IERG 6154: Network Information Theory Homework 1

Due: Jan 17, 2019

- 1. Inequalities: Show that
 - (a) $H(X|Z) \le H(X|Y) + H(Y|Z)$
 - (b) $I(X_1, X_2; Y_1, Y_2) \le I(X_1; Y_1) + I(X_2; Y_2)$ if $p(x_1, x_2, y_1, y_2) = p(x_1, x_2)p(y_1|x_1)p(y_2|x_2)$
 - (c) $I(X_1, X_2; Y_1, Y_2) \ge I(X_1; Y_1) + I(X_2; Y_2)$ if $p(x_1, x_2, y_1, y_2) = p(x_1)p(x_2)p(y_1, y_2|x_1, x_2)$
 - (d) $h(X+aY) \ge h(X+Y)$ when $Y \sim N(0,1), a \ge 1$. Assume that X and Y are independent.
- 2. Prove or give a counterexample to the following inequalities:
 - (a) $H(X_1, X_2, X_3) + H(X_1, X_2, X_4) + H(X_2, X_3, X_4) + H(X_1, X_3, X_4) \le \frac{3}{2} [H(X_1, X_2) + H(X_1, X_4) + H(X_2, X_3) + H(X_3, X_4)].$
 - (b) $H(X_1, X_2, X_3) + H(X_1, X_2, X_4) + H(X_2, X_3, X_4) + H(X_1, X_3, X_4) \le 3[H(X_1, X_2) + H(X_3, X_4)].$
- 3. (Korner-Marton Identity) Show the following equality for any pair of sequences Y^n, Z^n

$$\sum_{i=1}^{n} I(Y^{i-1}; Z_i | Z_{i+1}^n) = \sum_{i=1}^{n} I(Z_{i+1}^n; Y_i | Y^{i-1})$$

Note: Y^m and Y_n^m denote $Y_1, Y_2, ... Y_m$ and $Y_n, Y_{n+1}, ... Y_m$ respectively. There is an abuse of notation at the edges. Take Y^0 and Z_{n+1}^n as trivial random variables.

Remark: This equality was discovered by Marton and was first used in a joint paper with Korner (1977). This was conveyed to me by Janos Korner. Unfortunately, it has been mistakenly called Csiszar-sum Lemma by the community after attributing the result to a later paper by Csiszar and Korner.

4. Prove that when $f: \mathbb{R} \to [0, \frac{1}{2}]$ such that H(f(u)) is convex and f is twice differentiable, then H(f(u)*p) is convex in u for $p \in [0,1]$. (Fan Cheng's extension of Mrs. Gerber's lemma). In the above, for $0 \le x, y \le 1$ define x*y = x(1-y) + y(1-x). As a corollary note that $H(H^{-1}(u)*p)$ is convex in u for $p \in [0,1]$ where $H^{-1}(u)$ as a mapping from [0,1] to $[0,\frac{1}{2}]$.

- 5. (Pinsker's Inequality)
 - Let $x, y \in (-1, 1)$. Show that (assume natural logarithms)

$$\frac{1+x}{2}\log\left(\frac{1+x}{1+y}\right) + \frac{1-x}{2}\log\left(\frac{1-x}{1-y}\right) = \sum_{k=1}^{\infty} \left(\frac{x^{2k} + (2k-1)y^{2k} - 2kxy^{2k-1}}{2k(2k-1)}\right),$$

and in particular that all the terms on the right-hand-side are non-negative for each $k \geq 1$. Hence conclude that

$$\frac{1+x}{2}\log\left(\frac{1+x}{1+y}\right) + \frac{1-x}{2}\log\left(\frac{1-x}{1-y}\right) \ge \frac{1}{2}(x-y)^2.$$

• Given two distributions p and q, the total variation distance between them is defined as

$$d_{TV}(p,q) = \frac{1}{2} \sum_{x} |p(x) - q(x)|.$$

Show that

$$D(p||q) \ge 2d_{TV}^2(p,q).$$

(Hint: use data-processing inequality of D(p||q).)

- 6. (Han's inequality or Shearer's lemma)
 - Show that

$$H(X^n) \le \frac{1}{n-1} \sum_{i=1}^n H(X^{n \setminus i})$$

• More generally, show that given any collection $A_i \subseteq [1:n]$, then

$$a_*H(X^n) \le \sum_i H(X_{A_i}),$$

where $a_* = \min_k \left(\sum_i 1_{k \in A_i} \right)$.

Remark: The second generalization appears in a paper by Madiman and Tetali.

7. (Strong data processing inequalities for relative entropy) Given a channel W(y|x), define

$$\eta_W = \sup_{p,q:q \ll p} \frac{D(Wq||Wp)}{D(q||p)}.$$

Show the following:

- When W is $BEC(\epsilon)$ then $\eta_W = 1 \epsilon$
- When W is $BSC\left(\frac{1+\rho}{2}\right)$ then $\eta_W = \rho^2$
- When W is a Z-channel with W(Y=0|X=0)=1 and W(y=0|x=1)=z, then $\eta_W=1-z$.