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## Exercise 4

### - Proposed Solution -

Friday, May 13, 2016

#### Solution of Problem 1

Theorem 4.3 shall be proven.

a)  $X$  is a discrete random variable with  $p_i = P(X = x_i)$ ,  $i = 1, \dots, m$ . It holds

$$H(X) = - \sum_i p_i \log(p_i) \geq 0,$$

as  $p_i \geq 0$  and  $-\log(p_i) \geq 0$  for  $0 < p_i \leq 1$  and  $0 \cdot \log 0 = 0$  per definition.

Equality holds, if all addends are zero, i.e.,

$$p_i \log(p_i) = 0 \Leftrightarrow p_i \in \{0, 1\} \quad i = 1, \dots, m,$$

as  $p_i > 0$  and  $-\log(p_i) > 0$ , thus,  $-p_i \log(p_i) > 0$  for  $0 < p_i < 1$ .

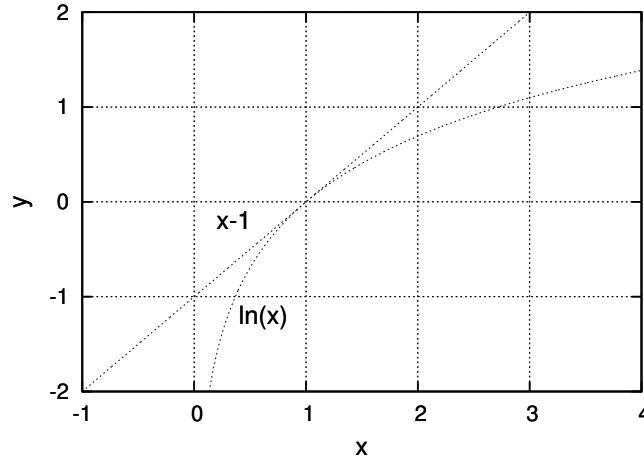
b) It holds

$$\begin{aligned} H(X) - \log(m) &= - \sum_i p_i \log(p_i) - \underbrace{\sum_i p_i}_{=1} \log(m) \\ &= \sum_{i:p_i>0} p_i \log\left(\frac{1}{p_i m}\right) \\ &= (\log e) \sum_{i:p_i>0} p_i \ln\left(\frac{1}{p_i m}\right) \\ &\stackrel{\ln(x) \leq x-1}{\leq} (\log e) \sum_{i:p_i>0} p_i \left(\frac{1}{p_i m} - 1\right) \\ &= (\log e) \sum_{i:p_i>0} \left(\frac{1}{m} - p_i\right) = 0 \end{aligned}$$

As  $\ln(x) = x - 1$  only holds for  $x = 1$  it follows that equality holds iff  $p_i = 1/m$ ,  $i = 1, \dots, m$ . In particular, as  $p_i = \frac{1}{m}$ , it follows  $p_i > 0$ ,  $i = 1, \dots, m$ .

c) Define for  $i = 1, \dots, m$  and  $j = 1, \dots, d$

$$p_{i|j} = P(X = x_i \mid Y = y_j).$$



Show  $H(X | Y) - H(X) \leq 0$  which is equivalent to the claim.

$$\begin{aligned}
 H(X | Y) - H(X) &= - \sum_{i,j} p_{i,j} \log(p_{i|j}) + \sum_i p_i \log(p_i) \\
 &= - \sum_{i,j} p_{i,j} \log\left(\frac{p_{i,j}}{p_j}\right) + \sum_i \underbrace{\sum_j p_{i,j}}_{=p_i} \log(p_i) \\
 &= (\log e) \sum_{i,j:p_{i,j}>0} p_{i,j} \ln\left(\frac{p_i p_j}{p_{i,j}}\right) \\
 &\stackrel{\ln(x) \leq x-1}{\leq} (\log e) \sum_{i,j:p_{i,j}>0} p_{i,j} \left(\frac{p_i p_j}{p_{i,j}} - 1\right) \\
 &= (\log e) \sum_{i,j:p_{i,j}>0} (p_i p_j - p_{i,j}) = 0
 \end{aligned}$$

Note that from  $p_{i,j} > 0$  it follows  $p_i, p_j > 0$ . Equality holds for  $p_i p_j = p_{i,j}$  which is equivalent to  $X$  and  $Y$  being stochastically independent.

This means that the mutual information  $I(X, Y) = H(X) - H(X | Y)$  is nonnegative.

d) It holds

$$\begin{aligned}
 H(X, Y) &= - \sum_{i,j} p_{i,j} \log(p_{i,j}) \\
 &= - \sum_{i,j} p_{i,j} [\log(p_{i,j}) - \log(p_i) + \log(p_i)] \\
 &= - \sum_{i,j} p_{i,j} \log\left(\frac{p_{i,j}}{p_i}\right) - \sum_i \underbrace{\sum_j p_{i,j}}_{=p_i} \log(p_i) \\
 &= H(Y | X) + H(X).
 \end{aligned}$$

e) It holds

$$H(X, Y) \stackrel{(d)}{=} H(X) + H(Y | X) \stackrel{(c)}{\leq} H(X) + H(Y)$$

with equality as in (c) iff  $X$  and  $Y$  are stochastically independent.

## Solution of Problem 2

Recall:  $H(X) = -\sum_i p_i \log(p_i)$ .

$$\begin{aligned} \text{a) } H(\hat{M}) &= -\frac{1}{4} \log_2\left(\frac{1}{4}\right) - \frac{3}{4} \log_2\left(\frac{3}{4}\right) = \frac{1}{2} + \frac{3}{2} - \frac{3}{4} \log_2(3) \approx 0.811 \\ H(\hat{K}) &= -\frac{1}{2} \log_2\left(\frac{1}{2}\right) - 2 \frac{1}{4} \log_2\left(\frac{1}{4}\right) = \frac{1}{2} + 1 = 1.5 \end{aligned}$$

c	$K_1$	$K_2$	$K_3$	
a	1	2	3	$\frac{1}{4}$
b	2	3	4	$\frac{3}{4}$
	$\frac{1}{2}$	$\frac{1}{4}$	$\frac{1}{4}$	1

$$\begin{aligned} P(\hat{C} = 1) &= P(\hat{M} = a) \cdot P(\hat{K} = K_1) = \frac{1}{4} \cdot \frac{1}{2} = \frac{1}{8} \\ P(\hat{C} = 2) &= P(\hat{M} = a) \cdot P(\hat{K} = K_2) + P(\hat{M} = b) \cdot P(\hat{K} = K_1) = \frac{1}{4} \cdot \frac{1}{4} + \frac{3}{4} \cdot \frac{1}{2} = \frac{7}{16} \\ P(\hat{C} = 4) &= P(\hat{M} = b) \cdot P(\hat{K} = K_3) = \frac{3}{4} \cdot \frac{1}{4} = \frac{3}{16} \\ \Rightarrow P(\hat{C} = 3) &= 1 - P(\hat{C} = 1) - P(\hat{C} = 2) - P(\hat{C} = 4) = 1 - \frac{2}{16} - \frac{7}{16} - \frac{3}{16} = \frac{4}{16} \\ \Rightarrow H(\hat{C}) &= -\frac{1}{8} \log_2\left(\frac{1}{8}\right) - \frac{7}{16} \log_2\left(\frac{7}{16}\right) - \frac{3}{16} \log_2\left(\frac{3}{16}\right) - \frac{1}{4} \log_2\left(\frac{1}{4}\right) \approx 1.850 \\ \Rightarrow H(\hat{K} | \hat{C}) &\stackrel{\text{Thm. 4.7}}{=} H(\hat{M}) + H(\hat{K}) - H(\hat{C}) \approx 0.811 + 1.5 - 1.850 = 0.461 \end{aligned}$$

- b) Lem. 4.12 b) demands  $|\mathcal{C}_+| \leq |\mathcal{K}_+|$  for perfect secrecy.  
But in this case, we get  $4 = |\mathcal{C}_+| > |\mathcal{K}_+| = 3 \not\leq$

## Solution of Problem 3

Show for any function  $f : X(\Omega) \times Y(\Omega) \rightarrow \mathbb{R}$ , that  $H(X, Y, f(X, Y)) = H(X, Y)$ .

By definition, we have:

$$H(X, Y, Z = f(X, Y)) \stackrel{\text{Def.}}{=} \sum_{X, Y, Z} P(X = x, Y = y, Z = z) \log(P(X = x, Y = y, Z = z))$$

With

$$P(X = x, Y = y, Z = z) = \begin{cases} P(X = x, Y = y) & , \text{ if } Z = f(X, Y) \\ 0 & , \text{ if } Z \neq f(X, Y) \end{cases}$$

it follows that

$$H(X, Y, Z = f(X, Y)) = \sum_{X, Y} P(X = x, Y = y) \log(P(X = x, Y = y)) = H(X, Y).$$

**Note:** It holds  $0 \cdot \log 0 = 0$ .

## Solution of Problem 4

a)

$$H(M) = - \sum_i P(M_i) \log_2 P(M_i) = - \left( \frac{1}{3} \log_2 \frac{1}{3} + \frac{2}{3} \log_2 \frac{2}{3} \right)$$

b) (i) For each  $M \in \mathcal{M}_N, C \in \mathcal{C}_N$  there exists exactly one  $K \in \mathcal{K}_N$  such that  $e(M, K) = C$ , namely  $K = (s_1, \dots, s_N)$  with  $s_j = (c_j - a_j) \pmod m$ .

(ii)  $\tilde{K}_N$  is uniformly distributed over  $\mathcal{K}_N$ , as

$$P(\tilde{K}_N = K) = P(\tilde{K}_1 = s_1, \dots, \tilde{K}_N = s_N) = \prod_{i=1}^N P(\tilde{K}_i = s_i) = \frac{1}{m^N} = \frac{1}{|\mathcal{K}_N|}$$

$$\forall K = (s_1, \dots, s_N)$$

(iii) Disadvantage of Vernam Cipher: The main disadvantage of the Vernam Cipher is that :  $|\mathcal{K}_+| \geq |\mathcal{M}_+|$  (one needs at least as many keys as plaintexts) and these keys need to be communicated over a secure channel in advance.

c)

$$H(C, M) \stackrel{\text{chain-rule}}{=} H(C) + H(M|C) \stackrel{\text{perf.sec.}}{=} H(C) + H(M)$$

d) Due to the independence of  $X$  and  $Y$  we have  $p_Y(y|x) = p_Y(y)$ , and

$$\tilde{H}(Y|X) = - \sum_x \sum_y p_Y(y) \log_2 p_Y(y) = |\mathcal{X}| H(Y) \geq H(Y)$$