

Solutions to Homework Set #7

1. Postprocessing the output.

One is given a communication channel with transition probabilities $p(y | x)$ and channel capacity $C = \max_{p(x)} I(X; Y)$. A helpful statistician postprocesses the output by forming $\tilde{Y} = g(Y)$, yielding a channel $p(\tilde{y}|x)$. He claims that this will strictly improve the capacity.

- (a) Show that he is wrong.
- (b) Under what conditions does he not strictly decrease the capacity?

Solution: Postprocessing the output.

- (a) The statistician calculates $\tilde{Y} = g(Y)$. Since $X \rightarrow Y \rightarrow \tilde{Y}$ forms a Markov chain, we can apply the data processing inequality. Hence for every distribution on x ,

$$I(X; Y) \geq I(X; \tilde{Y}).$$

Let $\tilde{p}(x)$ be the distribution on x that maximizes $I(X; \tilde{Y})$. Then

$$C = \max_{p(x)} I(X; Y) \geq I(X; Y)_{p(x)=\tilde{p}(x)} \geq I(X; \tilde{Y})_{p(x)=\tilde{p}(x)} = \max_{p(x)} I(X; \tilde{Y}) = \tilde{C}.$$

Thus, the helpful suggestion is wrong and processing the output does not increase capacity.

- (b) We have equality (no decrease in capacity) in the above sequence of inequalities only if we have equality in data processing inequality, i.e., for the distribution that maximizes $I(X; \tilde{Y})$, we have $X \rightarrow \tilde{Y} \rightarrow Y$ forming a Markov chain. Thus, \tilde{Y} should be a sufficient statistic.

2. Noisy typewriter.

Consider a 26-key typewriter.

- (a) If pushing a key results in printing the associated letter, what is the capacity C in bits?
- (b) Now suppose that pushing a key results in printing that letter or the next (with equal probability). Thus $A \rightarrow A$ or B , and $Z \rightarrow Z$ or A . What is the capacity?
- (c) What is the highest rate code with block length one that you can find that achieves *zero* probability of error for the channel in part (b) .

Solution: Noisy typewriter.

- (a) If the typewriter prints out whatever key is struck, then the output Y is the same as the input X and

$$C = \max I(X; Y) = \max H(X) = \log 26, \quad (1)$$

attained by a uniform distribution over the letters.

- (b) In this case, the output is either equal to the input (with probability $\frac{1}{2}$) or equal to the next letter (with probability $\frac{1}{2}$). Hence $H(Y|X) = \log 2$ independent of the distribution of X , and hence

$$C = \max I(X; Y) = \max H(Y) - \log 2 = \log 26 - \log 2 = \log 13, \quad (2)$$

which is attained for a uniform distribution over the output, which in turn is attained by a uniform distribution on the input.

- (c) A simple zero error block length one code is the one that uses every alternate letter, say A,C,E,...,W,Y. In this case, none of the codewords will be confused, since A will produce either A or B, C will produce C or D, etc. The rate of this code,

$$R = \frac{\log(\# \text{ codewords})}{\text{Block length}} = \frac{\log 13}{1} = \log 13. \quad (3)$$

In this case, we can achieve capacity with a simple code with zero error. Note that the uniform distribution over the output is attained also by this input distribution.

3. The Z Channel.

The Z-channel has binary input and output alphabets and transition probabilities $p(y|x)$ given by the following matrix:

$$Q = \begin{bmatrix} 1 & 0 \\ 1/2 & 1/2 \end{bmatrix} \quad x, y \in \{0, 1\}$$

Find the capacity of the Z-channel and the maximizing input probability distribution.

Solution: The Z channel.

First we express $I(X; Y)$, the mutual information between the input and output of the Z-channel, as a function of $\alpha = \Pr(X = 1)$:

$$\begin{aligned}H(Y|X) &= \Pr(X = 0) \cdot 0 + \Pr(X = 1) \cdot 1 = \alpha \\H(Y) &= \mathbf{H}(\Pr(Y = 1)) = \mathbf{H}(\alpha/2) \\I(X; Y) &= H(Y) - H(Y|X) = \mathbf{H}(\alpha/2) - \alpha\end{aligned}$$

Since $I(X; Y)$ is strictly concave on α (why?) and $I(X; Y) = 0$ when $\alpha = 0$ and $\alpha = 1$, the maximum mutual information is obtained for some value of α such that $0 < \alpha < 1$.

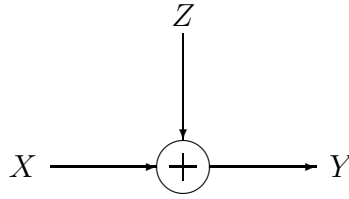
Using elementary calculus, we determine that

$$\frac{d}{d\alpha} I(X; Y) = \frac{1}{2} \log_2 \frac{1 - \alpha/2}{\alpha/2} - 1,$$

which is equal to zero for $\alpha = 2/5$. (It is reasonable that $\Pr(X = 1) < 1/2$ since $X = 1$ is the noisy input to the channel.) So the capacity of the Z-channel in bits is $H(1/5) - 2/5 = 0.722 - 0.4 = 0.322$.

4. **An additive noise channel.**

Find the channel capacity of the following discrete memoryless channel:



where $\Pr\{Z = 0\} = \Pr\{Z = a\} = \frac{1}{2}$. The alphabet for x is $\mathcal{X} = \{0, 1\}$. $Y = X + Z$ with real addition. Assume that Z is independent of X .

Observe that the channel capacity depends on the value of a .

Solution: An additive noise channel.

$$Y = X + Z \quad X \in \{0, 1\}, \quad Z \in \{0, a\}$$

We have to distinguish various cases depending on the values of a .

$a = 0$ In this case, $Y = X$, and $\max I(X; Y) = \max H(X) = 1$. Hence the capacity is 1 bit per transmission.

$a \neq 0, \pm 1$ In this case, Y has four possible values $0, 1, a$ and $1 + a$. Knowing Y , we know the X which was sent, and hence $H(X|Y) = 0$. Hence $\max I(X; Y) = \max H(X) = 1$, which is achieved for an uniform distribution on the input X .

$a = 1$ In this case Y has three possible output values, $0, 1$ and 2 , and the channel is identical to the binary erasure channel discussed in class, with $\alpha = 1/2$. As derived in class, the capacity of this channel is $1 - \alpha = 1/2$ bit per transmission.

$a = -1$ This is similar to the case when $a = 1$ and the capacity here is also $1/2$ bit per transmission.

5. **Channels with memory have higher capacity.**

Consider a binary symmetric channel with $Y_i = X_i \oplus Z_i$, where \oplus is mod 2 addition, and $X_i, Y_i \in \{0, 1\}$.

Suppose that $\{Z_i\}$ has constant marginal probabilities $P\{Z_i = 1\} = p = 1 - P\{Z_i = 0\}$, but that Z_1, Z_2, \dots, Z_n are not necessarily independent. Let $C = 1 - H(p, 1 - p)$. Show that $\max_{p(x_1, x_2, \dots, x_n)} I(X_1, X_2, \dots, X_n; Y_1, Y_2, \dots, Y_n) \geq nC$. Comment on the implications.

Solution: Channels with memory have higher capacity.

$$Y_i = X_i \oplus Z_i,$$

where

$$Z_i = \begin{cases} 1 & \text{with probability } p \\ 0 & \text{with probability } 1 - p \end{cases}$$

and Z_i are not independent.

When X_1, X_2, \dots, X_n are chosen i.i.d. $\sim \text{Bern}(\frac{1}{2})$,

$$\begin{aligned} I(X_1, X_2, \dots, X_n; Y_1, Y_2, \dots, Y_n) &= H(X_1, X_2, \dots, X_n) - H(X_1, X_2, \dots, X_n | Y_1, Y_2, \dots, Y_n) \\ &= H(X_1, X_2, \dots, X_n) - H(Z_1, Z_2, \dots, Z_n | Y_1, Y_2, \dots, Y_n) \\ &\geq H(X_1, X_2, \dots, X_n) - H(Z_1, Z_2, \dots, Z_n) \\ &\geq H(X_1, X_2, \dots, X_n) - \sum H(Z_i) \\ &= n - nH(p). \end{aligned}$$

Hence, the capacity of the channel with memory over n uses of the channel is

$$\begin{aligned} nC^{(n)} &= \max_{p(X_1, X_2, \dots, X_n)} I(X_1, X_2, \dots, X_n; Y_1, Y_2, \dots, Y_n) \\ &\geq I(X_1, X_2, \dots, X_n; Y_1, Y_2, \dots, Y_n)_{p(x_1, x_2, \dots, x_n) = \text{Bern}(\frac{1}{2})} \\ &\geq n(1 - H(p)) \\ &= nC. \end{aligned}$$

Hence, channels with memory have higher capacity. The intuitive explanation for this result is that the correlation between the noise decreases the effective noise; one could use the information from the past samples of the noise to combat the present noise.

6. Channel capacity.

Consider the channel $Y = X + Z \pmod{13}$, where

$$Z = \begin{pmatrix} 1, & 2, & 3, & 4 \\ \frac{1}{4}, & \frac{1}{4}, & \frac{1}{4}, & \frac{1}{4} \end{pmatrix}$$

and $X \in \{0, 1, \dots, 12\}$.

- (a) Find the capacity.
- (b) What is the maximizing $p^*(x)$?

Solution: Channel capacity.

$$Y = X + Z \pmod{13}$$

where

$$Z = \begin{cases} 1 & \text{with probability } 1/4 \\ 2 & \text{with probability } 1/4 \\ 3 & \text{with probability } 1/4 \\ 4 & \text{with probability } 1/4 \end{cases}.$$

Then,

$$H(Y|X) = H(Z|X) = H(Z) = \log 4,$$

which is independent of the distribution of X . Hence the capacity of the channel is

$$\begin{aligned} C &= \max_{p(x)} I(X; Y) \\ &= \max_{p(x)} H(Y) - H(Y|X) \\ &= \max_{p(x)} H(Y) - \log 4 \\ &= \log 13 - \log 4, \end{aligned}$$

which is attained when Y has a uniform distribution, which occurs (by symmetry) when X has a uniform distribution.

- (a) The capacity of the channel is $\log \frac{13}{4}$ bits/transmission.
- (b) The capacity is achieved by a uniform distribution on the inputs, that is,

$$p(X = i) = \frac{1}{13} \quad \text{for } i = 0, 1, \dots, 12.$$

7. Using two channels at once.

Consider two discrete memoryless channels $(\mathcal{X}_1, p(y_1 | x_1), \mathcal{Y}_1)$ and $(\mathcal{X}_2, p(y_2 | x_2), \mathcal{Y}_2)$ with capacities C_1 and C_2 respectively. A new channel $(\mathcal{X}_1 \times \mathcal{X}_2, p(y_1 | x_1) \times p(y_2 | x_2), \mathcal{Y}_1 \times \mathcal{Y}_2)$ is formed in which $x_1 \in \mathcal{X}_1$ and $x_2 \in \mathcal{X}_2$, are *simultaneously* sent, resulting in y_1, y_2 . Find the mutual information maximizing $p^*(x_1, x_2)$ in terms of the individual maximizing distributions $p^*(x_1)$ and $p^*(x_2)$. Find the capacity of this channel.

Solution: Using two channels at once.

To find the capacity of the product channel $(\mathcal{X}_1 \times \mathcal{X}_2, p(y_1, y_2 | x_1, x_2), \mathcal{Y}_1 \times \mathcal{Y}_2)$, we have to find the distribution $p(x_1, x_2)$ on the input alphabet $\mathcal{X}_1 \times \mathcal{X}_2$ that maximizes $I(X_1, X_2; Y_1, Y_2)$. Since the transition probabilities are given as $p(y_1, y_2 | x_1, x_2) = p(y_1 | x_1)p(y_2 | x_2)$,

$$\begin{aligned} p(x_1, x_2, y_1, y_2) &= p(x_1, x_2)p(y_1, y_2 | x_1, x_2) \\ &= p(x_1, x_2)p(y_1 | x_1)p(y_2 | x_2), \end{aligned}$$

Therefore, $Y_1 \rightarrow X_1 \rightarrow X_2 \rightarrow Y_2$ forms a Markov chain and

$$\begin{aligned} I(X_1, X_2; Y_1, Y_2) &= H(Y_1, Y_2) - H(Y_1, Y_2 | X_1, X_2) \\ &= H(Y_1, Y_2) - H(Y_1 | X_1, X_2) - H(Y_2 | X_1, X_2) \end{aligned} \quad (4)$$

$$= H(Y_1, Y_2) - H(Y_1 | X_1) - H(Y_2 | X_2) \quad (5)$$

$$\leq H(Y_1) + H(Y_2) - H(Y_1 | X_1) - H(Y_2 | X_2) \quad (6)$$

$$= I(X_1; Y_1) + I(X_2; Y_2),$$

where Eqs. (4) and (5) follow from Markovity, and Eq. (6) is met with equality if X_1 and X_2 are independent and hence Y_1 and Y_2 are independent. Therefore

$$\begin{aligned} C &= \max_{p(x_1, x_2)} I(X_1, X_2; Y_1, Y_2) \\ &\leq \max_{p(x_1, x_2)} I(X_1; Y_1) + \max_{p(x_1, x_2)} I(X_2; Y_2) \\ &= \max_{p(x_1)} I(X_1; Y_1) + \max_{p(x_2)} I(X_2; Y_2) \\ &= C_1 + C_2. \end{aligned}$$

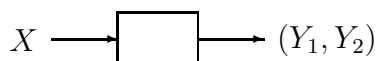
with equality iff $p(x_1, x_2) = p^*(x_1)p^*(x_2)$ and $p^*(x_1)$ and $p^*(x_2)$ are the distributions that maximize C_1 and C_2 respectively.

8. A channel with two independent looks at Y .

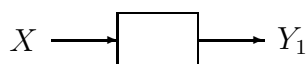
Let Y_1 and Y_2 be conditionally independent and conditionally identically distributed given X . Thus $p(y_1, y_2 | x) = p(y_1 | x)p(y_2 | x)$.

(a) Show $I(X; Y_1, Y_2) = 2I(X; Y_1) - I(Y_1; Y_2)$.

(b) Conclude that the capacity of the channel



is less than twice the capacity of the channel



(c) How about 3 independent looks? Compare $I(X; Y_1, Y_2, Y_3)$ to $3I(X; Y_1)$.

Solution: A channel with two independent looks at Y .

- (a) First note that Y_1 and Y_2 are identically distributed since $p(y_1|x) = p(y_2|x)$ and hence $p(y_1) = \sum_x p(y_1|x)p(x) = \sum_x p(y_2|x)p(x) = p(y_2)$. Therefore,

$$\begin{aligned} I(X; Y_1, Y_2) &= H(Y_1, Y_2) - H(Y_1, Y_2|X) \\ &= H(Y_1) + H(Y_2) - I(Y_1; Y_2) - H(Y_1, Y_2|X) \\ &= H(Y_1) + H(Y_2) - I(Y_1; Y_2) - H(Y_1|X) - H(Y_2|X) \end{aligned} \quad (7)$$

$$\begin{aligned} &= 2H(Y_1) - 2H(Y_1|X) - I(Y_1; Y_2) \\ &= 2I(X; Y_1) - I(Y_1; Y_2), \end{aligned} \quad (8)$$

where Eq. (7) follows from the fact that Y_1 and Y_2 are conditionally independent given X and Eq. (8) follows from the fact that Y_1 and Y_2 are identically distributed and conditionally identically distributed given X .

- (b) The capacity of the single look channel $X \rightarrow Y_1$ is

$$C_1 = \max_{p(x)} I(X; Y_1).$$

The capacity of the channel $X \rightarrow (Y_1, Y_2)$ is

$$\begin{aligned} C_2 &= \max_{p(x)} I(X; Y_1, Y_2) \\ &= \max_{p(x)} 2I(X; Y_1) - I(Y_1; Y_2) \\ &\leq \max_{p(x)} 2I(X; Y_1) \\ &= 2C_1. \end{aligned}$$

Hence, two independent looks cannot be more than twice as good as one look.

- (c) Observe for Y^n conditionally independent given X that

$$\begin{aligned} I(X; Y_k | Y^{k-1}) &= H(Y_k | Y^{k-1}) - H(Y_k | Y^{k-1}, X) \\ &= H(Y_k | Y^{k-1}) - H(Y_k | X) \end{aligned} \quad (9)$$

$$\begin{aligned} &\leq H(Y_k) - H(Y_k | X) \\ &= I(X; Y_k) \end{aligned} \quad (10)$$

where Eq. (9) follows from the fact that Y_k and Y^{k-1} are conditionally independent given X and Eq. (10) follows from the fact that conditioning reduces entropy.

Using the above relationship,

$$\begin{aligned} I(X; Y_1, Y_2, Y_3) &= I(X; Y_1) + I(X; Y_2 | Y_1) + I(X; Y_3 | Y_1, Y_2) \\ &\leq I(X; Y_1) + I(X; Y_2) + I(X; Y_3) \end{aligned} \quad (11)$$

$$= 3I(X; Y_1) \quad (12)$$

where Eq. (11) follows from the relationship shown above and Eq. (12) follows from the fact that Y_1, Y_2 and Y_3 are identically distributed.

Thus it can be shown that the capacity C_3 of three looks is less than three times the capacity C_1 of one look, $C_3 < 3C_1$.

9. **Can signal alternatives lower capacity?**

Show that adding a row to a channel transition matrix does not decrease capacity.

Solution: Can signal alternatives lower capacity?

Adding a row to the channel transition matrix is equivalent to adding a symbol to the input alphabet \mathcal{X} . Suppose there were m symbols and we add an $(m + 1)$ st. We can always choose not to use this extra symbol.

Alternatively, let C_m and C_{m+1} denote the capacity of the original channel and the new channel, respectively. Since using the original m symbols is same as using $m + 1$ symbols with the newly added symbol assigned zero probability,

$$\begin{aligned} C_{m+1} &= \max_{p(x_1, \dots, x_{m+1})} I(X; Y) \\ &\geq \max_{p(x_1, \dots, x_m, 0)} I(X; Y) \\ &= C_m. \end{aligned}$$