

C&O 739 Information Theory and Applications
University of Waterloo, Winter 2024
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Assignment 1, Jan. 26, 2024
Due: Fri., Feb. 9, 2024

Question 1. The two parts of this question are unrelated.

- (a) Give an example of a random variable X , a prefix-free code C for X , and a Shannon code C' for X such that
- $\mathbb{E}|C(X)| < \mathbb{E}|C'(X)|$, and
 - a codeword $C(x)$ for some x in the support of X is *longer* than the codeword $C'(x)$.
- (b) Let X be a random variable over $[m]$ with distribution p , with $p_1 \geq p_2 \geq \dots \geq p_m > 0$. Let the probability that $X < i$ be denoted by q_i , i.e., $q_i = \sum_{j=1}^{i-1} p_j$. Define a code C as follows: C_i is the first $\lceil \log(1/p_i) \rceil$ bits of the binary expansion of q_i . We may verify that $\mathbb{E}|C(X)|$ is within 1 bit of the entropy $H(X)$. Construct the code for the distribution $(0.5, 0.25, 0.125, 0.125)$. Then prove for any random variable X as above, that the code C is prefix-free.

Question 2. Let p, q be distributions over the same sample space \mathcal{X} such that $\text{supp}(p) \subseteq \text{supp}(q)$. Define $M(p||q) := \sum_{x \in \mathcal{X}} p_x \log \frac{1}{q_x}$. We abbreviate $M(p||q)$ by m below.

Let X_1, X_2, \dots, X_n be i.i.d. $\sim p$, and let $\mathbf{X} := X_1 X_2 \dots X_n$. Fix $\epsilon > 0$.

- (a) [5 marks] For a sequence $\mathbf{x} := x_1 x_2 \dots x_n \in \mathcal{X}^n$, define $q_{\mathbf{x}} := q_{x_1} q_{x_2} \dots q_{x_n}$. Let $S_{n,\epsilon} \subseteq \mathcal{X}^n$ be defined as the following set of sequences

$$S_{n,\epsilon} := \left\{ \mathbf{x} \in \mathcal{X}^n : 2^{-n(m+\epsilon)} \leq q_{\mathbf{x}} \leq 2^{-n(m-\epsilon)} \right\} .$$

Prove that $\Pr(\mathbf{X} \in S_{n,\epsilon}) \rightarrow 1$ as $n \rightarrow \infty$.

- (b) [5 marks] Suppose q is our guess for the distribution p (which is not known to us). Explain how we may compress the sequence \mathbf{X} (i) losslessly to at most $m + \epsilon$ bits per sample *in expectation*; and (ii) to at most $m + \epsilon$ bits per sample in the worst case such that a receiver can recover \mathbf{X} with probability arbitrarily close to 1.

Question 3. The two parts of this question are unrelated.

- (a) Suppose we define an equivalence relation on random variables X, Y on the same sample space \mathcal{X} , so that $X \equiv Y$ iff there is a bijection f on \mathcal{X} such that $Y = f(X)$. Let $\rho(X, Y) = H(X|Y) + H(Y|X)$. Prove that ρ is a metric on the set of equivalence classes of random variables.
- (b) Let X, Y be real-valued random variables, with finite support. Let $Z = X + Y$. State and prove a necessary and sufficient condition for when the entropy of the sum equals the sum of the entropies, i.e., $H(Z) = H(X) + H(Y)$.

Question 4. Let $G := (A, B, E)$ be an n -regular bi-partite graph with $|A| = |B| = m$. Following the steps below, give an information-theoretic proof of the property that the number of independent sets in G is at most $(2^{n+1} - 1)^{m/n}$.

Let \mathbf{X} denote a uniformly random independent set in G , represented by its characteristic vector. For $v \in A \cup B$, let $N(v)$ denote the set of neighbours of v in G , and let $Y_v := \mathbb{1}(\mathbf{X}_{N(v)} = \mathbf{0})$ the Bernoulli random variable indicating whether $\mathbf{X}_{N(v)} = \mathbf{0}$ or not.

Prove that

- (a) $H(\mathbf{X}_B) \leq \frac{1}{n} \sum_{v \in A} H(\mathbf{X}_{N(v)})$;
- (b) $H(\mathbf{X}_A | \mathbf{X}_B) \leq \sum_{v \in A} H(X_v | \mathbf{X}_{N(v)})$; and
- (c) $H(\mathbf{X}_{N(v)}) \leq H(p_v) + (1 - p_v) \log(2^n - 1)$, where $p_v := \Pr(Y_v = 1)$.

Conclude the bound on the number of independent sets stated above.