

Solution of exercise sheet n° 5 :

**5-1.**

Channel capacity :  $C = \max_{p(x)} I(X : Y)$ . There are three ways to calculate  $I(X : Y)$  :

1.  $I(X : Y) = \sum_{x,y} p(x, y) \log_2 \frac{p(x,y)}{p(x)p(y)}$ .
2.  $I(X : Y) = H(X) - H(X|Y)$ .
3.  $I(X : Y) = H(Y) - H(Y|X)$ .

We remark that for the (memoryless) additive noise channel where the input  $X$  and the noise  $Z$  are uncorrelated we can use the relation  $H(Y|X) = H(X + Z|X) = H(Z)$ . Therefore, for the calculation of the capacity we can use in this exercise the equation

$$C = \max_{p(x)} \{H(Y)\} - H(Z), \tag{1}$$

where the second term  $H(Z)$  no longer depends on  $X$  (and thus,  $p(x)$ ).

In the following we need to distinguish three cases : I)  $a = 0$ , II)  $a > 1$  and III)  $a = 1$ .

- I)  $a = 0$  : No noise is added, thus  $Y = X$  and  $H(Z) = 0$ . The capacity is therefore  $C = \max_{p(x)} H(p, 1 - p) = 1$  bit.
- II)  $a > 1$  : The output alphabet reads  $\mathcal{Y} = \{0, 1, a, 1 + a\}$ . For the input variable  $X$  we define the general probability distribution  $P(X = 0) = p, P(X = 1) = 1 - p$ . Then we can compute probability distribution of the output  $Y$ , *i.e.*  $P(Y = 0) = P(X = 0)P(Z = 0) = p/2, P(Y = 1) = P(X = 1)P(Z = 0) = (1 - p)/2, P(Y = a) = P(X = 0)P(Z = a) = p/2$  and  $P(Y = a + 1) = P(X = 1)P(Z = a) = (1 - p)/2$ . We conclude that each output can be associated to a unique combination of input  $X$  and noise  $Z$  and thus, we make no error. We confirm that indeed like in I) again  $C = 1$  bit by injecting the probability distribution  $p(y)$  in equation (1) :

$$C = \max_p \left\{ -p \log_2(p/2) - (1 - p) \log_2\left(\frac{1 - p}{2}\right) \right\} - 1 \text{ bit.}$$

This is maximized by  $p = 1/2$  for which  $C = 1$  bit.

- III)  $a = 1$  : In this case,  $\mathcal{Y} = \{0, 1, 2\}$ . We see like above that if one obtains  $Y = 0, 2$  one does not do any error when estimating  $X$ . However,  $Y = 1$  corresponds to either  $X = 0, Z = 1$  or  $X = 1, Z = 0$ . We find the output probability distribution  $p(y) = \{p/2, 1/2, (1 - p)/2\}$ . Injected in the capacity formula (1) we find that  $C = 1/2$  bits for  $p = 1/2$ .

**5-2.** This exercise can be solved in the same way as in ex. **5-1**, *i.e.* because the input  $X$  and noise  $Z$  are independent we can use again Eq. (1) (careful : in general (1) is **is not valid!**).

(a) We again parametrize the input probability distribution, but now, as the input takes 4 values we set it to  $p(x) = \{a, b, c, d\}$  where  $a + b + c + d = 1$  (alternatively one can include the constraint into the parametrization, so write  $p(x) = \{a, b, c, 1 - a - b - c\}$ ). We remark, that  $-1 \bmod 4 = 3$ , so the output alphabet reads  $\mathcal{Y} = \{0, 1, 2, 3\}$ . Now, we have to find the parameters  $a, b, c, d$  that maximize the output entropy  $H(Y)$ . We can express (similar to ex. **5-1**)  $p(y)$  as

a function of  $a, b, c, d$ , which reads  $p(y) = \{\frac{a}{4} + \frac{c}{4} + \frac{d}{2}, \frac{b}{4} + \frac{c}{2} + \frac{d}{4}, \frac{a}{4} + \frac{b}{2} + \frac{c}{4}, \frac{a}{2} + \frac{b}{4} + \frac{d}{4}\}$ . We try now to find  $a, b, c, d$  such that the (optimal) uniform distribution  $p(y) = \{1/4, 1/4, 1/4, 1/4\}$  is reached (careful : in general it may **not be possible** to achieve this! Then one needs to apply the method of Lagrange multipliers). We have thus, 4 equations + 1 equation for the constraint  $a + b + c + d = 1$  to solve (one equation is linear dependent on the others). We find that  $a = c$  and  $b = d$ . Namely, there is an infinite number of solutions and among them  $a = b = c = d = 1/4$ , *i.e.*, the uniform distribution for  $X$  is a solution.

(b) When we inject the solution of (a) in Eq. (1) we find  $C = 1/2$  bits.

(c) We need to sum both noises :  $Z_{total} = Z_1 + Z_2$  and again follow a calculation like in (a). One obtains the new probability distribution (attention :  $-2 = 2$  since we apply “mod 4”);  $Z_{total}$  takes the values : -1, 0, 1, 2 with probabilities 1/4, 3/8, 1/4, 1/8. We have thus  $H(Z) = H(1/4, 3/8, 1/4, 1/8)$  and we find  $C = \log_2 4 - 1.91 = 0.09$  bits.

(d)  $C_{total} = C_1 + C_2 = 1$  bit. (The capacity is additive!)

### 5-3.

(a) If  $p(X = 1) = p$  and  $p(X = 0) = 1 - p$ . One obtains :

$$I(X : Y) = H(1 - p/2, p/2) - p$$

Taking into account that :

$$\frac{\partial H(x, 1 - x)}{\partial x} = \log_2 \frac{1 - x}{x} \quad (2)$$

The maximum is found for  $p^* = 2/5$  and  $C = I_{p=2/5}(X, Y) = \log_2 5 - 2$  bits.

(b) The binary symmetric channel has the capacity  $C = 1 - H(\alpha)$  bits, where  $\alpha$  is the error rate of the channel, because  $I(X : Y) = H(Y) - \sum p(x)H(Y|X = x) = H(Y) - H(\alpha) \leq 1 - H(\alpha)$  bits.

(c) We have  $C = \max_{p(x)} I(X : Y)$ . OÙ  $I(X : Y) = H(Y) - H(Y|X)$ . The entropy at the output is given by  $H(Y) = H(1 - pq, pq)$  and the conditional entropy reads  $H(Y|X) = pH(q, 1 - q)$ .  $I(X : Y)$  is maximal if  $\frac{\partial I(X:Y)}{\partial p} = 0$ , which implies

$$q \log_2 \frac{1 - pq}{pq} = H(q, 1 - q). \quad (3)$$

To simplify notations we write  $H(q, 1 - q) = H$ . The distribution  $p$  that maximizes  $I(X : Y)$  is

$$p = \frac{1}{q(1 + 2^{H/q})}. \quad (4)$$

We finally obtain the capacity of the channel :

$$C = \log_2(1 + 2^{H/q}) - \frac{H}{q}. \quad (5)$$

To check consistency we can test Eq. (5) for  $q = 0.5$ . Since  $H(q = 0.5) = 1$  we confirm the result of (a), *i.e.*  $C(q = 0.5) = \log_2 5 - 2$  bits.

### 5-4.

(a) It does not attain the capacity because the equiprobable distribution does not maximize the mutual information of the channel of exercises 5-3.

(b) For a probability distribution  $p(x)$ , the maximal transmission rate  $R$  is bounded from above by  $I(X : Y)$ . For the channel of exercise 5-3 :  $R_{p=1/2} < I_{p=1/2}(X : Y) = 0.3113$  bits.

**5-5.**

(a)  $C = \max_{p(x)} I(X : Y)$  and  $\tilde{C} = \max_{p(x)} I(X : \tilde{Y})$ .

We have  $I(X : Y, \tilde{Y}) = H(X : Y) + H(X : \tilde{Y}|Y)$ ,

and  $I(X : Y, \tilde{Y}) = H(X : \tilde{Y}) + H(X : Y|\tilde{Y})$ .

Since  $H(X : Y|\tilde{Y}) \geq 0$  and  $H(X : \tilde{Y}|Y) = 0$  (see exercise 2-3), we deduce that  $I(X : \tilde{Y}) \leq I(X : Y)$ . Thus,  $\tilde{C} > C$  is impossible.

(b) One requires  $H(X : Y|\tilde{Y}) = 0$ . This chain thus must satisfy  $X \rightarrow \tilde{Y} \rightarrow Y$ . This is only possible if  $Y \leftrightarrow \tilde{Y}$ , *i.e.* iff  $\tilde{Y} = f(Y)$  is a bijective function.