

# Chapter 4 : Solutions to the exercises

## Exercise 4.1 Solution

$$H(X) = 2.4087 \text{ bits/symbol.}$$

In the case where the symbols are equally probable

$$H(X) = \log_2(6) = 2.585 \text{ bits/symbol.}$$

## Exercise 4.2 Solution

a) Defining  $\alpha = p(1 - \varepsilon) + (1 - p)\varepsilon$

$$H(X) = H_b(p) = -p \log_2(p) - (1 - p) \log_2(1 - p).$$

$$H(Y) = H_b(\alpha) = -\alpha \log_2(\alpha) - (1 - \alpha) \log_2(1 - \alpha).$$

$$H(Y|X) = H_b(\varepsilon).$$

$$H(X, Y) = H_b(p) + H_b(\varepsilon).$$

$$H(X|Y) = H_b(p) + H_b(\varepsilon) - H_b(\alpha)$$

$$I(X, Y) = H_b(\alpha) - H_b(\varepsilon).$$

b)  $p = \frac{1}{2}$

c)  $\varepsilon = \frac{1}{2}$

## Exercise 4.3 Solution

a)  $H(X) = 2.5282$  bits/symbol. For equiprobable symbols

$$H(X) = \log_2(7) = 2.8074 \text{ bits/symbol.}$$

b)  $H(Y) = 1.4060$  bits/symbol. For equiprobable symbols

$$H(Y) = \log_2(3) = 1.585 \text{ bits/symbol}$$

c)

$$H(X, Y) = 2.5282 \text{ bits/symbol.}$$

$$H(X|Y) = 1.1222 \text{ bits/symbol.}$$

$$H(Y|X) = 0 \text{ bits/symbol.}$$

$$I(X, Y) = 1.4060 \text{ bits/symbol.}$$

**Exercise 4.4** Solution

The channel capacity is

$$C = (1 - \varepsilon) \text{ bits/usage,}$$

and that this capacity is achieved for some a priori probabilities

$$p_X(0) = p_X(1) = \frac{1}{2}.$$

The capacity is represented as a function of  $\varepsilon$  in Figure 4.1.

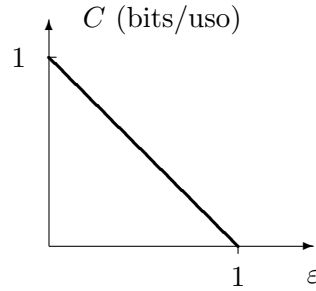


Figure 4.1: Channel capacity as a function of  $\varepsilon$ , Exercise 4.4.

**Exercise 4.5** Solution

a)  $C = \varepsilon$ , for  $p_X(x_0) = p_X(x_1) = \frac{1}{2}$ .

b)

$$H(X, Y) = H_b(p) + H_b(\varepsilon)$$

$$H(X|Y) = (1 - \varepsilon)H_b(p)$$

**Exercise 4.6** Solution

a) For  $\varepsilon = 0$

$$C = \log_2(3) = 1.585 \text{ bits/use.}$$

The input probabilities for which the channel capacity is reached are those that satisfy

$$p_X(x_1) = p_X(x_3) = \frac{1}{3}, \quad [p_X(x_0) + p_X(x_2)] = \frac{1}{3}.$$

For  $\varepsilon = 1$

$$C = \log_2(2) = 1 \text{ bit/use.}$$

The input probabilities for which the channel capacity is reached are those that satisfy

$$[p_X(x_0) + p_X(x_2)] = \frac{1}{2}, \quad [p_X(x_1) + p_X(x_3)] = \frac{1}{2}.$$

b)  $\varepsilon = 0.75$

$$H(X|Y) = 0.931 \text{ bits/symbol.}$$

c) The probabilities for which the channel capacity is obtained are

$$p_X(x_0) = 0.403, \quad p_X(x_1) = 0.339, \quad p_X(x_2) = 0.258.$$

The channel capacity is the maximum of the mutual information

$$C = H(Y) - H(Y|X) = 1.308 \text{ bits/use}$$

#### Exercise 4.7 Solution

a) The channel capacity is

$$C = \log_2(3) = 1.585 \text{ bits/use}$$

for

$$p_X(x_0) = p_X(x_1) = p_X(x_2) = \frac{1}{3}.$$

b) Any of these two values is possible

$$\varepsilon = \begin{cases} 0.25 \\ 0.75 \end{cases}.$$

$$H(Y) = \begin{cases} 1.5546, & \varepsilon = 0.25 \\ 1.2807, & \varepsilon = 0.75 \end{cases}.$$

$$I(X, Y) = \begin{cases} 1.0137, & \varepsilon = 0.25 \\ 0.7398, & \varepsilon = 0.75 \end{cases}.$$

$$H(X|Y) = H(X, Y) - H(Y) = \begin{cases} 0.5712, & \varepsilon = 0.25 \\ 0.8451, & \varepsilon = 0.75 \end{cases}.$$

#### Exercise 4.8 Solution

a) The channel capacity is  $C = 1$  bit/usage and is reached for

$$p_X(x_1) = 1/2, \quad p_X(x_0) + p_X(x_2) = 1/2.$$

b)  $H(X|Y) = 0$  and  $H(X, Y) = H_b(p) + p H_b(\varepsilon)$ . For  $\varepsilon = \frac{1}{2}$ ,  $H(X, Y) = H_b(p) + p$ .

#### Exercise 4.9 Solution

a)  $C = 1$  bit/use, for  $p_X(x_0) = \frac{1}{2}$  and  $p_X(x_1) + p_X(x_2) = \frac{1}{2}$ .

b)

$$H(X, Y) = 2p + H_b(2p) + (1 - 2p)H_b(\varepsilon)$$

$$H(X|Y) = 2p$$

#### Exercise 4.10 Solution

a)

$$\mathbf{P}^{DMC} = \begin{bmatrix} 1 - \varepsilon & \varepsilon & 0 & 0 \\ \varepsilon & 1 - 2\varepsilon & \varepsilon & 0 \\ 0 & \varepsilon & 1 - 2\varepsilon & \varepsilon \\ 0 & 0 & \varepsilon & 1 - \varepsilon \end{bmatrix}$$

$$\mathbf{P}^{Const} = \begin{bmatrix} 1 - Q\left(\frac{1}{\sqrt{N_0/2}}\right) & Q\left(\frac{1}{\sqrt{N_0/2}}\right) - Q\left(\frac{3}{\sqrt{N_0/2}}\right) & Q\left(\frac{3}{\sqrt{N_0/2}}\right) - Q\left(\frac{5}{\sqrt{N_0/2}}\right) & Q\left(\frac{5}{\sqrt{N_0/2}}\right) \\ Q\left(\frac{1}{\sqrt{N_0/2}}\right) & 1 - 2Q\left(\frac{1}{\sqrt{N_0/2}}\right) & Q\left(\frac{1}{\sqrt{N_0/2}}\right) - Q\left(\frac{3}{\sqrt{N_0/2}}\right) & Q\left(\frac{3}{\sqrt{N_0/2}}\right) \\ Q\left(\frac{3}{\sqrt{N_0/2}}\right) & Q\left(\frac{1}{\sqrt{N_0/2}}\right) - Q\left(\frac{3}{\sqrt{N_0/2}}\right) & 1 - 2Q\left(\frac{1}{\sqrt{N_0/2}}\right) & Q\left(\frac{1}{\sqrt{N_0/2}}\right) \\ Q\left(\frac{5}{\sqrt{N_0/2}}\right) & Q\left(\frac{3}{\sqrt{N_0/2}}\right) - Q\left(\frac{5}{\sqrt{N_0/2}}\right) & Q\left(\frac{1}{\sqrt{N_0/2}}\right) - Q\left(\frac{3}{\sqrt{N_0/2}}\right) & 1 - Q\left(\frac{1}{\sqrt{N_0/2}}\right) \end{bmatrix}$$

The approximation that has been made is that only the probability of error between symbols that are at a minimum distance is considered, neglecting the probability of making errors with symbols at  $2d_{min}$  or  $3d_{min}$ . That means

$$Q\left(\frac{k}{\sqrt{N_0/2}}\right) \approx 0, \text{ for } k > 1,$$

and

$$\varepsilon = Q\left(\frac{1}{\sqrt{N_0/2}}\right).$$

b)

$$H(Y|X) = \frac{1}{2}H_b(\varepsilon) + \frac{1}{2}H_b(2\varepsilon) + \varepsilon$$

$$H(X|Y) = \frac{1}{2}H_b(\varepsilon) + \frac{1}{2}H_b(2\varepsilon) + \varepsilon$$

$$H(X, Y) = 2 + \frac{1}{2}H_b(\varepsilon) + \frac{1}{2}H_b(2\varepsilon) + \varepsilon$$

$$I(X, Y) = 2 - \frac{1}{2}H_b(\varepsilon) - \frac{1}{2}H_b(2\varepsilon) - \varepsilon$$

The value of  $\varepsilon$  that minimizes  $H(Y|X)$  and  $H(X|Y)$  and maximizes  $I(X, Y)$  is  $\varepsilon = 0$  (in which case  $H(X|Y) = H(Y|X) = 0$ , and  $I(X, Y) = 2$ ).

c) The minimum and maximum values that  $\varepsilon$  can take are  $\varepsilon = 0$  and  $\varepsilon = \frac{1}{2}$ , respectively.

i) For  $\varepsilon = 0$ ,  $C = 2$  bits/use for  $p_X(x_i) = \frac{1}{4}$ ,  $i = 0, 1, 2, 3$ .

ii) For  $\varepsilon = \frac{1}{2}$ ,  $C = 1$  bits/use for  $p_X(x_0) = p_X(x_3)$  and  $p_X(x_1) = p_X(x_2)$ .

### Exercise 4.11 Solution

a) The channel matrix for the DMC is

$$\mathbf{P}^{DMC} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 - \varepsilon & \varepsilon & 0 \\ 0 & \varepsilon & 1 - \varepsilon & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$

And for the constellation

$$\mathbf{P}^{Const.} = \begin{bmatrix} 1 - Q\left(\frac{2}{\sqrt{N_0/2}}\right) & Q\left(\frac{2}{\sqrt{N_0/2}}\right) - Q\left(\frac{5}{\sqrt{N_0/2}}\right) & Q\left(\frac{5}{\sqrt{N_0/2}}\right) - Q\left(\frac{8}{\sqrt{N_0/2}}\right) & Q\left(\frac{8}{\sqrt{N_0/2}}\right) \\ Q\left(\frac{2}{\sqrt{N_0/2}}\right) & 1 - Q\left(\frac{1}{\sqrt{N_0/2}}\right) - Q\left(\frac{2}{\sqrt{N_0/2}}\right) & Q\left(\frac{1}{\sqrt{N_0/2}}\right) - Q\left(\frac{4}{\sqrt{N_0/2}}\right) & Q\left(\frac{4}{\sqrt{N_0/2}}\right) \\ Q\left(\frac{4}{\sqrt{N_0/2}}\right) & Q\left(\frac{1}{\sqrt{N_0/2}}\right) - Q\left(\frac{4}{\sqrt{N_0/2}}\right) & 1 - Q\left(\frac{1}{\sqrt{N_0/2}}\right) - Q\left(\frac{2}{\sqrt{N_0/2}}\right) & Q\left(\frac{2}{\sqrt{N_0/2}}\right) \\ Q\left(\frac{8}{\sqrt{N_0/2}}\right) & Q\left(\frac{5}{\sqrt{N_0/2}}\right) - Q\left(\frac{8}{\sqrt{N_0/2}}\right) & Q\left(\frac{2}{\sqrt{N_0/2}}\right) - Q\left(\frac{5}{\sqrt{N_0/2}}\right) & 1 - Q\left(\frac{2}{\sqrt{N_0/2}}\right) \end{bmatrix}.$$

The approximation that has been made is that the probability of transition with symbols further away from distance 2 is zero, or what is the same, that errors only occur between the symbols  $\mathbf{a}_1$  and  $\mathbf{a}_2$ , and with this assumption

$$Q\left(\frac{k}{\sqrt{N_0/2}}\right) \approx 0, \text{ for } k > 1.$$

Therefore

$$\varepsilon = Q\left(\frac{1}{\sqrt{N_0/2}}\right).$$

b)

$$H(X, Y) = 2 + \frac{1}{2} H_b(\varepsilon)$$

$$H(Y|X) = \frac{1}{2} H_b(\varepsilon).$$

$$H(X|Y) = \frac{1}{2} H_b(\varepsilon).$$

$$I(X, Y) = 2 - \frac{1}{2} H_b(\varepsilon).$$

The representation of these functions is shown in Figure 4.2.

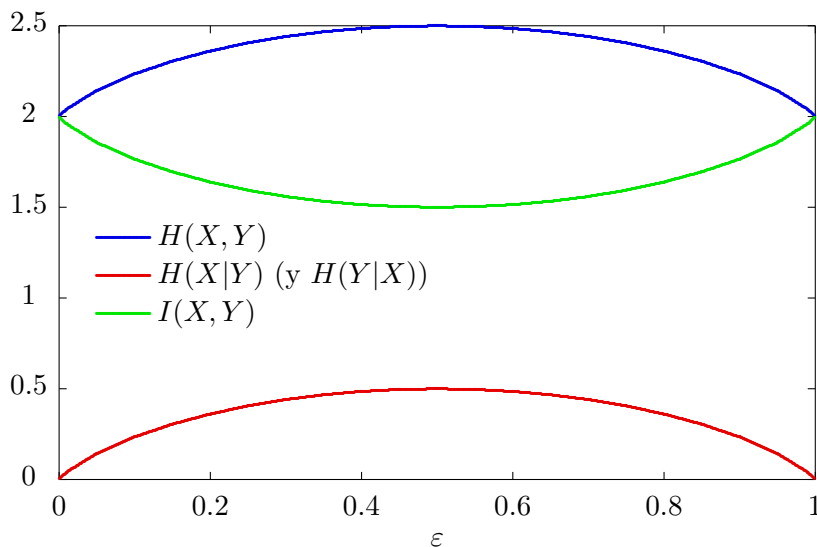


Figure 4.2: Plot of  $H(X, Y)$ ,  $H(X|Y)$  (equal to  $H(Y|X)$ ) and  $I(X, Y)$  as a function of  $\varepsilon$ .

The maximum values of  $H(X, Y)$ ,  $H(X|Y)$ , and  $H(Y|X)$  are obtained for  $\varepsilon = 1/2$ , the same as the minimum value of  $I(X, Y)$ .

c) The channel capacity is

$$C = 1 + H_b \left( \frac{2^{H_b(\varepsilon)}}{1 + 2^{H_b(\varepsilon)}} \right) - \left( \frac{1}{1 + 2^{H_b(\varepsilon)}} \right) H_b(\varepsilon),$$

value obtained for

$$p_X(x_0) = p_X(x_3) = \frac{2^{H_b(\varepsilon)-1}}{1 + 2^{H_b(\varepsilon)}}, \quad p_X(x_1) = p_X(x_2) = \frac{1}{2(1 + 2^{H_b(\varepsilon)})}.$$

**Exercise 4.12** Solution

1.  $a = d = 0, b = c = \varepsilon_1$ .

$$H(Y) = 2 + \frac{1}{2} \left[ (1 - \varepsilon_0) \log_2 \frac{1}{1 - \varepsilon_0} + (1 + \varepsilon_0) \log_2 \frac{1}{1 + \varepsilon_0} \right].$$

If we take into account that  $\frac{1}{2}(1 + \varepsilon_0) = 1 - \frac{1}{2}(1 - \varepsilon_1)$ , we could regroup terms and write

$$H(Y) = 1 + H_b \left( \frac{1}{2}(1 - \varepsilon_0) \right).$$

$$H(X, Y) = 2 + \frac{1}{2} H_b(\varepsilon_0) + \frac{1}{2} H_b(\varepsilon_1)$$

$$H(Y|X) = \frac{1}{2} H_b(\varepsilon_0) + \frac{1}{2} H_b(\varepsilon_1)$$

$$H(X|Y) = 1 + \frac{1}{2} H_b(\varepsilon_0) + \frac{1}{2} H_b(\varepsilon_1) - H_b \left( \frac{1}{2}(1 - \varepsilon_0) \right)$$

$$I(X, Y) = 1 + H_b \left( \frac{1}{2}(1 - \varepsilon_0) \right) - \frac{1}{2} H_b(\varepsilon_0) - \frac{1}{2} H_b(\varepsilon_1)$$

2. The channel capacity is

$$C = 1 + H_b(p(1 - \varepsilon_0)) - pH_b(\varepsilon_0) - (1 - p)H_b(\varepsilon_1) \text{ bits/usage}$$

for

$$p = \frac{1}{(1 - \varepsilon_0) \left[ 1 + 2^{\frac{H_b(\varepsilon_0) - H_b(\varepsilon_1)}{1 - \varepsilon_0}} \right]}.$$

The distribution that allow to achieve the capacity is

$$p_X(x_0) = p_X(x_3) = \frac{p}{2}, \quad p_X(x_1) = p_X(x_2) = \frac{1 - p}{2}$$

**Exercise 4.13** Solution

a)

$$H(Y) = \frac{2}{3}(1 - \varepsilon_0 + \varepsilon_1) + H_b \left( \frac{2}{3}(1 - \varepsilon_0 + \varepsilon_1) \right)$$

$$H(Y|X) = \frac{2}{3} H_b(\varepsilon_0) + \frac{1}{3} H_b(2\varepsilon_1) + \frac{2}{3} \varepsilon_1.$$

$$H(X, Y) = \log_2(3) + \frac{2}{3} H_b(\varepsilon_0) + \frac{1}{3} H_b(2\varepsilon_1) + \frac{2}{3} \varepsilon_1.$$

$$H(X|Y) = H(X, Y) - H(Y)$$

$$I(X, Y) = H(X) - H(X|Y) = H(Y) - H(Y|X)$$

b) In this case the channel is the *binary erasure channel* (BEC)

$$C = (1 - \varepsilon_0) \text{ bits/symbol, for } p_X(x_0) = p_X(x_2) = \frac{1}{2}.$$

c)

$$C = p(1 - \varepsilon_0) + (1 - p)2\varepsilon_1 + H_b(p(1 - \varepsilon_0) + (1 - p)2\varepsilon_1) - pH_b(\varepsilon_0) - (1 - p)(H_b(2\varepsilon_1) + 2\varepsilon_1)$$

for the value

$$p = \frac{(1 - 2\varepsilon_1) 2^{\frac{H_b(\varepsilon_0) - H_b(2\varepsilon_1) - 2\varepsilon_1}{2\varepsilon_1 + \varepsilon_0 - 1}} - \varepsilon_1}{(1 - 2\varepsilon_1 - \varepsilon_0) 2^{\frac{H_b(\varepsilon_0) - H_b(2\varepsilon_1) - 2\varepsilon_1}{2\varepsilon_1 + \varepsilon_0 - 1}} + \frac{1}{2}(1 - \varepsilon_0 - 2\varepsilon_1)}$$

being the input distribution

$$p_X(x_0) = p_X(x_2) = \frac{p}{2}, p_X(x_1) = 1 - p$$

d) If we have that  $\varepsilon_0 = \varepsilon_1$ , it is very easy to see that for  $\varepsilon_0 = \varepsilon_1 = 0$  we have an ideal channel, in which  $H(Y|X) = 0$ , and capacity is achieved with equiprobable input symbols, which generate equiprobable output symbols and therefore the channel capacity is  $\log_2(3)$ , which is the maximum possible value for a system with three inputs and three outputs.

**Exercise 4.14 Solution**

a)  $a = b = c = d = 0$ . They are two BSC channels in parallel with different error probabilities. Therefore it would be a good model for a system that is transmitting bits through two independent subsystems.

b)

$$H(X) = 1 + H_b(p).$$

$$H(Y) = 1 + H_b(p)$$

$$H(X, Y) = 1 + H_b(p) + pH_b(\varepsilon_1) + (1 - p)H_b(\varepsilon_0)$$

$$H(Y|X) = pH_b(\varepsilon_1) + (1 - p)H_b(\varepsilon_0)$$

$$H(X|Y) = pH_b(\varepsilon_1) + (1 - p)H_b(\varepsilon_0)$$

c)

$$C = 1 + H_b(p) - pH_b(\varepsilon_1) - (1 - p)H_b(\varepsilon_0)$$

with

$$p = \frac{1}{1 + 2^{H_b(\varepsilon_1) - H_b(\varepsilon_0)}}$$

being the input distribution

$$p_X(x_0) = p_X(x_1) = \frac{p}{2}, p_X(x_2) = p_X(x_3) = \frac{1 - p}{2}$$

**Exercise 4.15 Solution**

a)

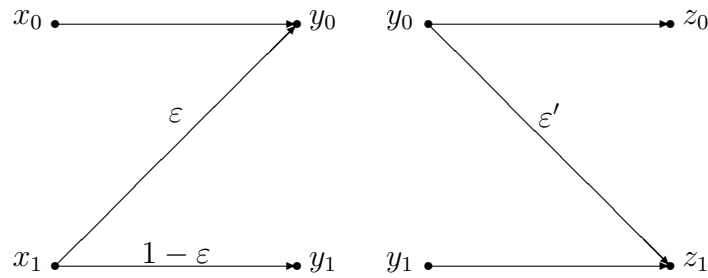
$$H(X) = H_b(\alpha), H(Y) = H_b((1 - \alpha)(1 - \varepsilon)), H(Z) = H_b(\beta(1 - \varepsilon'))$$

$$H(X, Y) = H_b(\alpha) + (1 - \alpha)H_b(\varepsilon)$$

$$H(Y|Z) = H_b(\beta) + \beta H_b(\varepsilon') - H_b(\beta(1 - \varepsilon'))$$

b) The transition probabilities, given in the channel matrix, are

$$\mathbf{P}_{Y|X} = \begin{bmatrix} 1 & 0 \\ \varepsilon & 1 - \varepsilon \end{bmatrix}, \quad \mathbf{P}_{Z|Y} = \begin{bmatrix} 1 - \varepsilon' & \varepsilon' \\ 0 & 1 \end{bmatrix}$$



c)  $I(X, Y) = H_b((1 - \alpha)(1 - \varepsilon)) - (1 - \alpha)H_b(\varepsilon)$

d)

$$\mathbf{P}_{Z|X} = \mathbf{P}_{Y|X} \times \mathbf{P}_{Z|Y} = \begin{bmatrix} 1 - \varepsilon' & \varepsilon' \\ \varepsilon(1 - \varepsilon') & \varepsilon\varepsilon' + 1 - \varepsilon \end{bmatrix}$$

e) The concatenation will be a BSC if  $\varepsilon = \frac{\varepsilon'}{1 - \varepsilon'}$

**Exercise 4.16** Solution

a) The channel matrix is

$$\mathbf{P} = \begin{bmatrix} 1 & 0 \\ \varepsilon & 1 - \varepsilon \\ 0 & 1 \end{bmatrix}.$$

b) Entropy values in the two scenarios

i)

$$H(X) = \log_2 3 = 1.585 \text{ bits/symbol.}$$

$$H(Y) = H_b\left(\frac{1}{3}(1 + \varepsilon)\right) = H_b\left(\frac{1}{3}(2 - \varepsilon)\right) \text{ bits/symbol.}$$

$$H(Y|X) = p_1 H_b(\varepsilon) = \frac{1}{3} H_b(\varepsilon) \text{ bits/symbol.}$$

$$H(X, Y) = \log_2 3 + \frac{1}{3} H_b(\varepsilon) \text{ bits/symbol.}$$

$$H(X|Y) = \log_2 3 + \frac{1}{3} H_b(\varepsilon) - H_b\left(\frac{1}{3}(1 + \varepsilon)\right) \text{ bits/symbol.}$$

$$I(X, Y) = H_b\left(\frac{1}{3}(1 + \varepsilon)\right) - \frac{1}{3} H_b(\varepsilon) \text{ bits/symbol.}$$

ii)

$$H(X) = 0 \text{ bits/symbol,}$$

$$H(Y) = H_b(\varepsilon) \text{ bits/symbol.}$$

$$H(Y|X) = H_b(\varepsilon) \text{ bits/symbol.}$$

$$H(X, Y) = H_b(\varepsilon) \text{ bits/symbol.}$$

$$H(X|Y) = 0 \text{ bits/symbol,}$$

$$I(X, Y) = 0 \text{ bits/symbol.}$$



c) The channel capacity, for each case, is

i)  $C = \max_{p_X(x_i)} I(X, Y) = 1$  bit/use, value that is reached for  $p = 0$ , which means that

$$p_0 = \frac{1}{2}, p_1 = 0, p_2 = \frac{1}{2}.$$

ii)

$$C = H_b \left( \frac{1}{1 + 2^{\frac{H_b(\varepsilon)}{1-\varepsilon}}} \right) - \frac{1}{(1-\varepsilon)} \frac{1}{\left(1 + 2^{\frac{H_b(\varepsilon)}{1-\varepsilon}}\right)} H_b(\varepsilon),$$

what is obtained for

$$p_X(x_0) = 1 - p = 1 - \frac{1}{(1-\varepsilon) \left(1 + 2^{\frac{H_b(\varepsilon)}{1-\varepsilon}}\right)}, p_X(x_1) = p = \frac{1}{(1-\varepsilon) \left(1 + 2^{\frac{H_b(\varepsilon)}{1-\varepsilon}}\right)}.$$

### Exercise 4.17 Solution

a) The joint entropy is

$$\begin{aligned} H(X, Y) &= - \sum_{i=0}^1 \sum_{j=0}^1 p_{X,Y}(x_i, y_j) \log_2 p_{X,Y}(x_i, y_j) \\ &= -0.1 \log_2 0.1 - 0.2 \log_2 0.2 - 0.3 \log_2 0.3 - 0.4 \log_2 0.4 = 1.8464 \text{ bits/symbol} \end{aligned}$$

b) The probabilities of the random variable  $X$  are

$$\begin{cases} p_X(x_0) = p_{X,Y}(x_0, y_0) + p_{X,Y}(x_0, y_1) = 0.1 + 0.2 = 0.3 \\ p_X(x_1) = p_{X,Y}(x_1, y_0) + p_{X,Y}(x_1, y_1) = 0.3 + 0.4 = 0.7 \end{cases}$$

so the entropy is

$$H(X) = -0.3 \log_2 0.3 - 0.7 \log_2 0.7 = H_b(0.3) = H_b(0.7) = 0.8813 \text{ bits/symbol}$$

The probabilities of the random variable  $Y$  are

$$\begin{cases} p_Y(y_0) = p_{X,Y}(x_0, y_0) + p_{X,Y}(x_1, y_0) = 0.1 + 0.3 = 0.4 \\ p_Y(y_1) = p_{X,Y}(x_0, y_1) + p_{X,Y}(x_1, y_1) = 0.2 + 0.4 = 0.6 \end{cases}$$

so the entropy is

$$H(Y) = -0.4 \log_2 0.4 - 0.6 \log_2 0.6 = H_b(0.4) = H_b(0.6) = 0.971 \text{ bits/symbol}$$

The conditional entropy  $H(X|Y)$  can be calculated in several ways. One is by taking advantage of the relationship

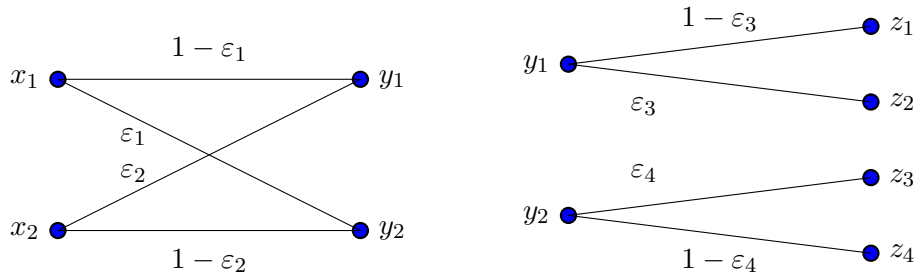
$$H(X|Y) = H(X, Y) - H(Y) = 1.8464 - 0.971 = 0.8754 \text{ bits/symbol}$$

c) Mutual information can also be calculated in several ways. One of them is

$$I(X, Y) = H(X) - H(X|Y) = 0.8813 - 0.8754 = 0.0059 \text{ bits/symbol}$$

### Exercise 4.18 Solution

a) The schematic representation of the two DMCs is shown in the figure



b) The entropy of this variable, being binary, is

$$H(X) = \Omega(\alpha) \text{ bits/symbol.}$$

The probabilities of the exit symbols are

$$\begin{aligned} P_Y(y_1) &= \alpha(1 - \varepsilon_1) + (1 - \alpha)\varepsilon_2 \equiv \gamma \\ P_Y(y_2) &= \alpha\varepsilon_1 + (1 - \alpha)(1 - \varepsilon_2) \equiv 1 - \gamma \end{aligned}$$

so since it is also a binary variable it has entropy

$$H(Y) = \Omega(\gamma) \text{ bits/symbol}$$

For the joint entropy it is convenient to calculate the conditional

$$H(Y|X) = \alpha\Omega(\varepsilon_1) + (1 - \alpha)\Omega(\varepsilon_2) \text{ bits/symbol, since } \begin{cases} H(Y|X = x_1) = \Omega(\varepsilon_1) \\ H(Y|X = x_2) = \Omega(\varepsilon_2) \end{cases}$$

Now the joint entropy is

$$H(X, Y) = H(X) + H(Y|X) = \Omega(\gamma) + \alpha\Omega(\varepsilon_1) + (1 - \alpha)\Omega(\varepsilon_2) \text{ bits/symbol}$$

c) The probabilities of  $Z$  are now

$$\begin{aligned} P_Y(z_1) &= \beta(1 - \varepsilon_3) \\ P_Y(z_2) &= \beta\varepsilon_3 \\ P_Y(z_3) &= (1 - \beta)\varepsilon_4 \\ P_Y(z_4) &= (1 - \beta)(1 - \varepsilon_4) \end{aligned}$$

so the entropy of  $Z$  is

$$\begin{aligned} H(Z) &= -\beta(1 - \varepsilon_3) \log_2 \beta(1 - \varepsilon_3) - \beta\varepsilon_3 \log_2 \beta\varepsilon_3 \\ &\quad - (1 - \beta)\varepsilon_4 \log_2 (1 - \beta)\varepsilon_4 - (1 - \beta)(1 - \varepsilon_4) \log_2 (1 - \beta)(1 - \varepsilon_4) \\ &= -\beta [\beta \log_2 \beta + (1 - \beta) \log_2 (1 - \beta)] - \beta [\varepsilon_3 \log_2 \varepsilon_3 + (1 - \varepsilon_3) \log_2 (1 - \varepsilon_3)] \\ &\quad - (1 - \beta) [\varepsilon_4 \log_2 \varepsilon_4 + (1 - \varepsilon_4) \log_2 (1 - \varepsilon_4)] \\ &= \Omega(\beta) + \beta\Omega(\varepsilon_3) + (1 - \beta)\Omega(\varepsilon_4) \text{ bits/symbol} \end{aligned}$$

As for  $H(Y|Z)$ , knowing the value of  $Z$  there is no uncertainty about the value of  $Y$ , so

$$H(Y|Z) = 0 \text{ bits/symbol}$$

d) For DMC A

$$I(X, Y) = H(Y) - H(Y|X) = \Omega(\gamma) - \alpha\Omega(\varepsilon_1) - (1 - \alpha)\Omega(\varepsilon_2)$$

For DMC B

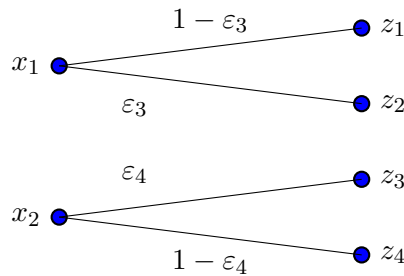
$$I(Y, Z) = H(Y) - H(Y|Z) = \Omega(\beta).$$

It could also be calculated as

$$I(Y, Z) = H(Z) - H(Z|Y),$$

taking into account that for this channel it is very easy to obtain  $H(Z|Y) = \beta\Omega(\varepsilon_3) + (1 - \beta)\Omega(\varepsilon_4)$ .

e) The concatenated channel, for  $\varepsilon_1 = \varepsilon_2 = 0$ , with its transition probabilities, is shown in the figure



f) The mutual information between input and output of this concatenated channel is

$$I(X, Z) = H(X) - H(X|Z) = \Omega(\alpha),$$

because for this channel  $H(X|Z) = 0$ . Since the channel capacity is the maximum value that this mutual information can take depending on the input probabilities, in this case

$$C = 1 \text{ bit/use, for } p_X(x_1) = p_X(x_2) = \frac{1}{2}$$

since the maximum value of  $\Omega(\alpha)$  is one and is obtained for  $\alpha = \frac{1}{2}$ .