

«Calcul formel avancé et application». Very brief lecture notes.

## 21.09.2023. Lecture 2.

### 1. Discussion of the homework.

The exercise about an optimal sorting for  $k = 4$  and  $k = 5$  objects remains unsolved.

**2. The game guess a number revisited.** We discussed the version of the game “guess a number,” where the first player chooses at random an integer number between 1 and  $k$  with (some fixed and known in advance) probabilities  $p_1, \dots, p_k$ , and the second player should reveal this number by asking questions with answers *yes* or *no*, with the minimal *on average* number of questions.

**Proposition 1.** *For every random variable  $\alpha$  distributed on a set of  $n$  values*

$$0 \leq H(\alpha) \leq \log n.$$

*Moreover,  $H(\alpha) = 0$  if and only if the distribution is concentrated at one point (one probability  $p_i$  is equal to 1, and the other  $p_j$  for  $j \neq i$  are equal to 0), and  $H(\alpha) = \log n$  if and only if the distribution is uniform ( $p_1 = \dots = p_n = \frac{1}{n}$ ).*

*Sketch of proof:* We use the concavity of the function  $\log x$  and Jensen’s inequality for the concave functions.

**Proposition 2.** *For every random variable  $\alpha$  and for every (deterministic) function  $F$ , Shannon’s entropy of the random variable  $\beta = F(\alpha)$  is not greater than Shannon’s entropy of  $\alpha$ .*

*Sketch of proof:* First of all, we observed that  $H(\alpha) = H(\beta)$ , if  $F$  is a bijection. Then, we proved that the entropy of a distribution decreases, when we merge together two points in this distribution; in other words,  $H(\alpha) \geq H(F(\alpha))$ , if  $F$  merges together two points from the range of  $\alpha$  and leaves distinct the other values of  $\alpha$ . By iterating the basic “merging” operations, we prove the inequality  $H(\alpha) \geq H(F(\alpha))$  for an arbitrary function  $F$ .

Given a pair of jointly distributed random variables  $(\alpha, \beta)$  we can apply the definition of Shannon’s entropy three times, with three potentially different distributions: we have Shannon’s entropy of the entire distribution (denoted  $H(\alpha, \beta)$ ) and the entropies of two marginals,  $H(\alpha)$  and  $H(\beta)$ .

We have proved earlier that

**Proposition 1.** *In the game “guess a number,” where the first player chooses at random an integer number between 1 and  $k$  with (known in advance) probabilities  $p_1, \dots, p_k$ , the average number of questions cannot be less than*

$$\sum_{i=1}^k p_i \log \frac{1}{p_i}$$

Now we proved an upper bound for the same game:

**Proposition 2.** *For the game “guess a number,” where the first player chooses at random an integer number between 1 and  $k$  with (known in advance) probabilities  $p_1, \dots, p_k$ , there exists a strategy that requires on average less than*

$$\sum_{i=1}^k p_i \log \frac{1}{p_i} + 1$$

questions.

*Sketch of the proof:* W.l.o.g. we assume that  $p_1 \geq p_2 \geq \dots \geq p_n$ . We define  $\ell_i = \lceil \log_2 \frac{1}{p_i} \rceil$ . Observe that  $\sum 2^{-\ell_i} \leq 1$ . Then, we construct a binary tree with  $n$  leaves and branches of length  $\ell_1, \dots, \ell_n$ .

On the first stage we choose the leftmost branch of length  $\ell_1$ , then we choose the leftmost branch of  $\ell_2$  that is incompatible with the first branch, and so on. On the  $k$ -th step we choose the leftmost a branch of length  $\ell_i$  that is not a continuation of any branch chosen on the stages  $1, \dots, (k - 1)$ . We show that this procedure can be repeated until stage  $n$  due to two key facts:

- the sum  $\sum 2^{-\ell_i}$  is not greater than 1,
- $\ell_1 \leq \ell_2 \leq \dots \leq \ell_n$ .

*End of proof.*

We observed that strategies in the guessing number game are equivalent to prefix-free binary codes. Thus, we have shown that for every probability distribution  $(p_1, \dots, p_n)$  the minimal average length of a binary code  $\sum p_i |c_i|$  is a number between  $\sum_{i=1}^k p_i \log \frac{1}{p_i}$  and  $\sum_{i=1}^k p_i \log \frac{1}{p_i} + 1$ .

**3. Huffman's encoding.** We discussed the construction of Huffman's code and proved its optimality. For a detailed explanation see the textbook *Elements of information theory* by T. M. Cover and J. A. Thomas.

**Exercise 2.1.** Construct Huffman's code for the distribution of probabilities (0.33, 0.34, 0.2, 0.1, 0.05) and find the average length of the codewords for this code.

**4. Block coding.** We discussed the problem of optimal compression for texts of length  $N$  over an alphabet  $\{a_1, \dots, a_k\}$  with known frequencies of letters  $(p_1, \dots, p_k)$ . Using a counting (based on Stirling's formula) we showed that we need

$$\left( \sum_{i=1}^k p_i \log \frac{1}{p_i} \right) \cdot N + o(N)$$

binary digits.