

# B8.4 Information Theory

## Sheet 3 — MT25

### Section A

1. Let  $|\mathcal{X}| = 100$  and  $p$  the uniform distribution on  $\mathcal{X}$ . How many codewords are there of length  $l = 1, 2, \dots$  in a Huffman binary code?

**Solution:** By the Huffman procedure, we can see that there are 28 codewords of length 6 and 72 of length 7.

Another way to get these numbers is as follows:

Consider the optimization of  $l_x$  for optimal code

$$\min \sum_{i=1}^{100} p_i l_i \quad \text{subject to} \quad \sum_{i=1}^{100} 2^{-l_i}$$

The optimal  $l_i$  should be integers close to  $-\log(p_i)$ , i.e. 6 or 7 in this question.

To prove this, write  $\Gamma = \{u = (u_1, \dots, u_{100}) : \sum 2^{-u_i} \leq 1\}$  for the set of feasible solutions (without integer constraint), and  $J(u) = \sum u_i$  for the objective function.

Defining  $u^* = \log(100) * (1, 1, 1, \dots, 1)$ ,  $A = \{6, 7\}^{100} \cap \Gamma$ , and  $\bar{A}$  be the convex hull of  $A$ , which is contained in  $\Gamma$ .

Then for any feasible solution in  $\bar{A}$ , the segment between  $u$  and  $u^*$  must intersect with  $\bar{A}$ , hence intersect with the surface of  $\bar{A}$ . So, there exists a  $\lambda \in (0, 1)$  such that  $u^\lambda = \lambda u + (1 - \lambda)u^*$  is on the surface of  $\bar{A}$ , and  $J(u^\lambda) = \lambda J(u) + (1 - \lambda)J(u^*)$ . Since  $J(u^*) < J(u)$ , so  $J(u^\lambda) < J(u)$ . Furthermore,  $u^*\lambda$  is on the surface of  $\bar{A}$ , so there exists a  $\hat{u} \in A$  such that  $J(\hat{u}) \leq J(u^\lambda)$ , which implies  $u$  cannot be optimal.

2. Consider an alphabet  $\mathcal{X} = \{A, B, C\}$ , with probabilities  $p(A) = 0.3, p(B) = 0.5, p(C) = 0.2$ . Consider building an arithmetic code with these probabilities (in this order). Compute the interval associated with the input string 'ABBA' and the first 5 digits of the string associated with the number  $1/\pi$ .

**Solution:** To encode the string ABBA, we have the sequence of intervals (written as

decimals):

$$A \mapsto [0, 0.3)$$

$$AB \mapsto [0.3 \times (0.3 - 0), 0.8 \times (0.3 - 0)) = [0.09, 0.24)$$

$$ABB \mapsto [0.09 + 0.3 \times (0.24 - 0.09), 0.09 + 0.8 \times (0.24 - 0.09)) = [0.135, 0.21)$$

$$ABBA \mapsto [0.135 + 0 \times (0.21 - 0.135), 0.135 + 0.3 \times (0.21 - 0.135)) = [0.135, 0.1575)$$

As our smallest probability is 0.2, after 5 iterations our interval will have width at least  $0.2^5 = 0.00032$ , so  $10^{-4}$  is enough accuracy for our codeword! We therefore wish to decode  $1/\pi \approx 0.3183$ , as

$$\begin{aligned} 0.3183 &\in [0.3, 0.8) \mapsto B \\ \frac{0.3183 - 0.3}{0.5} &= 0.0366 \in [0, 0.3) \mapsto A \\ \frac{0.0366 - 0}{0.3} &= 0.122 \in [0, 0.3) \mapsto A \\ \frac{0.122 - 0}{0.3} &= 0.40667 \in [0.3, 0.8) \mapsto B \\ \frac{0.40667 - 0.3}{0.5} &= 0.21334 \in [0, 0.3) \mapsto A \end{aligned}$$

Hence  $1/\pi \mapsto BAABA\dots$ . Observe that we do not need to construct the full codebook, but we do require high precision arithmetic here.

3. Consider a DMC with  $\mathcal{X} = \mathcal{Y} = \{0, 1, 2, \dots, 10\}$  and  $M = (\mathbb{P}(Y = y|X = x))_{x \in \mathcal{X}, y \in \mathcal{Y}}$ . It is known that  $Y = (X + Z) \bmod 11$ , where  $Z$  is independent of  $X$  and has pmf  $p_Z(i) = \frac{1}{3}$  for  $i \in \{1, 2, 3\}$ . Find the capacity of this channel and the distribution of  $X$  that achieves the capacity.

**Solution:**  $C = \max_{p_X} \{I(X, Y)\}$ . For any  $p_X$  over  $\mathcal{X}$ ,

$$\begin{aligned} I(X, Y) &= H(Y) - H(Y|X) = H(Y) - H(X + Z|X) \\ &= H(Y) - H(Z|X) = H(Y) - H(Z) \\ &= H(Y) - \log(3) \\ &\leq \log(11) - \log(3), \end{aligned}$$

and the equality holds in the last inequality if and only if  $Y$  follows the uniform distribution, which is realised when  $p_X$  is the uniform distribution.

## Section B

4. Consider an alphabet  $\mathcal{X} = A, B$  with pmf  $p(A) = 1 - 2^{-3}$  and  $p(B) = 2^{-3}$ , and binary output alphabet  $\mathcal{Y} = 0, 1$ .
  - (a) Construct the binary Shannon codes for blocks of length 1, 2 and 3.
  - (b) For each of these codes, determine whether it is an optimal code.
  - (c) Construct an optimal block code for this data with length three blocks. Compute its average blocklength and compare with  $H(X)$ , where  $X \sim p$ .
5. Let  $X$  be uniformly distributed over a finite set  $\mathcal{X}$  with  $|\mathcal{X}| = 2^n$  for some  $n \in \mathbb{N}$ . Given a sequence  $A_1, A_2, \dots$  of subsets of  $\mathcal{X}$  we ask a sequence of questions of the form  $X \in A_1, X \in A_2$ , etc.
  - (a) We can choose the sequence of subsets, but cannot vary them depending on the answers to previous questions. How many questions do we need to determine the value of  $X$ ? What is the most efficient way to do so?  
 [Note: If we regard all questions as a mapping from  $\mathcal{X}$  to  $\{\text{Yes}, \text{No}\}^*$ , we can even think about how to design the sequence of subsets to minimise the expected number of questions to ask to get the value of a random variable  $X$  with any given distribution.]
  - (b) We now randomly (i.i.d. and uniformly) draw a sequence of sets  $A_1, A_2, \dots$  from the set of all subsets of  $\mathcal{X}$ . Fix  $x, y \in \mathcal{X}$ . Conditional on  $\{X = x\}$ :
    - (i) What is the probability that  $x$  and  $y$  are indistinguishable after the first  $k$  random questions?
    - (ii) What is the expected number of elements in  $\mathcal{X} \setminus \{x\}$  that are indistinguishable from  $x$  after the first  $k$  questions?
6. Consider a DMC  $(\mathcal{X}, M, \mathcal{Y})$  with  $|\mathcal{X}| = |\mathcal{Y}| = 3$  and the stochastic matrix

$$M = \begin{pmatrix} 2/3 & 1/3 & 0 \\ 1/3 & 1/3 & 1/3 \\ 0 & 1/3 & 2/3 \end{pmatrix}.$$

- (a) Calculate the capacity of this DMC.
- (b) Give an intuitive argument why the capacity is achieved with a distribution that places zero probability on an input symbol.

7. Let  $\mathcal{X}$  be a set of symbols with  $|\mathcal{X}| = m$ . For a variable-to-fixed length code we take a set of strings  $\mathcal{W}_d$  in  $\mathcal{X}^*$  and encode them by a set of binary symbols in blocks of fixed length  $d$ . We can define the efficiency of the code to be

$$R(\mathcal{W}_d) = \frac{d}{\mathbb{E}(L(\mathcal{W}_d))},$$

where  $\mathbb{E}(L(\mathcal{W}_d))$  is the expected length of a string from  $\mathcal{W}_d$ .

Let  $\mathbf{X}_1^n = (X_i)_{i=1}^n$  be an iid sequence of symbols from  $\mathcal{X}$  with pmf  $p$ . We set  $Y$  to be the random variable taking values in  $\mathcal{W}_d$  according to the induced probabilities from  $p$ . We also let  $L = L(\mathcal{W}_d)$  be the random variable denoting the length of the first set of symbols in  $\mathbf{X}_1^n$  that match an element of  $\mathcal{W}_d$ . We assume that  $n$  is greater than the maximal length of strings in  $\mathcal{W}_d$ .

- (a) Show that  $H(Y) \leq d$ .  
 (b) Note that  $Y = (X_1, \dots, X_L)$ , and that  $H(\mathbf{X}_1^n) = H(Y, L, \mathbf{X}_{L+1}^n)$ . Show that

$$H(\mathbf{X}_{L+1}^n | L) = (n - \mathbb{E}(L))H(X).$$

and hence that  $H(Y) = \mathbb{E}(L)H(X)$ .

- (c) Hence show that  $R(\mathcal{W}_d) \geq H(X)$

Note that it is possible to show that  $R(\mathcal{W}_d^*) < H(X) + 1$  for large enough  $d$  using ideas from renewal theory.

8. Write an implementation of Tunstall's code. Your code should take as input a vector of probabilities over  $m$  input messages, and a length  $d$  of the binary encoding, and return a dictionary of source messages and their binary encodings. You should also provide code to encode and decode a message using this dictionary (including right-padding the input with underscores if needed).

For submission, apply your code using the 27 letter alphabet and probabilities as on Sheet 1 Q6, to construct an optimal variable-to-fixed code with  $d = 8$  bit output. State the length of the encodings, and whether right-padding is needed, for the input strings:

- 'HELLO'
- 'THE\_RAIN\_IN\_SPAIN\_FALLS\_MAINLY\_ON\_THE\_PLAIN'
- 'YESTERDAY'

## Section C

9. For a set  $\mathcal{X} = \{1, \dots, m\}$  with corresponding pmf  $p$ , give a necessary and sufficient condition on  $p$  such that there exists a  $d$ -ary code with average per-character length  $H_d(X)$ .

**Solution:** We know that  $H(X)$  is an upper bound on the average per-character length, and from the source coding theorem we have equality iff there exists a code with  $c(x) = -\log_d(p(x))$  for all  $x \in \mathcal{X}$ . Therefore, clearly, it is necessary that all probabilities are (negative) powers of  $d$ .

Conversely, suppose all probabilities are negative powers of  $d$ , that is  $p(x) = d^{-l_x}$ . Let  $l_{\max} = \max_x l_x$ . We know that  $1 = \sum_x d^{-l_x}$ , in particular

$$1 = \sum_x d^{-l_x} = \sum_{x: l_x < l_{\max}} d^{-l_x} + d^{-l_{\max}} \#\{x : l_x = l_{\max}\}$$

and hence,

$$d^{l_{\max}-1} - \sum_{x: l_x < l_{\max}} d^{l_{\max}-l_x-1} = d^{-1} \#\{x : l_x = l_{\max}\}$$

The left hand side of this equation is an integer, so we know that  $\#\{x : l_x = l_{\max}\}$  is divisible by  $d$ . We can therefore group the least likely symbols together, as in the Huffman construction, without using any symbols with higher probability. Repeating this argument, we see that the Huffman construction will always yield a codeword of length precisely  $l_x$  for each codeword. Finally, we observe that  $\sum_x p_x l_x = -\sum_x p_x \log_d(p_x) = H_d(X)$ , as desired.

10. **(Information theory and gambling)** Suppose  $m$  horses run a race, and the  $i$ th horse wins with probability  $p_i$ . An investment of one pound returns  $o(i)$  pounds if horse  $i$  wins, otherwise the investment is lost. A gambler distributes all of his wealth across the horses:  $b(i) \geq 0$  denotes the fraction of the gambler's wealth that he bets on horse  $i$  and  $\sum_{i=1}^m b(i) = 1$ . We now consider repeating this game over and over.

If  $S_n$  denotes the gambler's wealth after the  $n^{\text{th}}$  race, then

$$S_n = \prod_{i=1}^n b(X_i) o(X_i),$$

where  $X_i$  is the horse that wins the  $i^{\text{th}}$  race and  $S_0 = 1$  is the start capital.

- (a) If  $X_i$  are i.i.d., show that for given  $\mathbf{b} = (b(1), \dots, b(m))$ ,  $\mathbf{p} = (p_1, \dots, p_m)$ , the wealth evolves exponentially, i.e.  $\lim_{n \rightarrow +\infty} \frac{1}{n} \log \left( \frac{S_n}{2^{nW(\mathbf{b}, \mathbf{p})}} \right) = 0$  almost surely, where  $W(\mathbf{b}, \mathbf{p})$  is to be determined. *Hint: Strong law of large numbers.*
- (b) Define  $W^*(\mathbf{p}) := \max_{\mathbf{b}: \sum b(i)=1, b(i) \geq 0} W(\mathbf{b}, \mathbf{p})$  and find  $\mathbf{b}$  that achieves this maximum. *Hint: You can find a candidate by using Lagrange multipliers.*
- (c) (Informal.) We can regard  $q_i := \frac{1}{o(i)}$  as the “probabilities” the bookmaker implicitly assigns to  $o(i)$  outcomes. Considering the cases  $\sum q_i = 1$ ,  $\sum q_i < 1$  and  $\sum q_i > 1$ , discuss the fairness of the game.

**Solution:**

- (a) Since

$$\frac{1}{n} (\log(S_n / 2^{nW(\mathbf{b}, \mathbf{p})})) = \frac{1}{n} \sum_{i=1}^n \log(\mathbf{b}(X_i) o(X_i)) - W(\mathbf{b}, \mathbf{p}),$$

and by the law of large number

$$\lim_{n \rightarrow +\infty} \frac{1}{n} \sum_{i=1}^n \log(\mathbf{b}(X_i) o(X_i)) = \mathbb{E}[\log(\mathbf{b}(X_1) o(X_1))] = \sum_{i=1}^m \log(\mathbf{b}(i) o(i)) p_i.$$

Hence  $W(\mathbf{b}, \mathbf{p}) = \sum_{i=1}^m \log(\mathbf{b}(i) o(i)) p_i$ .

- (b) By the last part, we have

$$\begin{aligned} W(\mathbf{b}, \mathbf{p}) &= \sum_{i=1}^m \log(\mathbf{b}(i) o(i)) p_i \\ &= \sum_{i=1}^m \log(p(i)) p_i + \sum_{i=1}^m \log\left(\frac{\mathbf{b}(i)}{p_i}\right) p_i + \sum_{i=1}^m \log(o(i)) p_i \\ &= -H(\mathbf{p}) - D(\mathbf{p} \parallel \mathbf{b}) + \sum_{i=1}^m \log(o(i)) p_i \\ &\leq -H(\mathbf{p}) + \sum_{i=1}^m \log(o(i)) p_i, \end{aligned}$$

and the equality in the last inequality holds iff  $\mathbf{b} = \mathbf{p}$ .

(c) We know  $W^*(\mathbf{p}) = \sum_{i=1}^m \log(p_i o(i)) p_i$ . In terms of  $q_i = \frac{1}{o(i)}$ , we can write it into

$$W^*(\mathbf{p}) = \sum_{i=1}^m \log(p_i / q(i)) p_i.$$

Denote  $K = \sum_{j=1}^m q_j$ , then we can define  $\hat{q}_i = \frac{q_i}{K}$ , with which  $\hat{\mathbf{q}} = (\hat{q}_1, \dots, \hat{q}_m)$  is a pmf, and

$$\begin{aligned} W^*(\mathbf{p}) &= \sum_{i=1}^m \log(p_i / \hat{q}(i)) p_i - \log(K) \\ &= D(\mathbf{p} || \hat{\mathbf{q}}) - \log(K). \end{aligned}$$

In conclusion,

- If  $K < 1$ , then  $W^*(\mathbf{p}) > 0$ , which is favourable for the gambler;
- If  $K = 1$ , this game is still favourable unless  $\mathbf{q}$  is parallel to  $\mathbf{p}$ .
- If  $K > 1$ , then this game can be favourable for the bookmaker.