

**ECE 745 - Advanced Communication Theory, Spring 2007**  
**Homework #1**

**Due: March 12, 2007**

1. A Huffman code finds the optimal codeword to assign to a given block of  $N$  source symbols.
  - (a) Suppose that the probability of the most probable block is greater than 0.4; show that this implies that the codeword generated by a Huffman code for this block has length 1.
  - (b) Suppose that the probability of the most probable block is less than  $\frac{1}{3}$ ; show that this implies that the codeword generated by a Huffman code for this block has length  $\geq 2$ .
  - (c) Show that  $\{01, 100, 101, 1110, 1111, 0011, 0001\}$  cannot be a Huffman code, regardless of the source distribution and the choice of  $N$ .
2. (a) Describe a pair of random variables  $X$  and  $Y$  such that there exists  $y_1$  and  $y_2$  such that

$$H(X|Y = y_1) \leq H(X) \leq H(X|Y = y_2)$$

- (b) Describe a pair of random variables  $X$  and  $Y$  for which  $I(X; Y) = H(X) < H(Y)$ .
- (c) Let  $Z = X + Y$  for real-valued discrete random variables  $X$  and  $Y$ . Show that  $H(Z) \leq H(X) + H(Y)$ . Give an example where  $H(Z) = H(X) + H(Y)$  and another where  $H(Z) < H(X) + H(Y)$ .
- (d) Show that if  $X$  and  $Z$  are conditionally independent given  $Y$ , then

$$H(X|Z) \geq H(X|Y)$$

3. For a given independent and identically distributed (IID) discrete source, the best lossless source code for  $N = 2$  has average length 4 and the best lossless source code for  $N = 3$  has average length 4.8.
  - (a) What can be deduced about the entropy of the source?
  - (b) What can be deduced about the size of the source alphabet?
4. Define the *variational distance* between probability mass functions  $p(x)$  and  $q(x)$ , each with alphabet  $\mathcal{X}$  as

$$V(p, q) = \sum_x |p(x) - q(x)|$$

- (a) Show that  $V(p, q)$  has all of the properties of a metric; that is, show (i)  $V(p, q) \geq 0$  with equality if and only if  $p = q$ , (ii)  $V(p, q) = V(q, p)$ , and (iii)  $V(p, q) \leq V(p, r) + V(r, q)$  for any three probability mass functions  $p, q, r$ , each with alphabet  $\mathcal{X}$ .
- (b) For  $|\mathcal{X}| < \infty$ , show that for any probability mass function  $p(x)$  and any  $\epsilon > 0$ , there exists a  $\delta > 0$  so small that  $D(p||q) \leq \epsilon$  whenever  $V(p, q) \leq \delta$ .

5. A discrete-valued source outputs an independent and identically distributed (IID) sequence of random variables  $(X_i)$ , each drawn from the alphabet  $\mathcal{X} = \{a_1, a_2, \dots, a_Q\}$ . I have not been able to find the probability mass function  $p_{X_i}(x)$ ; however, I have been able to show that for any  $\epsilon > 0$ , there are sets  $B_1, B_2, B_3, \dots$  that have the following three properties:

- (i)  $B_N$  is a subset of  $\mathcal{X}^N$ .
- (ii)  $P((X_1, X_2, \dots, X_N)^T \in B_N) \rightarrow 1$  as  $N \rightarrow \infty$ .
- (iii)  $|B_N| = 4^{3\epsilon N + 5N}$ .

What can be said about the entropy of the source?

6. The following is a formal restatement of our theorem for almost lossless fixed length source codes:

Let  $(X_i)$  be an IID sequence of discrete random variables with common distribution  $p_X(x)$  and first-order entropy  $H(X)$ .

(1) Then, for any  $R > H(X)$ , any  $\delta > 0$ , there exists a fixed length source code with rate less than  $R$  and probability of decoding error  $P_e < \delta$ .

(2) Conversely, for any  $R < H(X)$  and any  $\epsilon > 0$ , there is a (large) integer  $N_0$  such that for all  $N \geq N_0$ , a fixed length source code that operates on  $N$  symbols at a time has probability of decoding error bounded as:

$$P_e > 1 - \epsilon - 2^{N(R - H(X) + \epsilon)}$$

By applying the Typical Sequences Theorem, show:

- (a) The positive statement (statement (1)).
- (b) The converse (statement (2)).
- (c) Show that for  $R < H(X)$ ,  $\lim_{N \rightarrow \infty} P_e = 1$ .

7. Suppose that we are trying to "guess"  $X$  from an observation of  $Y$  with the decision rule  $\hat{X}(Y)$ . Let  $P_e = P(\hat{X}(Y) \neq X)$ .

- (a) Show that any optimum decision rule can never have  $P_e$  greater than  $\frac{|\mathcal{X}|-1}{|\mathcal{X}|}$ .
- (b) Show that Fano's function  $\mathcal{F}_{|\mathcal{X}|}(p)$  is maximized at  $p = \frac{|\mathcal{X}|-1}{|\mathcal{X}|}$ .
- (c) Suppose  $I(X; Y) = H(X)$ . Show that an optimal decision rule has  $P_e = 0$ .