

# Homework 1 Solution and Grading Policy

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## Homework Policy: (READ BEFORE YOU START TO WORK)

- Copying from other students' solution is not allowed. If caught, all involved students get 0 point on that particular homework. Caught twice, you will be asked to drop the course.
- Collaboration is welcome. You can work together with **at most one partner** on the homework problems which you find difficult. However, you should write down your own solution, not just copying from your partner's.
- Your partner should be the same for the entire homework.
- Put your collaborator's name beside the problems that you collaborate on.
- When citing known results from the assigned references, be as clear as possible.

## 1. (Phase transition in error probability) [20]

For a lossless source coding problem, let  $\epsilon^*(n, k)$  denote the smallest possible error probability that any  $(n, k)$  source code can ever achieve, that is,

$$\epsilon^*(n, k) := \min \left\{ \epsilon \mid \text{there exists an } (n, k, \epsilon) \text{ source code} \right\}.$$

For a discrete memoryless source  $\{S_i \mid i \in \mathbb{N}\}$ ,  $S_i$ 's being i.i.d. copies of a discrete random variable  $S$  with entropy  $H(S)$ , prove the following statements.

a) If  $R > H(S)$ , then

$$\lim_{n \rightarrow \infty} \epsilon^*(n, \lfloor nR \rfloor) = 0 \quad [10]$$

b) If  $R < H(S)$ , then

$$\lim_{n \rightarrow \infty} \epsilon^*(n, \lfloor nR \rfloor) = 1. \quad [10]$$

Note that these are alternative ways to state the achievability part and the converse part of the lossless source coding theorem in the lecture, respectively.

**Solution:****1. a)**

Claim:  $\forall \epsilon \in (0, 1), \exists N_0 \in \mathbb{N}$  such that a  $(n, \lfloor nR \rfloor, \epsilon)$  code exists for all  $n \geq N_0$ .

Consider a  $\delta$ -typical set  $\mathcal{A}_\delta^{(n)}$  whose parameter  $\delta > 0$  will be determined later. If we select this set to be the range of the decoding function, then the error probability is  $\Pr \{S_n \notin \mathcal{A}_\delta^{(n)}\}$ .

To make it a  $(n, \lfloor nR \rfloor)$  code, we select  $\delta \leq \frac{R-H(S)}{2}$  so that

$$\lceil \log |\mathcal{A}_\delta^{(n)}| \rceil \leq n(H(S) + \delta) + 1 \leq nR,$$

when  $n$  is large enough. By Proposition 1.2, we have:

For any  $\epsilon \in (0, 1), \exists N_0 \in \mathbb{N}$  such that  $\Pr \{S_n \notin \mathcal{A}_\delta^{(n)}\} < \epsilon$  for all  $n \geq N_0$ . This perfectly proves the claim.

Therefore, for any  $\epsilon \in (0, 1)$ , exist  $N_0$  such that an  $(n, k, \epsilon)$  code exists for all  $n \geq N_0$ . We can then conclude that  $\lim_{n \rightarrow \infty} \epsilon^*(n, k) = 0$ .

**Grading Policy**

- (10) Flawless proof
- (8) Reasonable procedure but with some missing detail, e.g. undetermined  $\delta$
- (6) Conceptually correct but lack of detail
- (4) Conceptually correct but no explanation
- (2) Wrong but I can tell you have tried

**1. b)**

Claim:  $\forall \epsilon \in (0, 1), \exists N_0 \in \mathbb{N}$  such that  $\forall (n, k)$  code with  $n \geq N_0, P_\epsilon^{(n)} > 1 - \epsilon$ .

For any  $(n, k)$  code, denote its range of decoding function by  $\mathcal{B}_n$ .

Then we have for any  $\delta' \in (0, 1)$ ,

$$\begin{aligned} \Pr \{\hat{S}_n = S_n\} &\leq \Pr \{S_n \in \mathcal{B}_n\} \\ &= \Pr \{S_n \in \mathcal{B}_n \cap \mathcal{A}_{\delta'}^{(n)}\} + \Pr \{S_n \in \mathcal{B}_n \cap \mathcal{A}_{\delta'}^{(n)c}\} \quad (\text{by total probability}) \\ &\leq |\mathcal{B}_n| \cdot 2^{-n(H(S)-\delta')} + \Pr \{S_n \in \mathcal{A}_{\delta'}^{(n)c}\} \quad (\text{by Proposition 1.1}) \\ &\leq 2^{nR} \cdot 2^{-n(H(S)-\delta')} + \Pr \{S_n \in \mathcal{A}_{\delta'}^{(n)c}\} \\ &= 2^{n(H(S)-\delta)} \cdot 2^{-n(H(S)-\delta')} + \Pr \{S_n \in \mathcal{A}_{\delta'}^{(n)c}\} \quad (\text{say } R = H(S) - \delta) \\ &= 2^{n(\delta'-\delta)} + \Pr \{S_n \in \mathcal{A}_{\delta'}^{(n)c}\} \end{aligned}$$

Now, we aim to select  $N_0$  according to any given  $\epsilon \in (0, 1)$ .

First, we choose  $\delta' = \frac{\delta}{2}$ , and we will have  $N_1 = \frac{2}{\delta} \log \frac{2}{\epsilon}$  so that  $2^{(\delta' - \delta)} < \frac{\epsilon}{2}$ ,  $\forall n \geq N_1$ .

Next, by Proposition 1.2, there exists  $N_2 \in \mathbb{N}$  such that  $\Pr \left\{ S_n \in \mathcal{A}_{\frac{\delta}{2}}^{(n)^c} \right\} < \frac{\epsilon}{2}$ ,  $\forall n \geq N_2$ .

Finally, by taking  $N_0 = \max \{N_1, N_2\}$ , we can conclude that  $\Pr \left\{ \hat{S}_n = S_n \right\} \leq \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon$ ,  $\forall n \geq N_0$ .

Then the claim is proved.

Therefore, for any  $\epsilon \in (0, 1)$ , once  $n$  is large enough, any  $(n, k)$  code has error probability greater than  $1 - \epsilon$ . So, we can conclude that  $\lim_{n \rightarrow \infty} \epsilon^*(n, k) = 1$ .

### Grading Policy

(10) Flawless proof

(8-6) Reasonable procedure but with some missing detail, e.g. undetermined  $\delta$

(4) Conceptually wrong, e.g. bound the error probability for only typicality-based code

(2) Wrong but I can tell you have tried

## 2. (Entropy calculation) [16]

- a) Let  $X_1, X_2, \dots, X_n$  be  $n$  discrete random variables with disjoint alphabets  $\mathcal{X}_1, \mathcal{X}_2, \dots, \mathcal{X}_n$  respectively. Let  $J$  be a random index, independent of everything else, taking values in  $\{1, 2, \dots, n\}$  with

$$\Pr\{J = j\} = p_j, \quad j = 1, 2, \dots, n.$$

Find  $H(X_J)$  in terms of  $H(X_1), H(X_2), \dots, H(X_n)$  and  $p_1, p_2, \dots, p_n$ . [8]

- b) A biased coin (**Head** with probability  $p$  and **Tail** with probability  $(1 - p)$ ) is flipped until the first **Head** occurs. Let  $N$  denote the number of flips required.

Find  $H(N)$  if it exists, or show that it does not exist.

### Solution:

#### 2. a)

Note that  $H(J|X_J) = 0$  since the alphabets are disjoint. Hence,  $H(X_J) = H(X_J, J) - H(J|X_J) = H(X_J, J)$ .

Then,

$$H(X_J, J) = H(J) + H(X_J|J) \quad (\text{by chain rule})$$

$$= \sum_{j=1}^n -p_j \log p_j + \sum_{j=1}^n p_j H(X_J|J = j) \quad (\text{by definition})$$

$$= \sum_{j=1}^n -p_j \log p_j + \sum_{j=1}^n p_j H(X_j)$$

### Grading Policy

- (8) Flawless proof and correct answer
- (6) Correct answer but missing detail about the use of chain rule
- (4-0) Wrong answer

### 2. b)

Note that  $\mathbf{p}_N(n) := \Pr\{N = n\} = (1-p)^{n-1}p$ .

Thus, by definition,  $H(N) = \sum_{n=1}^{\infty} -\mathbf{p}_N(n) \log \mathbf{p}_N(n)$  if it exists.

$$\begin{aligned} \sum_{n=1}^{\infty} -\mathbf{p}_N(n) \log \mathbf{p}_N(n) &= \sum_{n=1}^{\infty} -(1-p)^{n-1}p \log((1-p)^{n-1}p) \\ &= -p \log(1-p) \sum_{n'=0}^{\infty} n'(1-p)^{n'} - p \log p \sum_{n'=0}^{\infty} (1-p)^{n'} \\ &= -p \log(1-p) \cdot \frac{1-p}{p^2} - p \log p \cdot \frac{1}{p} \\ &= \frac{-(1-p) \log(1-p) - p \log p}{p}, \text{ exist if } p > 0 \end{aligned}$$

The third equality comes from the following. Let  $r < 1$  and  $S_n := \sum_{i=0}^n ir^i = 1 \cdot r + 2 \cdot r^2 + \dots + n \cdot r^n$ .

Thus,  $rS_n = 1 \cdot r^2 + 2 \cdot r^3 + \dots + n \cdot r^{n+1}$ .

Then we have  $(1-r)S_n = r + r^2 + \dots + r^n - n \cdot r^{n+1} = \frac{r(1-r^n)}{1-r} - n \cdot r^{n+1}$ .

That results in  $S_n = \frac{r(1-r^n)}{(1-r)^2} - \frac{n \cdot r^{n+1}}{1-r} \rightarrow \frac{r}{(1-r)^2}$  as  $n \rightarrow \infty$ .

### Grading Policy

- (8) Flawless proof and correct answer
- (7) Almost correct but ignore the case when  $p = 0$
- (6-0) More mistakes in the derivation

### 3. (Mixing Increases Entropy) [14]

Consider a probability vector  $\mathbf{p} = (p_1, \dots, p_i, \dots, p_j, \dots, p_d)$ .

- a) For  $i \neq j \in \{1, 2, \dots, d\}$ , an  $\{i, j\}$ -mixing of  $\mathbf{p}$ , called  $\mathbf{p}_{\{i,j\}}$ , is another probability vector where both the  $i$ -th and the  $j$ -th coordinates are replaced by  $\frac{p_i + p_j}{2}$ .

Show that

$$H(\mathbf{p}) \leq H(\mathbf{p}_{\{i,j\}}). \quad [8]$$

- b) For  $\mathcal{I} \subseteq \{1, 2, \dots, d\}$ , an  $\mathcal{I}$ -mixing of  $\mathbf{p}$ ,  $\mathbf{p}_{\mathcal{I}}$ , is another probability vector where for all the  $i \in \mathcal{I}$ , the  $i$ -th coordinate  $p_i$  is replaced by  $\frac{\sum_{i \in \mathcal{I}} p_i}{|\mathcal{I}|}$ .

Show that

$$H(\mathbf{p}) \leq H(\mathbf{p}_{\mathcal{I}}). \quad [6]$$

**Solution:**

**3. a) b)**

We claim that: The function  $f : \mathbb{R}_+ \rightarrow \mathbb{R}$  defined by  $f(p) := \begin{cases} 0 & , \text{ if } p = 0 \\ -p \log p & , \text{ if } p > 0 \end{cases}$  is concave in  $[0, 1]$ .

The claim can be easily shown by showing that of second derivative of  $f(\cdot)$  in  $(0, 1]$  is negative and that  $f((1 - \alpha) \cdot 0 + \alpha p) \geq (1 - \alpha)f(0) + \alpha f(p)$ .

In this case, for any  $\mathcal{I} \subseteq \{1, 2, \dots, d\}$ ,

$$\begin{aligned} H(\mathbf{p}_{\mathcal{I}}) &:= \sum_{i \in \{1, \dots, d\} \setminus \mathcal{I}} -p_i \log p_i + \sum_{i \in \mathcal{I}} -\bar{p} \log \bar{p} \quad , \text{ where } \bar{p} := \frac{\sum_{j \in \mathcal{I}} p_j}{|\mathcal{I}|} \\ &= \sum_{i \in \{1, \dots, d\} \setminus \mathcal{I}} -p_i \log p_i + |\mathcal{I}| f\left(\frac{\sum_{j \in \mathcal{I}} p_j}{|\mathcal{I}|}\right) \\ &\geq \sum_{i \in \{1, \dots, d\} \setminus \mathcal{I}} -p_i \log p_i + |\mathcal{I}| \sum_{j \in \mathcal{I}} \frac{1}{|\mathcal{I}|} f(p_j) \\ &= H(\mathbf{p}) \end{aligned}$$

And clearly, b) implies a).

### Grading Policy

- (14) Flawless proof
- (13-12) Almost correct but fail to cover the case when the entry is 0
- (10) Reasonable procedure but with unclear derivation and abuse of notation
- (8-0) More mistakes in derivation