

**ECE 499/599 Data Compression/Information Theory
Spring 06**

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**Homework 1 Solution
Due 04/18/06 at the beginning of the class**

Problem 1:

- a) 2 bits.
- b) 1.75 bits.
- c) 1.739818 bits.

Problem 2: Do problem 4 in chapter 2 of the textbook.

First show that $f(x) = -x \log x - (a-x) \log(a-x)$ is maximum for $x = a/2$, then use this fact to prove whether $H(Q)$ is greater or smaller than $H(P)$. This can be done by taking first derivative to establish a local extreme point at $a/2$. Then, taking the second derivative showing $f''(a/2) > 0$, hence $f(a/2)$ is maximum. We then proceed as follows.

$$\begin{aligned} H_Q - H_P &= -\sum_{i=1}^m q_i \log_2 q_i + \sum_{i=1}^m p_i \log_2 p_i \\ &= -q_{j-1} \log_2 q_{j-1} - q_j \log_2 q_j + p_{j-1} \log_2 p_{j-1} + p_j \log_2 p_j \end{aligned}$$

Given a function

$$f_a(x) = -x \log x - (a-x) \log(a-x)$$

we can easily show that $f_a(x)$ is maximum for $x = \frac{a}{2}$. Let

$$p_{j-1} + p_j = c$$

then

$$q_{j-1} = q_j = \frac{c}{2}$$

Then

$$\begin{aligned} H_Q - H_P &= -\frac{c}{2} \log_2 \frac{c}{2} - \frac{c}{2} \log_2 \frac{c}{2} + p_j \log_2 p_j + (c - p_j) \log_2 (c - p_j) \\ &= f_c\left(\frac{c}{2}\right) - f_c(p_j) \\ &\geq 0 \end{aligned}$$

Therefore $H_Q \geq H_P$.

Another neat proof:

We know the highest entropy is obtained when all the probabilities are equal to each other. In this case, we only need to consider the amount of information of $\{q_{j-1}, q_j\}$ and $\{p_{j-1}, p_j\}$ since the probabilities of other symbols are identical for p and q . Since $q_{j-1} = q_j$, $\{q_{j-1}, q_j\}$ must be maximum. Since entropy is additive (with statistical weight), we have $H(Q) > H(P)$.

Note: I will give bonus points for anyone who comes up with elegant proof like one above.

Problem 3: For a certain exam, 75% of the participating students in the exam pass, 25% do not pass. Of the students who have passed, 10% own a car. Of the students who have failed, 50% own a car.

- a) How much information does one receive if one is told the result of a student's exam?

There are four possible situations, passing, not passing, owning a car, not owning a car, are denoted by s , s' , c and c' , respectively. When the result of an exam is announced, this delivers an amount of information

$$H(\text{result}) = -p(s)\log p(s) - p(s')\log p(s') = -(3/4)\log(3/4) - (1/4)\log(1/4) = 0.81 \text{ bit}$$

- b) How much information is contained in the announcement of a student who has passed that he does or does not have a car?

If a student, who has passed, announces whether or not he has a car, then there are two possibilities c and c' with the given probabilities. Thus,

$$H(\text{car owner}/\text{passed}) = -p(c|s)\log[p(c|s)] - p(c'|s)\log[p(c'|s)] =$$

$$-(1/10)\log(1/10) - (9/10)\log(9/10) = 0.47 \text{ bits}$$

c) How much uncertainty remains concerning the car ownership of a student if he announces the result of his exam?

There are four possibilities in total. The corresponding probabilities are

$$p(s,c) = (3/4)(1/10) = 3/40$$

$$p(s,c') = (3/4)(9/10) = 27/40$$

$$p(s',c) = (1/4)(1/2) = 1/8$$

$$p(s',c') = (1/4)(1/2) = 1/8$$

The amount of information that is delivered by announcing the result of the exam as well as possible car ownership is then

$$H(\text{car ownership, result}) = -(3/40)\log(3/40) - (27/40)\log(27/40) - 2(1/8)\log(1/8) = 1.41 \text{ bits}$$

The remaining uncertainty about car ownership, if the result of the exam is given, is then

$$H(\text{car ownership/result}) = H(\text{car ownership, result}) - H(\text{result}) = 1.41 - .81 = 0.6 \text{ bits.}$$

One can also obtain this result by calculating the conditional amount of information directly as per the definition for the conditional amount of information.

