

1. (14 points) *Entropy and source coding.*

A discrete memoryless source X of alphabet size K has an entropy of $H(X)$ bits. A binary prefix code for this source is found, and has an average length of $E[l(x)] = \frac{H(X)}{\log_2(3)}$ bits per source symbol. Prove that each letter has a probability of the form 3^{-i} for i an integer that can depend on the symbol.

$$E[l(x)] = \frac{H_2(x)}{\log_2 3} \text{ bits / source symbol} \rightarrow \text{in base 2}$$

The entropy in base 3 would be $H_3(x) = \frac{H_2(x)}{\log_2(3)}$, so $H_2(x) = \log_2 3 H_3(x)$

Then, the expected codeword length ~~is~~ if using a ~~code~~ code in base D would be

$$E[l(x)] = \frac{H_2(x)}{\log_2 3} \text{ bits} = \frac{\log_2 3 \times H_3(x)}{\log_2 3} = H_3(x) \text{ ternary bits / source symbol}$$

Since this code would hit the entropy lower bound exactly, the PMF must be 3-adic, i.e. each symbol has a probability of the form 3^{-i} for some i .

2. (14 points) Consider the binary multiplier channel $Y = XZ$, where X and Z are independent binary random variables that take on values 0 and 1, and Z is Bernoulli α , i.e. $P(Z = 1) = \alpha$. Calculate $I(X; Y)$ when $p(x = 1) = q$ and hence $p(x = 0) = 1 - q$ for some $0 < q < 1$.

$$\left. \begin{array}{l} p(x=0) = 1-q, \quad p(x=1) = q \\ p(z=0) = 1-\alpha, \quad p(z=1) = \alpha \end{array} \right\} \begin{array}{l} Y=0 \text{ if } (x \cdot z) = 0, \text{ i.e. if} \\ \quad (x=0, z=0) \text{ or } (x=1, z=0) \\ \quad \text{or } (x=0, z=1) \\ Y=1 \text{ if } x \cdot z = 1, \text{ i.e. if} \\ \quad (x=1, z=1) \end{array}$$

$$I(X; Y) = H(Y) - H(Y|X)$$

$$= H(Y) - \underbrace{H(Y|X=0)}_{\text{if } x=0 \text{ then } y=0 \text{ automatically, so } H(Y|X=0)=0} p(x=0) - \underbrace{H(Y|X=1)}_{\text{if } x=1, \text{ then } y=Z} p(x=1)$$

if $x=0$ then
 $y=0$ automatically,
so $H(Y|X=0)=0$

if $x=1$, then $y=Z$.
 $H(Y|X=1) = H(Z|X=1)$
 $= H(Z)$
 $= -\alpha \log \alpha - (1-\alpha) \log (1-\alpha)$

To calculate $H(Y)$, notice that

$$\begin{aligned} p(y=0) &= (1-q)(1-\alpha) + (1-q)\alpha + (1-\alpha)q \\ p(y=1) &= \alpha \cdot q \end{aligned}$$

call this g^*

so,

$$I(X; Y) = \cancel{H(Y)} - \cancel{H(Y|X)}$$

$$= -\alpha \cdot q \log(\alpha \cdot q) - g^* \log g^* + q \cdot (-\alpha \log \alpha - (1-\alpha) \log (1-\alpha))$$

3. (10 points) For X, Y, Z random variables (not necessarily independent). Determine whether $H(X, Y, Z) - H(X, Y) \leq H(X, Z) - H(X)$ is true or not. If true, find the conditions for equality.

$$\begin{aligned} H(X, Y, Z) - H(X, Y) &= H(X, Z) + H(Y|X, Z) - H(X) - H(Y|X) \\ &= \underbrace{H(X, Z) - H(X)} + [H(Y|X, Z) - H(Y|X)] \\ &= H(X, Z) - H(X) + I(Y; Z|X) \end{aligned}$$

since Mutual information is non-negative this inequality must be true.

Equality holds when $I(Y; Z|X) = 0$, i.e. when Y and Z are conditionally independent given X , which happens when

$Y \rightarrow X \rightarrow Z$ form a Markov chain.

4. Let X, Y, Z all be binary random variables (not necessarily independent).

- (a) (6 points) Find a joint distribution $p(x, y, z)$ such that $I(X; Y) = 0$ and $I(X; Y|Z) = 1$ bit.
 (b) (6 points) Find a joint distribution $p(x, y, z)$ such that $I(X; Y) = 1$ bit and $I(X; Y|Z) = 0$.

(a) As seen in class, take X, Y to be independent Bernoulli($1/2$) (or coin tosses), and let $Z = X \oplus Y$. Then

$$I(X; Y) = 0 \quad (\text{by independence})$$

and

$$\begin{aligned} I(X; Y|Z) &= H(X|Z) - \underbrace{H(X|Y, Z)}_0 \\ &= \underbrace{H(X|Z=0)}_1 \underbrace{p(Z=0)}_{1/2} + \underbrace{H(X|Z=1)}_1 \underbrace{p(Z=1)}_{1/2} \\ &= 1 \end{aligned}$$

(b) let X be a coin toss, and let $Y = Z = X$.

Then,

$$I(X; Y) = H(X) - \underbrace{H(X|Y)}_0 = H(X) = 1$$

$$I(X; Y|Z) = \underbrace{H(Y|Z)}_0 - \underbrace{H(Y|X, Z)}_0 = 0$$

5. Correct the following incorrect statements.

- (a) (5 points) All typical sequences in $A_\epsilon^{(n)}$ have the same probability.
- (b) (5 points) The typical set $A_\epsilon^{(n)}$ is defined as the smallest set of sequences with $\Pr\{A_\epsilon^{(n)}\} \geq 1 - \epsilon$.
- (c) (5 points) The number of sequences in $A_\epsilon^{(n)}$ may be bounded as $|A_\epsilon^{(n)}| \leq 2^{-n(H(X)-\epsilon)}$.
- (d) (5 points) Let $I(X;Y)|_{p(x,y)}$ denote the mutual information between X, Y when they share a joint distribution $p(x, y)$. Then, for fixed $p_1(x), p_2(x)$ and any $0 \leq \lambda \leq 1$

$$I(X;Y)|_{p(y|x)(\lambda p_1(x)+(1-\lambda)p_2(x))} \leq \lambda I(X;Y)|_{p(y|x)p_1(x)} + (1-\lambda)I(X;Y)|_{p(y|x)p_2(x)}$$

(a) False. $A_\epsilon^{(n)}$ defined as $\{x^n: 2^{-n(H(X)+\epsilon)} \leq p(x^n) \leq 2^{-n(H(X)-\epsilon)}\}$
 so all typical sequences have approximately the same probability.

(b) False. As defined above. B_δ^n is the smallest set with $\Pr\{B_\delta^n\} \geq 1 - \delta$.

(c) False. $|A_\epsilon^{(n)}| \leq 2^{+n(H(X)+\epsilon)}$

(d) False. $I(X;Y)$ is concave in $p(x)$ for fixed $p(y|x)$, so the inequality should be the other way.



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6. (10 points) State Jensen's inequality (formally, but no need to prove it.) State where we have used it (to prove what, the basis of what)?

Jensen's inequality:

If $f(x)$ is a convex function and X is a random variable,

$$E[f(X)] \geq f(E[X])$$

If f is strictly concave, then equality in the above implies that $X = E[X]$ with probability 1.

We used Jensen's inequality to show that $D(p||q) \geq 0$, ^{for any} _{p, q}
which in turn was used to show that $I(X; Y) \geq 0$.

7. Your boss gives you a text document with letters distributed i.i.d. according to $p(x)$ and asks you to compress them as much as possible.
- (a) (10 points) How much can you compress them to, i.e. what is the fundamental limit of compression. Please be precise.
- (b) (10 points) If your boss does not care about hitting the limit exactly, but only wants to within 0.1 bits per source symbol of this limit, how would you go about designing a source code?

(a) If coding a block of n iid. symbols (a "super symbol") then we know that the minimal expected codeword length

$$\frac{1}{n} E[l(x_1, \dots, x_n)]$$

satisfies:

$$H(x) \leq \frac{1}{n} E[l(x_1, \dots, x_n)] \leq H(x) + \frac{1}{n}$$

(b) To hit it within 0.1 bits per source symbol, use supersymbols of length at least 10. ~~For each~~ For ~~each~~ ^{the} supersymbol, use either a Shannon code or a Huffman code to obtain a code that satisfies

$$\frac{1}{n} E[l(x_1, \dots, x_n)] \leq H(x) + \frac{1}{10}$$

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