$

6. **AEP.** Let  $X_i$  for  $i \in \{1, \dots, n\}$  be an i.i.d. sequence from the p.m.f.  $p(x)$  with alphabet  $\mathcal{X} = \{1, 2, \dots, m\}$ . Denote the expectation and entropy of  $X$  by  $\mu := \mathbb{E}[X]$  and  $H := -\sum p(x) \log p(x)$  respectively.

For  $\epsilon > 0$ , recall the definition of the typical set

$$A_\epsilon^{(n)} = \left\{ x^n \in \mathcal{X}^n : \left| -\frac{1}{n} \log p(x^n) - H \right| \leq \epsilon \right\}$$

and define the following set

$$B_\epsilon^{(n)} = \left\{ x^n \in \mathcal{X}^n : \left| \frac{1}{n} \sum_{i=1}^n x_i - \mu \right| \leq \epsilon \right\}.$$

In what follows,  $\epsilon > 0$  is fixed.

- (a) Does  $\mathbb{P} \left( X^n \in A_\epsilon^{(n)} \right) \rightarrow 1$  as  $n \rightarrow \infty$ ?
- (b) Does  $\mathbb{P} \left( X^n \in A_\epsilon^{(n)} \cap B_\epsilon^{(n)} \right) \rightarrow 1$  as  $n \rightarrow \infty$ ?

(c) Show that for all  $n$ ,

$$|A_\epsilon^{(n)} \cap B_\epsilon^{(n)}| \leq 2^{n(H+\epsilon)}.$$

(d) Show that for all  $n$  sufficiently large:

$$|A_\epsilon^{(n)} \cap B_\epsilon^{(n)}| \geq \left( \frac{1}{2} \right) 2^{n(H-\epsilon)}.$$

## 7. An AEP-like limit and the AEP (Bonus)

- (a) Let  $X_1, X_2, \dots$  be i.i.d. drawn according to probability mass function  $p(x)$ . Find the limit in probability as  $n \rightarrow \infty$  of

$$p(X_1, X_2, \dots, X_n)^{\frac{1}{n}}.$$

- (b) Let  $X_1, X_2, \dots$  be an i.i.d. sequence of discrete random variables with entropy  $H(X)$ . Let

$$C_n(t) = \{x^n \in \mathcal{X}^n : p(x^n) \geq 2^{-nt}\}$$

denote the subset of  $n$ -length sequences with probabilities  $\geq 2^{-nt}$ .

- i. Show that  $|C_n(t)| \leq 2^{nt}$ .
- ii. What is  $\lim_{n \rightarrow \infty} P(X^n \in C_n(t))$  when  $t < H(X)$ ? And when  $t > H(X)$ ?