

## Problem Set 1

15-859 Information Theory and Applications in TCS

Name: Shashank Singh

Email: sss1@andrew.cmu.edu

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### Problem 1

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All entropies are given in bits.

$$P(Y = 3) = \frac{1}{4}, P(Y = 4) = P(Y = 5) = \frac{3}{8}, \text{ so } H(Y) = \frac{1}{4} \log_2(4) + 2 \cdot \frac{3}{8} \log_2\left(\frac{8}{3}\right) = \boxed{\frac{11 - 3 \log_2 3}{4}}.$$

$$\text{Since } Y \text{ is a (deterministic) function of } X, \boxed{H(Y|X) = 0} \text{ and } I(X;Y) = H(Y) = \boxed{\frac{11 - \log_2 3}{4}}.$$

$H(X|Y = 3) = \log_2 2 = 1$ ,  $H(X|Y = 4) = \log_2 6 = 1 + \log_2 3$ , and  $H(X|Y = 5) = \log_2 12 = 2 + \log_2 3$  (computed by counting the number of possible series of each length). Thus,

$$H(X|Y) = \mathbb{E}_{y \in \{3,4,5\}} [H(X|Y = y)] = \frac{1}{4} \cdot 1 + \frac{3}{8} (3 + 2 \log_2 3) = \boxed{\frac{11 + 6 \log_2 3}{8}}.$$

$$\text{Thus, } H(X) = H(X|Y) + H(Y) = \boxed{\frac{33}{8}}.$$

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### Problem 2

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- (a) For  $i \in \{1, 2, \dots, n-2\}$ ,  $I(B_i, B_{i+1} | B_1, B_2, \dots, B_{i-1}) = 0$ , since  $B_i$  is independent of  $B_{i+1}$  given  $B_1, B_2, \dots, B_{i-1}$ . However, if  $i = n-1$ , then,  $I(B_i, B_{i+1} | B_1, B_2, \dots, B_{i-1}) = 1$ , since  $H(B_{i+1}) = 1$ , and  $B_{i+1}$  can be uniquely determined from  $B_i$ , given the values of  $B_1, B_2, \dots, B_{i-1}$ .
- (b) Since conditioning cannot reduce entropy,  $H(Y|X, Z) \leq H(Y|X)$ .

$$\begin{aligned} H(X, Y, Z) + H(X) &= H(Y, Z | X) + 2H(X) \\ &= H(Y | X, Z) + H(Z | X) + 2H(X) \\ &\leq H(Y | X) + H(Z | X) + 2H(X) \\ &= H(X, Z) + H(X, Z). \end{aligned}$$

This inequality can be re-written as the “submodular” inequality. Furthermore,  $H(Y|X, Z) = H(Y|X)$  if and only if  $Y$  is conditionally independent of  $Z$  given  $X$ , so that this is precisely the condition under which equality holds. ■

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**Problem 3**


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- (a) Since, for  $X = x$  fixed,  $Z = z$  if and only if  $Y = z - x$ ,

$$\begin{aligned} H(Z|X) &= \sum_x p(x) \sum_z p(Z = z|X = x) \log \left( \frac{1}{p(Z = z|X = x)} \right) \\ &= \sum_x p(x) \sum_z p(Y = z - x|X = x) \log \left( \frac{1}{p(Y = z - x|X = x)} \right) \\ &= \sum_x p(x) \sum_y p(Y = y|X = x) \log \left( \frac{1}{p(Y = y|X = x)} \right) = H(Y|X). \quad \blacksquare \end{aligned}$$

- (b) As shown in part (d) below, if  $X \perp Y$ , then  $H(Z) = H(X) + H(Y) \geq \max\{H(X), H(Y)\}$ , since entropy is non-negative.  $\blacksquare$
- (c) Suppose  $X \sim \text{Bernoulli}(1/2)$  and  $Y = -X$ . Then,  $H(X) = H(Y) = 1$ , but, since  $Z$  is always 0,  $H(Z) = 0 < 1 = \min\{H(X), H(Y)\}$ .  $\blacksquare$
- (d) By part (a),

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

Since  $I(X; Y) = 0$  if and only if  $X \perp Y$ ,  $H(Z) = H(X) + H(Y)$  if and only if  $X \perp Y$ .  $\blacksquare$

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**Problem 4**


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- (a) For  $x \in \{0, 1\}^n$ ,  $r \in [0, \infty)$ , let  $B(x, r) \subseteq \{0, 1\}^n$  denote the ball of Hamming radius  $r$  centered at  $x$ . By the given inequality, the cardinality of  $B(x, \tau n)$  is

$$|B(x, \tau n)| = \sum_{j=0}^{\tau n} \binom{n}{j} \leq 2^{h(\tau)n} \quad (1)$$

(there are  $\binom{n}{j}$  strings with Hamming distance exactly  $j$  from  $x$ , since we construct such a string by choosing  $j$  bits of  $x$  to flip). If  $C$  is a  $\tau$ -covering, then  $\{0, 1\}^n = \bigcup_{x \in C} B(x, \tau n)$ . Thus,

$$\begin{aligned} 2^n = |\{0, 1\}^n| &= \left| \bigcup_{x \in C} B(x, \tau n) \right| \leq \sum_{x \in C} |B(x, \tau n)| \quad (\text{by the Union Bound}) \\ &\leq \sum_{x \in C} 2^{h(\tau)n} = |C| 2^{h(\tau)n}, \quad (\text{by (1)}) \end{aligned}$$

which can be rewritten in the desired form:

$$|C| \geq \frac{2^n}{2^{h(\tau)n}} = 2^{(1-h(\tau))n}. \quad \blacksquare$$

(b) Couldn't get this one. ☹

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### Problem 5

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- (a) Clearly the leaves of the entire tree are  $\{a_1, a_2, \dots, a_n\}$ . Furthermore, if the leaves in the subtree rooted at some internal node  $N$  are  $\{a_i, a_{i+1}, \dots, a_j\}$  ( $1 \leq i < j \leq n$ ), then, since for some  $k \in \{i, i+1, \dots, j\}$ , the leaves in the subtrees rooted at the left and right children of  $N$  are  $\{i, i+1, \dots, k\}$  and  $\{k+1, k+2, \dots, j\}$ , so that both children of  $N$  have the desired property. Thus, by induction, the desired property holds for all internal nodes in the tree. ■
- (b) In the sum  $\sum_{[i,j] \in \mathcal{I}} q_{[i,j]}$ , each  $p_i$  appears once for each internal node which is an ancestor of  $[i, i]$ . Since the length  $l_i$  of the code for  $a_i$  is the number of ancestors of  $[i, i]$ ,

$$\sum_{[i,j] \in \mathcal{I}} q_{[i,j]} = \sum_{i \in \{1, 2, \dots, n\}} p_i l_i = L. \quad \blacksquare$$

- (c) For any node  $[i, j]$ , let  $S([i, j])$  denote the set of nodes in the subtree rooted at  $[i, j]$ . We show by induction that, for each internal node  $[i, j]$  in the tree,

$$H(X | X \in \{a_i, a_{i+1}, \dots, a_j\}) = \sum_{[i', j'] \in S([i, j])} \frac{q_{[i', j']}}{q_{[i, j]}} h\left(\frac{q_{[i', k']}}{q_{[i', j']}}\right),$$

which, in the case  $i = 1, j = n$ , reduces to the desired result. If  $i = j$ , this is trivial, since  $H(X | X = a_i) = 0 = h(1)$ . Suppose now, that the result holds for the left and right children of some internal node  $[i, j]$ . Then, letting  $D$  be a Bernoulli random variable with  $D = L$  if  $X \in \{a_i, a_{i+1}, \dots, a_k\}$  and  $D = R$  otherwise, computing conditional entropy as an expected value

$$\begin{aligned} H(X | X \in \{a_i, \dots, a_j\}) &= H(X | L, X \in \{a_i, \dots, a_j\}) + H(L) \\ &= \frac{q_{[i, k]}}{q_{[i, j]}} \left( \sum_{[i', j'] \in S([i, k])} \frac{q_{[i', j']}}{q_{[i, k]}} h\left(\frac{q_{[i', k']}}{q_{[i', j']}}\right) \right) \\ &\quad + \frac{q_{[k+1, j]}}{q_{[i, j]}} \left( \sum_{[i', j'] \in S([k+1, j])} \frac{q_{[i', j']}}{q_{[k+1, j]}} h\left(\frac{q_{[i', k']}}{q_{[i', j']}}\right) \right) + \frac{q_{[i, j]}}{q_{[i, j]}} h\left(\frac{q_{[i, k]}}{q_{[i, j]}}\right) \\ &= \sum_{[i', j'] \in S([i, j])} \frac{q_{[i', j']}}{q_{[i, j]}} h\left(\frac{q_{[i', k']}}{q_{[i', j']}}\right), \end{aligned}$$

so that the result holds for  $[i, j]$ . ■

(d) By the results of parts (b) and (c),

$$\begin{aligned}
L - H(X) &= \left( \sum_{[i,j] \in \mathcal{I}} q_{[i,j]} \right) - \sum_{[i,j] \in \mathcal{I}} q_{[i,j]} h\left(\frac{q_{[i,k]}}{q_{[i,j]}}\right) \\
&\leq \sum_{[i,j] \in \mathcal{I}} q_{[i,j]} - 2 \left( \frac{\min\{q_{[i,k]}, q_{[k+1,j]}\}}{q_{[i,j]}} \right) \quad (\text{since } h(1-x) = h(x) \text{ and } h(x) \geq 2x) \\
&= \sum_{[i,j] \in \mathcal{I}} |q_{[k+1,j]} - q_{[i,k]}|. \quad \blacksquare \quad (\text{since } q_{[i,j]} = q_{[i,k]} + q_{[k+1,j]})
\end{aligned}$$

(e) Suppose, for sake of contradiction, that, for some  $[i, j] \in \mathcal{I}$ ,  $|q_{[i,k]} - q_{[k+1,j]}| > \max\{p_k, p_{k+1}\}$ .  
If  $q_{[i,k]} < q_{[k+1,j]}$ ,

$$|q_{[i,k+1]} - q_{[k+2,j]}| = |q_{[i,k]} + p_{k+1} - (q_{[k+1,j]} - p_{k+1})| < |q_{[i,k]} - q_{[k+1,j]}|,$$

and, if  $q_{[i,k]} > q_{[k+1,j]}$

$$|q_{[i,k-1]} - q_{[k,j]}| = |q_{[i,k]} - p_k - (q_{[k+1,j]} + p_k)| < |q_{[i,k]} - q_{[k+1,j]}|,$$

Either case contradicts to choice of  $k = \operatorname{argmax}_{\ell: i \leq \ell < j} |q_{[i,\ell]} - q_{[\ell+1,j]}|$ .  $\blacksquare$

(f) By parts (d) and (e), since each  $p_i$  can be used at most twice (once as  $p_k$  and once as  $p_{k+1}$ )

$$L - H(X) \leq \sum_{[i,j] \in \mathcal{I}} |q_{[i,k]} - q_{[k+1,j]}| \leq \sum_{[i,j] \in \mathcal{I}} \max\{p_k, p_{k+1}\} \leq 2. \quad \blacksquare$$

## Problem 6

Pinsker's Inequality can be rewritten in the form

$$\sqrt{2D(p||q)} \geq \sum_{a \in A} |p(a) - q(a)|. \quad (2)$$

Thus, since mutual information is the divergence of joint and product distributions (shown in class),

$$\begin{aligned}
\sqrt{2I(X; Y)} &= \sqrt{2D(p(x, y); p(x)p(y))} \\
&\geq \sum_{(x,y) \in \mathcal{X} \times \mathcal{Y}} |p(x, y) - p(x)p(y)| \quad (\text{by (2)}) \\
&= \sum_{x \in \mathcal{X}} \sum_{y \in \mathcal{Y}} |p(y|x)p(x) - p(x)p(y)| \quad (\text{definition of conditional probability}) \\
&= \sum_{x \in \mathcal{X}} p(x) \sum_{y \in \mathcal{Y}} |p(y|x) - p(y)| \\
&= \sum_{x \in \mathcal{X}} p(x) d(x) = \mathbb{E}_{x \leftarrow \mathcal{X}} [d(x)]. \quad \blacksquare \quad (\text{definitions of } d, \text{ expected value})
\end{aligned}$$