

ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

Communication Systems Department

Handout 16
Solutions to Midterm

Information Theory and Coding
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Problem 1.

- (a) By the chain rule, the left hand side and the right hand side both equal $H(E, M|Y)$.
- (b) Since E is a function of M and Y , we have $H(E|M, Y) = 0$.
- (c) $H(M|E, Y) = \Pr(E = 0)H(M|Y, E = 0) + \Pr(E = 1)H(M|Y, E = 1)$. But $H(M|Y, E = 0)$ is zero since when $E = 0$, $g(Y) = M$ and thus Y determines M , so we have (i). On the other hand, given Y and $E = 1$, we know that M can take on all values except $g(Y)$. Thus M can take on at most $|\mathcal{M}| - 1$ values and its entropy can be at most $\log(|\mathcal{M}| - 1)$.
- (d) Conditioning does not increase entropy, hence $H(E|Y) \leq H(E)$.
- (e) E takes on the value 1 when and only when $\hat{M} \neq M$. This event has probability P_e , so $\Pr(E = 1) = P_e$, and $\Pr(E = 0) = 1 - P_e$. We then conclude that $H(E) = -P_e \log P_e - (1 - P_e) \log(1 - P_e) = h(P_e)$.
- (f) We have

$$\begin{aligned}
 H(M|Y) + H(E|M, Y) &= H(E|Y) + H(M|E, Y) && \text{from (a)} \\
 H(M|Y) &= H(E|Y) + H(M|E, Y) && \text{from (b)} \\
 H(M|Y) &\leq H(E|Y) + \Pr(E = 1) \log(|\mathcal{M}| - 1) && \text{from (c)} \\
 H(M|Y) &\leq H(E) + \Pr(E = 1) \log(|\mathcal{M}| - 1) && \text{from (d)} \\
 H(M|Y) &\leq h(P_e) + P_e \log(|\mathcal{M}| - 1) && \text{from (e)}.
 \end{aligned}$$

Problem 2.

- (a) We have
 - (i) $H(X|Y) = H(X)$ since X and Y are independent.
 - (ii) $H(X|K) = H(X)$ since X and K are independent.
 - (iii) $H(Y|X, K) = 0$ since X and K determine Y .
 - (iv) $H(X|Y, K) = 0$ since Y and K determine X by the decryptability condition.
 - (v) $I(X; Y|K) = H(X|K) - H(X|Y, K) = H(X)$ by (iv) and (ii).
 - (vi) $H(Y|K) = I(X; Y|K) + H(Y|X, K) = H(X)$ by (v) and (iii).
- (b) Suppose k a key common to both $\mathcal{K}(x_1)$ and $\mathcal{K}(x_2)$. Then, the pair y_0, k can be decrypted as either x_1 or x_2 , contradicting the decryptability condition.

- (c) Since $I(X;Y) = 0$ we know that X and Y are independent and thus, $\Pr(Y = y) = \Pr(Y = y|X = x)$ for all x and y . In particular

$$0 < \Pr(Y = y_0) = \Pr(Y = y_0|X = x).$$

Thus for each x , $\mathcal{K}(x)$ is not empty, for otherwise $\Pr(Y = y_0|X = x)$ would have been zero. If any $\mathcal{K}(x)$ had more than one element, then the total number of keys would exceed the number of source letters; thus each $\mathcal{K}(x)$ must have exactly one element.

- (d) Given that $X = x$, the only way $Y = y_0$ is when $K = k(x)$. Since X and K are independent this happens with probability $\Pr(K = k(x))$.
- (e) We have $\Pr(Y = y_0) = \Pr(Y = y_0|X = x) = \Pr(K = k(x))$. Since the left hand side does not depend on x , the same must be true for the right hand side. Since $k(x)$ exhausts all the keys as x ranges over the source letters, we see that $\Pr(K = k)$ does not depend on k and hence that K is uniformly distributed.

Problem 3.

- (a) If for some i , $q_i < p_{i-1}$, we can exchange the subtrees rooted at q_i and p_{i-1} . This would elongate by 1 the codewords for a set of source letters with probability q_i and shorten by 1 the codewords for a set of source letters with probability p_{i-1} . Since $q_i < p_{i-1}$ this shortens the expected codeword length by $p_{i-1} - q_i$, contradicting the optimality of the Huffman code. [Alternatively, if $q_i < p_{i-1}$, the Huffman procedure would have merged q_{i-1} with q_i , not p_{i-1} .]
- (b) We have $p_0 = F_0 p_0$ and $p_1 = q_0 + p_0 \geq F_1 p_0$. Using these facts as our induction base, suppose that $p_n \geq F_n p_0$ for all $n < i$. Then,

$$\begin{aligned} p_i &= p_{i-1} + q_{i-1} \\ &\geq p_{i-1} + p_{i-2} && \text{part (a)} \\ &\geq F_{i-1} p_0 + F_{i-2} p_0 && \text{induction hypothesis} \\ &= F_i p_0 && \text{Fibonacci recursion} \end{aligned}$$

completing the the proof by induction.

- (c) Since $1 = p_{n_0} \geq F_{n_0} p_0$, the claim follows.
- (d) If $q_i = p_{i-1}$ the Huffman procedure can choose to merge q_{i-1} with either q_i or p_{i-1} without loss of optimality. For three source letters, any distribution of the form $\{\alpha, \alpha, 1 - 2\alpha\}$ for $\alpha \geq 1/3$ yields a valid example. For larger source alphabets, $\{1/8, 1/8, 2/8, 4/8\}$, $\{1/16, 1/16, 2/16, 4/16, 8/16\}$, ... are other possible examples.
- (e) Since $q_0 > 0$, we have $p_1 > F_1 p_0$ which says that the bound in part (b) (and thus in part (c)) cannot be made to hold with equality. However, by letting $q_i = p_{i-1}$ for $i \geq 1$ as in part (d), one will get $p_i = F_i p_0 + F_{i-1} q_0$ for $i \geq 1$: for $i = 1, 2, 3$, the equality holds (induction base), assuming that it holds for all $n < i$,

$$\begin{aligned} p_i &= p_{i-1} + q_{i-1} \\ &= p_{i-1} + p_{i-2} \\ &= F_{i-1} p_0 + F_{i-2} q_0 + F_{i-2} p_0 + F_{i-3} q_0 \\ &= F_i p_0 + F_{i-1} q_0 \end{aligned}$$

proving the claim. Now, choose q_0 small enough to approach equality in $p_i \geq F_i p_0$ all i (upto n_0). Same construction yields

$$p_0 = (1 - F_{n_0-1} q_0) / F_{n_0}$$

which can be made arbitrarily close to $1/F_{n_0}$ by taking small enough q_0 .

The Fibonacci recursion can be solved to yield $F_n = [\phi^{n+1} - \phi^{-n-1}] / \sqrt{5}$ where $\phi = (1 + \sqrt{5})/2$. The last result shows that, for small p_0 one can get

$$n_0 \approx \frac{-\log_2 p_0}{\log_2 \phi} \approx -1.45 \log_2 p_0.$$

In other words, for some source letters, the Huffman procedure can yield a codeword that is much longer than one would expect, $-\log p_0$.