Homework 1

Due: February 2, 2022

1. Example of joint entropy. Let p(x,y) be given by

Find:

(a) H(X), H(Y).

(b) H(X|Y), H(Y|X).

(c) H(X,Y).

(d) H(Y) - H(Y|X).

(e) I(X;Y).

(f) Draw a Venn diagram for the quantities in parts (a) through (e).

2. Problem 2.4 from Cover-Thomas: Entropy of functions of a random variable. Let X be a discrete random variable. Show that the entropy of a function of X is less than or equal to the entropy of X by justifying the following steps:

$$\begin{split} H(X,g(X)) &\stackrel{\text{(a)}}{=} H(X) + H(g(X)|X) \\ &\stackrel{\text{(b)}}{=} H(X), \\ H(X,g(X)) &\stackrel{\text{(c)}}{=} H(g(X)) + H(X|g(X)) \\ &\stackrel{\text{(d)}}{\geq} H(g(X)). \end{split}$$

Thus, $H(g(X)) \leq H(X)$.

3. Yes/No question. Let X be a random variable with PMF given by

Let Y be a random variable with PMF given by

- (a) Devise a strategy to determine X by a series of Yes/No questions such that the expected number of questions is exactly equal to H(X). Each question may depend on the outcome of the previous question.
- (b) For Y, find a strategy of Yes/No questions to minimize the expected number of questions. What is the smallest possible expected number of questions? In this case, how does it compare to H(Y)?
- 4. Mutual information vs. conditional mutual information. In general, conditioning neither decreases nor increases mutual information; i.e. either the mutual information I(X;Y) or the conditional mutual information I(X;Y|Z) can be larger. This problem shows that under certain conditions, one can conclude that one of these quantities is larger than the other. Prove the following two statements:
 - (a) If $X \to Y \to Z$, then $I(X;Y|Z) \le I(X;Y)$.
 - (b) If X and Z are independent, then $I(X;Y) \leq I(X;Y|Z)$.

Hint: Expand I(X;Y,Z) using the chain rule in two different ways.

5. Problem 2.27 from Cover-Thomas: Grouping rule for entropy. Let $\mathbf{p} = (p_1, p_2, \dots, p_m)$ be a probability distribution on m elements (i.e., $p_i \geq 0$ and $\sum_{i=1}^m p_i = 1$). Define a new distribution \mathbf{q} on m-1 elements as $q_1 = p_1, q_2 = p_2, \dots, q_{m-2} = p_{m-2}$, and $q_{m-1} = p_{m-1} + p_m$ [i.e., the distribution \mathbf{q} is the same as \mathbf{p} on $\{1, 2, \dots, m-2\}$, and the probability of the last element in \mathbf{q} is the sum of the last two probabilities of \mathbf{p}]. Show that

$$H(\mathbf{p}) = H(\mathbf{q}) + (p_{m-1} + p_m)H\left(\frac{p_{m-1}}{p_{m-1} + p_m}, \frac{p_m}{p_{m-1} + p_m}\right).$$

6. Rényi Entropy. Rényi entropy is a different way of defining an "entropy" that generalizes Shannon's measure. Rényi entropy has a parameter α , which can be any positive number except 1. Rényi entropy is defined as

$$H_{\alpha}(X) = \frac{1}{1-\alpha} \log \left[\sum_{x \in \mathcal{X}} p(x)^{\alpha} \right].$$

- (a) Find $H_{\alpha}(X)$ if X is a uniform random variable with an alphabet of size m. Compare your answer to the corresponding value for the standard entropy.
- (b) Plot $H_{\alpha}(X)$ as a function of α where X is Bern(0.1).
- (c) Prove that $\lim_{\alpha\to 1} H_{\alpha}(X) = H(X)$, where H(X) is the standard entropy. For this reason, the standard entropy is sometimes written $H_1(X)$. Hint: Use L'Hôpital's rule.
- (d) If X and Y are independent, prove that $H_{\alpha}(X,Y) = H_{\alpha}(X) + H_{\alpha}(Y)$.
- (e) Show that Rényi entropy does *not* satisfy the grouping rule, as defined in Problem 5. To do this, find a distribution **p** where the formula in Problem 5 is violated for Rényi entropy.
- 7. Problem 2.32 from Cover-Thomas: Fano. We are given the following joint distribution on (X,Y):

Let $\hat{X}(Y)$ be an estimator for X (based on Y) and let $P_e = \Pr{\{\hat{X}(Y) \neq X\}}$.

- (a) Find the minimum probability of error estimator $\hat{X}(Y)$ and the associated probability P_e .
- (b) Evaluate Fano's inequality for this problem and compare.
- 8. Problem 2.37 from Cover-Thomas: Relative entropy. Let X, Y, Z be three random variables with a joint probability mass function p(x, y, z). The relative entropy between joint distribution and the product of the marginals is

$$D(p(x,y,z)\|p(x)p(y)p(z)) = E\left[\log\frac{p(x,y,z)}{p(x)p(y)p(z)}\right].$$

Expand this in terms of entropies. When is this quantity zero?

- 9. Problem 2.42 from Cover-Thomas: *Inequalities*. Which of the following inequalities are generally \geq , =, \leq ? Label each with \geq , =, or \leq .
 - (a) H(5X) vs. H(X)
 - (b) I(g(X); Y) vs. I(X; Y)
 - (c) $H(X_0|X_{-1})$ vs. $H(X_0|X_{-1},X_1)$
 - (d) H(X,Y)/(H(X) + H(Y)) vs. 1