ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

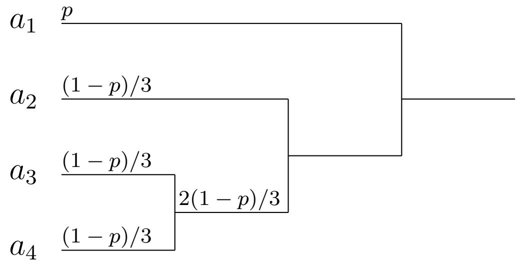
School of Computer and Communication Sciences

Handout 7 | Information Theory and Coding

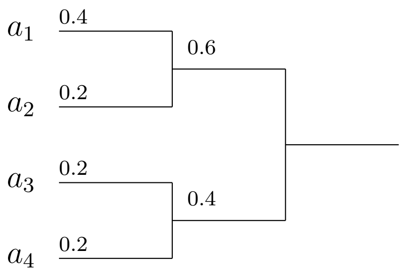
Solutions to Homework 3 | Oct. 10, 2022

Problem 1.

(a) Let $p = P(a_1)$, thus $P(a_2) = P(a_3) = P(a_4) = (1-p)/3$. By the Huffman construction (see figure below) we must have $p > 2(1-p)/3$, i.e., $q = 2/5$ in order to have $n_1 = 1$.



(b) With $P(a_1) = q$, the figure below illustrates that a Huffman code exists with $n_1 > 1$.



(c) & (d) For $K = 2$, n_1 is always 1. For $K = 3$, $n_1 = 1$ is guaranteed by $P(a_1) > P(a_2) \geq P(a_3)$. Now take $K \geq 4$ and assume $P(a_1) > 2/5$ and $P(a_1) > P(a_2) \geq \cdots \geq P(a_K)$. The Huffman procedure will combine a_{K-1} and a_K to obtain a super-symbol with probability

$$P(a_{K-1}) + P(a_K) < 2\frac{3/5}{K-1} \leq 2/5.$$

Thus, in the reduced ensemble a_1 is still the most likely element. Repeating the argument until $K = 3$, we see that $P(a_1) > q$ guarantees $n_1 = 1$ in all cases.

(e) For $K < 3$ no such q' exists. For $K \geq 3$, we claim $q' = 1/3$. Assume a_1 remains unpaired until the 2nd to last stage (otherwise there is nothing to prove). At this stage we have three nodes, and $P(a_1) < q'$ must be strictly less than one of the other two (otherwise all three would have been less than 1/3). Thus a_1 will be combined with one of them, leading to $n_1 > 1$.

PROBLEM 2.

(a) We already know that

$$H(X) + H(Y) \geq H(XY),$$
$$H(Y) + H(Z) \geq H(YZ),$$

and

$$H(Z) + H(X) \geq H(ZX).$$

Adding these inequalities together and diving by two gives

$$H(X) + H(Y) + H(Z) \geq \frac{1}{2}\big[H(XY) + H(YZ) + H(ZX)\big].$$

(b) The difference between the left and right sides, i.e.,

$$H(XY) + H(YZ) - H(XYZ) - H(Y),$$

equals

$$H(X|Y) - H(X|YZ) = I(X;Z|Y),$$

which is always positive.

(c) Using (b) with (YZX) and (ZXY) in the role of (XYZ) gives the inequalities

$$H(YZ) + H(ZX) \geq H(XYZ) + H(Z)$$

and

$$H(ZX) + H(XY) \geq H(XYZ) + H(X).$$

Adding the inequality in (b) to these two gives

$$2\big[H(XY) + H(YZ) + H(ZX)\big] \geq 3H(XYZ) + H(X) + H(Y) + H(Z).$$

(d) Since $H(X) + H(Y) + H(Z) \geq H(XYZ)$, (c) yields

$$2\big[H(XY) + H(YZ) + H(ZX)\big] \geq 4H(XYZ).$$

(e) Let $\big\{(x_i, y_i, z_i) : i = 1, \ldots, n\big\}$ be the xyz-coordinates of the n points. Let X, Y and Z be random variables with $\Pr\big((X,Y,Z) = (x_i, y_i, z_i)\big) = 1/n$ for every $1 \leq i \leq n$. Then, $H(XYZ) = \log_2 n$. Furthermore, the random pair (XY) takes values in the projection of the n points to the xy plane and similarly for (YZ) and (ZX). Thus $H(XY) \leq \log_2 n_{xy}$, $H(YZ) \leq \log_2 n_{yz}$, and $H(ZX) \leq \log_2 n_{zx}$. Part (d) now yields

$$\log_2[n_{xy}n_{yz}n_{zx}] \geq H(XY) + H(YZ) + H(ZX) \geq 2H(XYZ) = 2\log_2 n,$$

which implies that $n_{xy}n_{yz}n_{zx} \geq n^2$.

The relationship between $H(XYZ)$ and $H(XY)$, $H(YZ)$ and $H(ZX)$ is a special case of Han's inequality, which, for a collection of n random variables relates the sum of the $\binom{n}{k}$ joint entropies of k out of n random variables to the sum of the $\binom{n}{k+1}$ entropies of $k+1$ out of n random variables.

The combinatorial fact about the projections of points in 3D is known as Shearer's lemma.

Problem 3.

$$\begin{aligned}
H(X) &= -\sum_{k=1}^{M} P_X(a_k)\log P_X(a_k) \\
&= -\sum_{k=1}^{M-1}(1-\alpha)P_Y(a_k)\log[(1-\alpha)P_Y(a_k)] - \alpha\log\alpha \\
&= (1-\alpha)H(Y) - (1-\alpha)\log(1-\alpha) - \alpha\log\alpha
\end{aligned}$$

Since Y is a random variable that takes $M-1$ values $H(Y) \leq \log(M-1)$ with equality if and only if Y takes each of its possible values with equal probability.

Problem 4.

(a) Using the chain rule for mutual information,

$$I(X,Y;Z) = I(X;Z) + I(Y;Z \mid X) \geq I(X;Z),$$

with equality iff $I(Y;Z \mid X) = 0$, that is, when Y and Z are conditionally independent given X.

(b) Using the chain rule for conditional entropy,

$$H(X,Y \mid Z) = H(X \mid Z) + H(Y \mid X,Z) \geq H(X \mid Z),$$

with equality iff $H(Y \mid X,Z) = 0$, that is, when Y is a function of X and/or Z.

(c) Using first the chain rule for entropy and then the definition of conditional mutual information,

$$\begin{aligned}
H(X,Y,Z) - H(X,Y) &= H(Z \mid X,Y) = H(Z \mid X) - I(Y;Z \mid X) \\
&\leq H(Z \mid X) = H(X,Z) - H(X),
\end{aligned}$$

with equality iff $I(Y;Z \mid X) = 0$, that is, when Y and Z are conditionally independent given X.

(d) Using the chain rule for mutual information,

$$I(X;Z \mid Y) + I(Z;Y) = I(X,Y;Z) = I(Z;Y \mid X) + I(X;Z),$$

and therefore

$$I(X;Z \mid Y) = I(Z;Y \mid X) - I(Z;Y) + I(X;Z).$$

We see that this inequality is actually an equality in all cases.

Problem 5. Let X^i denote $X_1, \dots, X_i$.

(a) By stationarity we have for all $1 \leq i \leq n$,

$$H(X_n|X^{n-1}) \leq H(X_n|X_{n-i+1}, X_{n-i+2}, \dots, X_{n-1}) = H(X_i|X^{i-1}),$$

which implies that,

$$\begin{aligned}
H(X_n|X^{n-1}) &= \frac{\sum_{i=1}^{n} H(X_n|X^{n-1})}{n} && (1) \\
&\leq \frac{\sum_{i=1}^{n} H(X_i|X^{i-1})}{n} && (2) \\
&= \frac{H(X_1, X_2, \dots, X_n)}{n}. && (3)
\end{aligned}$$

(b) By the chain rule for entropy,

$$\frac{H(X_1, X_2, \ldots, X_n)}{n} = \frac{\sum_{i=1}^{n} H(X_i|X^{i-1})}{n} \tag{4}$$
$$= \frac{H(X_n|X^{n-1}) + \sum_{i=1}^{n-1} H(X_i|X^{i-1})}{n} \tag{5}$$
$$= \frac{H(X_n|X^{n-1}) + H(X_1, X_2, \ldots, X_{n-1})}{n}. \tag{6}$$

From stationarity it follows that for all $1 \le i \le n$,

$$H(X_n|X^{n-1}) \le H(X_i|X^{i-1}),$$

which further implies, by summing both sides over $i = 1, \ldots, n-1$ and dividing by $n-1$, that,

$$H(X_n|X^{n-1}) \le \frac{\sum_{i=1}^{n-1} H(X_i|X^{i-1})}{n-1} \tag{7}$$
$$= \frac{H(X_1, X_2, \ldots, X_{n-1})}{n-1}. \tag{8}$$

Combining (6) and (8) yields,

$$\frac{H(X_1, X_2, \ldots, X_n)}{n} \le \frac{1}{n}\left[\frac{H(X_1, X_2, \ldots, X_{n-1})}{n-1} + H(X_1, X_2, \ldots, X_{n-1})\right] \tag{9}$$
$$= \frac{H(X_1, X_2, \ldots, X_{n-1})}{n-1}. \tag{10}$$

Problem 6. By the chain rule for entropy,

$$H(X_0|X_{-1}, \ldots, X_{-n}) = H(X_0, X_{-1}, \ldots, X_{-n}) - H(X_{-1}, \ldots, X_{-n}) \tag{11}$$
$$= H(X_0, X_1, \ldots, X_n) - H(X_1, \ldots, X_n) \tag{12}$$
$$= H(X_0|X_1, \ldots, X_n), \tag{13}$$

where (12) follows from stationarity.

Problem 7. $X \multimap Y \multimap (Z, W)$ implies that $I(X; Z, W|Y) = 0$. Then,

$$I(X;Y) + I(Z;W) = I(X;Y) + I(X;Z,W|Y) + I(Z;W) = I(X;Y,Z,W) + I(Z;W)$$

Notice that $I(X;Y) + I(X;Z,W|Y) = I(X;Y,Z,W)$ follows from chain rule. Using the chain rule for a couple of times, we obtain the following steps.

$$I(X;Y,Z,W) + I(Z;W) = I(X;Z) + I(X;Y,W|Z) + I(Z;W) \tag{14}$$
$$= I(X;Z) + I(X;Y|W,Z) + I(X;W|Z) + I(Z;W) \tag{15}$$
$$= I(X;Z) + I(X;Y|W,Z) + I(X,Z;W) \tag{16}$$
$$\ge I(X;Z) + I(X;W) \tag{17}$$

as $I(X,Z;W) \ge I(X;W)$