

ECE 231A HW 3

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Problem 1

(a)

Then we have

$$I(X; Y|Z) = H(Y|Z) - H(Y|X, Z) = H(Y|Z) = \alpha H(p)$$

Thus we have that the capacity is α

(b)

We have

$$\begin{aligned} I(X; Y) &= H(Y) - H(Y|X) \\ &= -p\alpha \log_2(p\alpha) - (1 - p\alpha) \log_2((1 - p\alpha)) - H(Y|X) \\ &= -p\alpha \log_2(p\alpha) - (1 - p\alpha) \log_2((1 - p\alpha)) - pH(\alpha) \end{aligned}$$

To find its maximum, we take its derivative and set it to 0

$$\begin{aligned}
\frac{\partial I(X;Y)}{\partial p} &= 0 \\
-\alpha \log_2(p\alpha) - \frac{\alpha}{\ln(2)} + \alpha \log_2(1 - p\alpha) + \frac{\alpha}{\ln(2)} - H(\alpha) &= 0 \\
-H(\alpha) &= \alpha \log_2(p\alpha) - \alpha \log_2(1 - p\alpha) \\
2^{\frac{-H(\alpha)}{\alpha}} &= \frac{p\alpha}{1 - p\alpha} \\
p &= \boxed{\frac{1}{\alpha(2^{\frac{H(\alpha)}{\alpha}} + 1)}}
\end{aligned}$$

Problem 2

(a)

We have that,

$$\begin{aligned}
I(X;Y|S) &= H(Y|S) - H(Y|X, S) \\
&= H(Y, S) - H(S) - H(Y, X, S) - H(S, X) \\
&= I(X, S; Y, S) - H(S)
\end{aligned}$$

And thus

$$I(X, S; Y, S) = I(X; Y|S) + H(S)$$

(b)

We have that

$$\begin{aligned}
C &= \max_{p_S(s)} \left(\max_{p_X(x|s)} (I(X, S; Y, S)) \right) \\
&= \max_{p_S(s)} \left(\max_{p_X(x|s)} (I(X; Y|S) + H(S)) \right) \\
&= \max_{p_S(s)} \left(H(S) + \max_{p_X(x|s)} (I(X; Y|S)) \right) \\
&= \max_{p_S(s)} \left(H(S) + \max_{p_X(x|s)} (H(Y|S) + H(X|S) - H(Y, X|S)) \right) \\
&= \max_{p_S(s)} \left(H(S) + \max_{p_X(x|s)} \left(\sum_{s=1}^K p_S(s) (H(Y|S=s) + H(X|S=s) - H(Y, X|S=s)) \right) \right) \\
&= \max_{p_S(s)} \left(H(S) + \sum_{s=1}^K p_S(s) C_s \right)
\end{aligned}$$

(c)

We have that we want the following constraint to be true

$$\sum_{s=1}^K p_S(s) = 1$$

So then with a lagrangian we have

$$\frac{\partial}{\partial p_S(s)} \left(H(S) + \sum_{s=1}^K p_S(s) C_s + \lambda \sum_{s=1}^K p_S(s) - 1 \right) = -\log_2(p_S(s)) - \frac{1}{\ln(2)} + C_s + \lambda$$

Setting this equal to 0 we get that

$$p_S * (s) = 2^{-\frac{1}{\log 2(e)}} 2^\lambda 2^{C_s}$$

since $\sum_{s=1}^K p_S * (s) = 1$, we have

$$2^\lambda = \frac{1}{2^{-\frac{1}{\log 2(e)}} \sum_{s=1}^K 2^{C_s}}$$

And thus we have

$$p_S(s) = \frac{2^{C_s}}{\sum_{s=1}^K 2^{C_s}}$$

(d)

Then we have that

$$\begin{aligned} H(S) &= - \sum_{s=1}^K p_S(s) \log_2(p_S(s)) \\ &= \log_2 \left(\sum_{s=1}^K 2^{C_s} \right) - \sum_{s=1}^K p_S(s) C_s \end{aligned}$$

Thus we have that

$$C = H(S) + \sum_{s=1}^K p_S(s) C_s = \log_2 \left(\sum_{s=1}^K 2^{C_s} \right)$$

Problem 3

(a)

We want the outputs to Y_1 and Y_2 to be totally dependent on the input X . Thus we have that

$$I(S, Y_1, Y_2) = H(S) - H(S|Y_1, Y_2) = H(S)$$

(b)

Since X_1 is a BEC channel, the maximum distribution is just $P(X_1 = 0) = P(X_1 = 1) = \frac{1}{2}$. This will give us that the capacity of this channel is $1 - \alpha$. Thus we have that the minimum capacity of the noiseless channel is α .

(c)

Send one symbol through the erasure channel and one through the noiseless channel, then if the fraction of the symbols through the noiseless is greater than α , we send the next symbol through the noisy channel, and if it is less than α , we send the next symbol through the noiseless channel. This will converge to α with time.

(d)

We have that

$$\begin{aligned} I(X_1, Y_1) &= H(X_1) - H(X_1|Y_1) \\ &= (1 - \alpha)H(X_1) \end{aligned}$$

We want to maximize this subject to the constraints on $P(X_1 = 1) = p$

$$p \geq 0$$

$$1 - p \leq 1$$

$$E[q(X_1)] = 3p + 2(1 - p) = 2 + p \leq Q$$

Then with the first two conditions will be inactive constraints since entropy gets bigger as p gets closer to $\frac{1}{2}$, but the third one will be active. Thus we have that

$$f = (1 - \alpha)H(p) + \lambda(2 + p - Q)$$

$$\frac{\partial f}{\partial p} = (1 - \alpha)H'(p) + \lambda = 0$$

$$\frac{\partial f}{\partial \lambda} = 2 + p - Q = 0$$

Thus we have that

$$\log_2 \left(\frac{p}{1 - p} \right) = \frac{\lambda}{1 - \alpha}$$

$$p = \frac{1}{1 + 2^{-\frac{\lambda}{1 - \alpha}}}$$

plugging this back into the third constraint we get that

$$\begin{aligned}
p &= Q - 2 \\
1 &= (Q - 2)(1 + 2^{-\frac{\lambda}{1-\alpha}}) \\
\frac{3 - Q}{Q - 2} &= 2^{-\frac{\lambda}{1-\alpha}} \\
\lambda &= (1 - \alpha) \log_2 \left(\frac{Q - 2}{3 - Q} \right)
\end{aligned}$$

Thus we get that

$$p = 3 - Q$$

and thus the maximum $I(X_1, Y_1)$ is $(1 - \alpha)H_2(3 - Q)$, and thus the maximum of $I(S; Y_1)$ is $\boxed{(1 - \alpha)H_2(3 - Q)}$ as well.

(e)

Then we must have the rate of the other channel must be $1 - (1 - \alpha)H_2(3 - Q)$

Problem 4

(a)

We can intuitively see that the stationary distribution is $\frac{1}{3}$ for all, thus we have that the rate is

$$\mathcal{H}(S) = H_3(q, q, 1 - 2q)$$

Thus in order for lossless coding, we must have that q must satisfy the following

$$\begin{aligned}
R &= 1 - \alpha > \mathcal{H}(S) \\
\boxed{1 - \alpha > H_2(2q, 1 - 2q) + 2q}
\end{aligned}$$

(b)

From the source coding theorem we have that

$$1 - \alpha > \beta \mathcal{H}(S)$$

Thus the maximum beta is

$$\beta = \frac{2(1 - \alpha)}{3}$$

Problem 5

(a)

We have that

$$I(X; Y) = H(X) - H(X|Y) = H(X) - pH(X|Y = E) - (1 - p)H(X|Y = X)$$

$$I(X; Y) = H(X) - pH(X|Y = E) = (1 - p)H(X)$$

thus to maximize we just have all the values of X be equally likely thus we will have

$$C = (1 - p) \log_2(L)$$

(b)

Since feedback cannot increase the capacity of the channel, we have that the capacity still is $(1 - p) \log_2(L)$

(c)

Once again since feedback cannot increase the capacity of the channel, we have that the capacity still is $(1 - p) \log_2(L)$