

ECE 5630 - Extra Problem Set

May 18th, 2020

Instructions: This is an extra problem set, not for submission.

1) **Invariance to bijection:** Let X and Y be i.i.d. according to $\text{Unif}(\{0, 1, 2, 3\})$.

- Prove that $H(X + 4Y) = H(X, Y)$.
- Calculate $H(X + 4Y)$.

2) **KL-divergence computation:** Compute the following f -divergences:

- $D_{\text{KL}}(\text{Exp}(\eta_1) \parallel \text{Exp}(\eta_2))$, where $\text{Exp}(\eta)$ is the exponential distribution with parameter $\eta > 0$, i.e., the distribution whose PDF is $p^{(\eta)}(x) = \eta e^{-\eta x} \mathbb{1}_{\{x \geq 0\}}$.
- $D_{\text{KL}}(\text{Bin}(n, p_1) \parallel \text{Bin}(n, p_2))$, where $\text{Bin}(n, p)$ is the Binomial distribution with parameters $n \in \mathbb{N}$ and $p \in [0, 1]$, i.e., the distribution whose PMF is $p^{(n,p)}(k) = \binom{n}{k} p^k (1-p)^{n-k}$, for $k \in \{0, 1, 2, \dots, n\}$ (otherwise 0).
- $D_{\text{KL}}(\text{Geo}(p_1) \parallel \text{Geo}(p_2))$, where $\text{Geo}(p)$ is the geometric distribution with parameter $p \in [0, 1]$, i.e., the distribution whose PMF is $p^{(p)}(k) = (1-p)^{k-1} p$, for $k \in \mathbb{N}$ (otherwise 0).

3) **Shannon entropy and KL divergence**

Consider the following two distributions P and Q supported on $\{1, 2, \dots, L + M\}$, with PMFs:

$$p(x) = \begin{cases} p_x, & x = 1, \dots, L; \\ 0, & x = L + 1, \dots, L + M \end{cases}$$

$$q(x) = \begin{cases} \alpha p_x, & x = 1, \dots, L; \\ \frac{1-\alpha}{M}, & x = L + 1, \dots, L + M \end{cases}$$

where $0 < \alpha < 1$.

- Compute $D_{\text{KL}}(P \parallel Q)$
- Express $H(Q)$ in terms of $H(P)$, α and M .

4) **Differential entropy:** Let X , Z_1 , and Z_2 be independent Gaussian random variables with mean zero and variances $\mathbb{E}[X^2] = P$ and $\mathbb{E}[Z_1^2] = \mathbb{E}[Z_2^2] = N$. Let $Y_1 = g_1 X + Z_1$ and $Y_2 = g_2 X + Z_2$ for some constants $g_1, g_2 \in \mathbb{R}$. Express the following in terms of P, N, g_1 , and g_2 :

- $h(Z_1, Z_2)$.
- $h(Y_1, Y_2)$.
- $I(X; Y_1, Y_2)$.

5) **More differential entropy:** Let X, Y be jointly Gaussian with mean zero, variance one, and covariance $\rho \in (0, 1)$.

- What is $h(5X + 17)$
- What $h(X, Y)$
- What is $h(|X|)$?

6) **Entropy and KL divergence:** Let $\mathcal{X} = \{1, \dots, L + M\}$, for $M, N \in \mathbb{N}$ and consider the distributions $P, Q \in \mathcal{P}(\mathcal{X})$ whose PMFs are, respectively,

$$p(x) = \begin{cases} p_x, & x = 1, \dots, L; \\ 0, & x = L + 1, \dots, L + M. \end{cases}$$

$$q(x) = \begin{cases} \alpha p_x, & x = 1, \dots, L; \\ \frac{1-\alpha}{M}, & x = L + 1, \dots, L + M. \end{cases}$$

where $0 < \alpha < 1$.

- compute $D_{\text{KL}}(P||Q)$
- Express $H(Q)$ and in terms of $H(P)$, α and M .

7) **Axiomatic definition of entropy:** If a sequence of symmetric functions $H_m(p_1, p_2, \dots, p_m)$ satisfies the following properties:

- Normalization: $H_2(\frac{1}{2}, \frac{1}{2}) = 1$,
- Continuity: $H_2(p, 1 - p)$ is a continuous function of p ,
- Grouping: $H_m(p_1, p_2, \dots, p_m) = H_{m-1}(p_1 + p_2, p_3, \dots, p_m) + (p_1 + p_2)H_2\left(\frac{p_1}{p_1 + p_2}, \frac{p_2}{p_1 + p_2}\right)$,

prove that H_m must be of the form

$$H_m(p_1, p_2, \dots, p_m) = -\sum_{i=1}^m p_i \log(p_i).$$

Hint 1: Using induction show that

$$H_m(p_1, p_2, \dots, p_m) = H_{m-1}(p_1 + \dots + p_k, p_{k+1}, \dots, p_m) + (p_1 + \dots + p_k)H_2\left(\frac{p_1}{p_1 + \dots + p_k}, \dots, \frac{p_k}{p_1 + \dots + p_k}\right)$$

for all $k = 1, \dots, m$.

Hint 2: Let $f(m) = H_m(1/m, 1/m, \dots, 1/m)$. Show that for two integers i and j , it holds that $f(ij) = f(i) + f(j)$.

Hint 3: Prove that $H_2(p, 1 - p) = -p \log p - (1 - p) \log(1 - p)$ for any rational p . Use continuity to extend the argument to real numbers.

8) **Entropy under constraints:** Let $X, Y, Z \sim \text{Ber}(1/2)$ and pairwise independent ($I(X; Y) = I(Y; Z) = I(X; Z) = 0$).

- Under this constraint, what is the minimum value for $H(X, Y, Z)$?
- Given an example achieving this minimum.

Now instead of pairwise independence, assume that $I(X; Y) = I(Y; Z) = I(X; Z) = \alpha$ for some $\alpha \in [0, 1]$. Repeat parts (a) and (b).

- 9) **Directed information:** The *directed information* $I(X^n \rightarrow Y^n)$ from $X^n := (X_1, \dots, X_n)$ to $Y^n := (Y_1, \dots, Y_n)$ (random correlated sequences) is an information measure that appears in the context of interactive communication and communication with feedback. It is defined as

$$I(X^n \rightarrow Y^n) = \sum_{i=1}^n I(X^i; Y_i | Y^{i-1}) \quad (1)$$

That is, it is the sum of the the mutual information of inputs up to time i and the output at time i conditioned on the past outputs up to time $i - 1$. For this problem you can restrict yourself to considering discrete sources only (although this is not necessary).

- (a) Show that $I(X^n \rightarrow Y^n) \neq I(Y^n \rightarrow X^n)$ in general.

Hint: consider X^n and Y^n to be certain subsets of $\{Z_0, Z_1, \dots, Z_n\}$ that are i.i.d. Bern(1/2)).

- (b) Consider a DMC used for n channel uses with input X^n and output Y^n . Here we *do not* assume that X^n is i.i.d. Show that in general,

$$I(X^n \rightarrow Y^n) \leq \sum_{i=1}^n I(X_i; Y_i). \quad (2)$$

Make sure you justify each step.

- (c) What happens to (2) when Y_1, Y_2, \dots, Y_n are independent?

- 10) **Simulating a Gaussian distribution:** In this question we are going to write a code (Matlab/Python/etc.) for simulating a Gaussian distribution via the soft-covering lemma setup. Consider the additive white Gaussian noise channel (AWGN) whose output at time $i = 1, \dots, n$ is $V_i = u_i + Z_i$, where $u_i \in \mathbb{R}$ is the channel input and $Z_i \sim \mathcal{N}(0, \sigma^2)$ are i.i.d. Gaussians.

- a) Implement the AWGN channel output function $\text{AWGN}(n, \sigma, \mathbf{u})$, that takes as inputs a blocklength $n \in \mathbb{N}$, a noise parameter $\sigma \in \mathbb{R}_{>0}$ and an input sequence $\mathbf{u} \in \mathbb{R}^n$, and produces a sample of the (random) output sequence (Y_1, \dots, Y_n) , for Y_i as above.
- b) Next implement a randomly generated Gaussian code. Let $\text{Code}(n, W, \eta)$ be the function that takes as input the blocklength $n \in \mathbb{N}$, a codebook size $W \in \mathbb{N}$, and an input standard deviation parameter $\eta \in \mathbb{R}_{>0}$, and outputs a collection $\{\mathbf{u}(w)\}_{w=1}^W$ of i.i.d. (across codewords and across time) sequences of length n , where each symbol $u_i(w)$, for $i = 1, \dots, n$ and $w = 1, \dots, W$, is drawn according to $\mathcal{N}(0, \eta^2)$.
- c) Show that the induced output probability density function $q_{\mathbf{V}} : \mathbb{R}^n \rightarrow \mathbb{R}_{\geq 0}$ is the Gaussian mixture model

$$q_{\mathbf{V}}(\mathbf{v}) = \frac{1}{W} \sum_{w=1}^W \varphi_{\sigma}(\mathbf{v} - \mathbf{u}(w)), \quad (3)$$

where $\varphi_{\sigma}(\mathbf{x}) = \frac{1}{(2\pi\sigma)^{n/2}} e^{-\frac{\|\mathbf{x}\|_2^2}{2\sigma^2}}$ is the density of $\mathcal{N}(\mathbf{0}, \sigma^2 \mathbf{I}_n)$ and \mathbf{I}_n is the $n \times n$ identity matrix.

- d) Let $U \sim \mathcal{N}(0, \eta^2)$ be the coding variable and $V = U + Z$ be the (single-letter) channel output, where $Z \sim \mathcal{N}(0, \sigma^2)$ is independent of U . Show that the target distribution for fixed η and σ , i.e., the marginal distribution of V above, is $\mathcal{N}(\mathbf{0}, (\eta^2 + \sigma^2)\mathbf{I}_n)$, and write out its probability density function $\varphi_{\sqrt{\eta^2 + \sigma^2}}(\mathbf{v})$ explicitly.
- e) Compute $I(U, V)$ in terms of η and σ .
- f) Fix $\eta = \sigma = 1$, $n \in \{1, 2\}$ (implement both cases) and let W range from 1 to 10^4 (choose appropriate gaps). For

each W (and n), use the function $\text{Code}(n, W, \eta)$ to produce a Gaussian codebook. Compute and plot $q_{\mathbf{v}}(v)$ from (3) versus $\varphi_{\sqrt{\eta^2 + \sigma^2}}(\mathbf{v})$, for $v_i \in [-6, 6]$, $i = 1, \dots, n$. Also plot the (scaled) conditional distributions $q_{\mathbf{v}|W}(\mathbf{v}|w) = \varphi_{\sigma}(\mathbf{v} - \mathbf{u}(w))$, for $w = 1, \dots, W$, on the same axes. Repeat this experiment for each considered W . Does the approximation of $\varphi_{\sqrt{\eta^2 + \sigma^2}}$ via $q_{\mathbf{v}}$ improves as W grows? How do the results differ between the $n = 1$ and $n = 2$ cases?

g) Plot the total variation distance

$$\delta_{\text{TV}}(q_{\mathbf{v}}, \varphi_{\sqrt{\eta^2 + \sigma^2}}) = \frac{1}{2} \int_{\mathbb{R}^n} |q_{\mathbf{v}}(\mathbf{v}) - \varphi_{\sqrt{\eta^2 + \sigma^2}}(\mathbf{v})| d\mathbf{v}$$

versus the range of W values. Describe and explain the curve you obtain.

11) **Multiple cascaded BSCs:** In this problem we study a generalization of the cascade of BSCs from Question 7 of Homework Sheet 5. Consider a cascade of k identical and independent binary symmetric channels, each with crossover probability α .

- In the case where no encoding or decoding is allowed at the intermediate terminals, what is the capacity of this cascaded channel as a function of k, α ?
- Now, assume that encoding and decoding is allowed at the intermediate points, what is the capacity as a function of k, α ?
- What is the capacity of each of the above settings in the case where the number of cascaded channels, k , goes to infinity?

12) **Entropy power inequality:** A famous (and highly useful) information inequality is the *entropy power inequality (EPI)*.

Lemma (Entropy power inequality) *Let X and Y be two real-valued independent random variables. Then,*

$$e^{2h(X+Y)} \geq e^{2h(X)} + e^{2h(Y)}, \quad (4)$$

with equality if and only if X and Y are jointly Gaussian.

Let us consider a special case of that result. Suppose X and Y are two independent random variables with density functions

$$f_X(x) = \begin{cases} \frac{1}{2a} & |x| \leq a, \\ 0 & |x| > a \end{cases}$$

and

$$f_Y(y) = \begin{cases} \frac{1}{2b} & |y| \leq b, \\ 0 & |y| > b \end{cases}$$

for some arbitrary $0 < a \leq b$.

- Compute $h(X)$ and $h(Y)$.
- Find the probability density function of $Z = X + Y$. You may solve analytically or rely on a carefully labeled graphical solution.
- Find $h(Z)$.

Hint: For $\beta \geq \alpha$, we have

$$\int_{\alpha}^{\beta} x \log x dx = \frac{1}{2} \beta^2 \log \beta - \frac{1}{2} \alpha^2 \log \alpha - \frac{\log e}{4} (\beta^2 - \alpha^2).$$

- 13) **Erasures and errors in a binary channel:** Consider a binary channel with probability of error α and probability of erasure ϵ as depicted in Figure 1. More specifically, consider $(\mathcal{X}, \mathcal{Y}, P_{Y|X})$ where $\mathcal{X} = \{0, 1\}$, $\mathcal{Y} = \{0, 1, e\}$ and $P_{Y|X}$ described by the relation:

$$Y = \begin{cases} X, & \text{w.p. } 1 - \alpha - \epsilon, \\ 1 - X, & \text{w.p. } \alpha, \\ e, & \text{w.p. } \epsilon. \end{cases}$$

Find a closed form expression for the capacity $\max_{P_X} I(X; Y)$ of this channel.

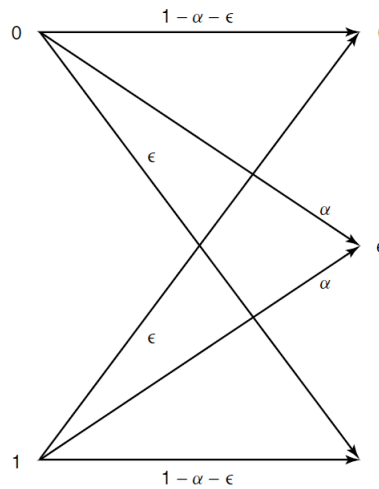


Fig. 1: Erasures and errors in a binary channel

- 14) **Modulus channel:** Consider a discrete channel with input alphabet $\mathcal{X} = \{0, 1, \dots, q-1\}$. The channel output is

$$Y = [X + Z] \bmod q$$

where Z is independent of X with $p_Z(0) = 1 - \beta$ and $p_Z(z) = \frac{\beta}{q-1}$ for $z = 1, 2, \dots, q-1$.

- What is $H(Z)$?
- What is the capacity of this channel?

- 15) **Time varying channels:** Consider a time varying binary symmetric channel. More specifically, at time $i = 1, \dots, n$, the channel is specified by $(\mathcal{X}, \mathcal{Y}, P_{Y_i|X_i})$, where $\mathcal{X} = \mathcal{Y} = \{0, 1\}$ and $P_{Y_i|X_i}$ is described by the relationship $Y_i = X_i \oplus Z_i$, where $Z_i \sim \text{Bern}(p_i)$ with $p_i \in (0, 1)$. Assume that $\{Z_i\}_{i=1}^n$ are independent, and, thus, Y_i 's are conditionally independent given X_i 's. Find $\max_{P_X^n} I(X^n; Y^n)$, where the underlying distribution is $P_X^n \prod_{i=1}^n P_{Y_i|X_i}$.

- 16) **Computing channel capacity:** Consider a channel $(\mathcal{X}, \mathcal{Y}, P_{Y|X})$, where $\mathcal{X} = \mathcal{Y} = \{0, 1, 2\}$ and $P_{Y|X}$ has a conditional PMF $p_{Y|X}$ given by

$$p_{Y|X} = \begin{bmatrix} 2/3 & 1/3 & 0 \\ 1/3 & 1/3 & 1/3 \\ 0 & 1/3 & 2/3 \end{bmatrix}$$

- Find the capacity $\max_{P_X} I(X; Y)$ and the distribution that achieves it.
- Qualitatively justify why the distribution found in part (a) achieves the capacity.