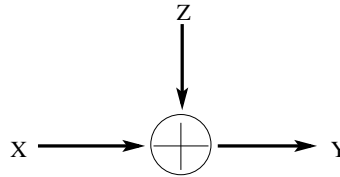


Homework 4 Solutions

1. 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 $\mathbf{X} = \{0, 1\}$. Assume that Z is independent of X . Observe that the channel capacity depends on the value of a .

Solution :

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

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 $\max I(X; Y) = \max H(X) = 1$, achieved for an uniform distribution on the input X .
 - $a = 1$. In this case, Y has three possible output values, 0, 1, and 2. The channel is identical to the binary erasure channel with $a = 1/2$. The capacity of this channel is $1/2$ 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.
2. 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 , \dots , $Z \rightarrow Z$ 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 :

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

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

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

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

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, \dots , 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,

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

In this case, we can achieve capacity with a simple code with zero error.

3. 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 $C = 1 - H(p)$. 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 :

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

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

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

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

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.

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

$$Z = \begin{cases} 1, & \text{with probability } \frac{1}{3} \\ 2, & \text{with probability } \frac{1}{3} \\ 3, & \text{with probability } \frac{1}{3} \end{cases}$$

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

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

Solution :

(a)

$$\begin{aligned}
 C &= \max_{p(x)} I(X; Y) \\
 &= \max_{p(x)} H(Y) - H(Y|X) \\
 &= \max_{p(x)} H(Y) - \log 3 \\
 &= \log 13 - \log 3 \\
 &= \log \frac{13}{3}
 \end{aligned}$$

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

- (b) The capacity is achieved by a uniform distribution on the inputs, that is,

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

5. Using two channels

- (a) 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 capacity of this channel.
- (b) Find the capacity C of the union 2 channels $(\mathcal{X}_1, p(y_1|x_1), \mathcal{Y}_1)$ and $(\mathcal{X}_2, p(y_2|x_2), \mathcal{Y}_2)$ where, at each time, one can send a symbol over channel 1 or channel 2 but not both. Assume the output alphabets are distinct and do not intersect. Show $2^C = 2^{C_1} + 2^{C_2}$.

Solution :

- (a) 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

$$\begin{aligned}
 I(X_1, X_2; Y_1, Y_2) &= H(Y_1, Y_2) - H(Y_1, Y_2|X_1, X_2) \\
 &= H(Y_1, Y_2) - H(Y_1|X_1, X_2) - H(Y_2|X_1, X_2)
 \end{aligned} \tag{1}$$

$$= H(Y_1, Y_2) - H(Y_1|X_1) - H(Y_2|X_2) \tag{2}$$

$$\leq H(Y_1) + H(Y_2) - H(Y_1|X_1) - H(Y_2|X_2) \tag{3}$$

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

where (1) and (2) follow from Markovity and (3) is met with equality of 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.

(b) Let

$$\theta = \begin{cases} 1, & \text{if the signal is sent over the channel 1} \\ 2, & \text{if the signal is sent over the channel 2} \end{cases}$$

Consider the following communication scheme: The sender chooses between two channels according to Bern(α) coin flip. Then the channel input is $X = (\theta, X_\theta)$.

Since the output alphabets \mathcal{Y}_1 and \mathcal{Y}_2 are disjoint, θ is a function of Y , i.e. $X \rightarrow Y \rightarrow \theta$. Therefore,

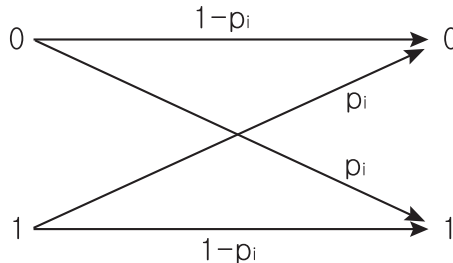
$$\begin{aligned} I(X; Y) &= I(X; Y, \theta) \\ &= I(X_\theta, \theta; Y, \theta) \\ &= I(\theta; Y, \theta) + I(X_\theta; Y, \theta | \theta) \\ &= I(\theta; Y, \theta) + I(X_\theta; Y | \theta) \\ &= H(\theta) + \alpha I(X_\theta; Y | \theta = 1) + (1 - \alpha) I(X_\theta; Y | \theta = 2) \\ &= H(\alpha) + \alpha I(X_1; Y_1) + (1 - \alpha) I(X_2; Y_2) \end{aligned}$$

Thus, it follows that

$$C = \sup_{\alpha} \{H(\alpha) + \alpha C_1 + (1 - \alpha) C_2\}$$

which is a strictly concave function on α . Hence, the maximum exists and by elementary calculus, one can easily show $C = \log_2(2^{C_1} + 2^{C_2})$, which is attained with $\alpha = 2^{C_1} / (2^{C_1} + 2^{C_2})$.

6. Consider a time-varying discrete *memoryless* binary symmetric channel. Let Y_1, Y_2, \dots, Y_n be conditionally independent given X_1, X_2, \dots, X_n , with conditional distribution given by $p(y^n | x^n) = \prod_{i=1}^n p_i(y_i | x_i)$, as shown below.



- (a) Find $\max_{p(x)} I(X^n; Y^n)$.

- (b) We now ask for the capacity for the time invariant version of this problem. Replace each p_i , $1 \leq i \leq n$, by the average value $\bar{p} = \frac{1}{n} \sum_{j=1}^n p_j$, and compare the capacity to part (a).

Solution :

(a)

$$\begin{aligned} I(X^n; Y^n) &= H(Y^n) - \sum_{i=1}^n H(Y_i | X_i) \\ &\leq \sum_{i=1}^n H(Y_i) - \sum_{i=1}^n H(Y_i | X_i) \\ &\leq \sum_{i=1}^n (1 - H(p_i)) \end{aligned}$$

with equality if X_1, \dots, X_n are chosen i.i.d. $\sim \text{Bern}(\frac{1}{2})$. Hence

$$\max_{p(x)} I(X_1, \dots, X_n; Y_1, \dots, Y_n) = \sum_{i=1}^n (1 - H(p_i))$$

- (b) Since $H(p)$ is concave on p , by Jensen's inequality,

$$\frac{1}{n} \sum_{i=1}^n H(p_i) \leq H\left(\frac{1}{n} \sum_{i=1}^n p_i\right) = H(\bar{p})$$

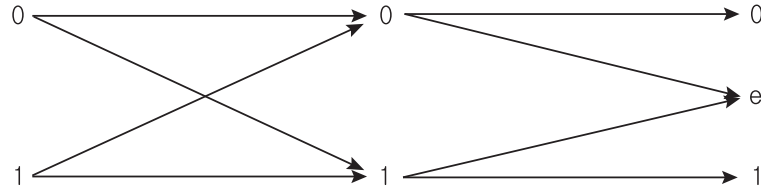
i.e.,

$$\sum_{i=1}^n H(p_i) \leq nH(\bar{p})$$

Hence,

$$\begin{aligned} C_{\text{time-varying}} &= \sum_{i=1}^n (1 - H(p_i)) \\ &= n - \sum_{i=1}^n H(p_i) \\ &\geq n - nH(\bar{p}) \\ &= \sum_{i=1}^n (1 - H(\bar{p})) \\ &= C_{\text{time invariant}} \end{aligned}$$

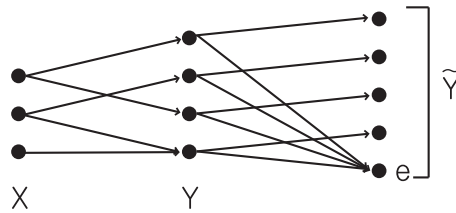
7. Suppose a binary symmetric channel of capacity C_1 is immediately followed by a binary erasure channel of capacity C_2 . Find capacity C of the resulting channel.



Now consider an arbitrary discrete memoryless channel $(\mathcal{X}, p(y|x), \mathcal{Y})$ followed by a binary erasure channel, resulting in an output

$$\tilde{Y} = \begin{cases} Y, & \text{with probability } 1 - \alpha \\ e, & \text{with probability } \alpha \end{cases}$$

where e denotes erasure. Thus the output \mathcal{Y} is erased with probability α . What is the capacity of this channel?



Solution :

- (a) Let $C_1 = 1 - H(p)$ be the capacity of the BSC with parameter p , and $C_2 = 1 - \alpha$ be the capacity of the BEC with parameter α . Let \tilde{Y} denote the output of the cascaded channel, and Y the output of the BSC. Then, the transition rule for the cascaded channel is simply

$$p(\tilde{y}|x) = \sum_{y=0,1} p(\tilde{y}|y)p(y|x)$$

for each (x, \tilde{y}) pair.

Let $X \sim \text{Bern}(\pi)$ denote the input to the channel. Then,

$$H(\tilde{Y}) = H((1 - \alpha)(\pi(1 - p) + p(1 - \pi)), \alpha, (1 - \alpha)(p\pi + (1 - p)(1 - \pi)))$$

and also

$$H(\tilde{Y}|X = 0) = H((1 - \alpha)(1 - p), \alpha, (1 - \alpha)p)$$

$$H(\tilde{Y}|X = 1) = H((1 - \alpha)p, \alpha, (1 - \alpha)(1 - p)) = H(\tilde{Y}|X = 0)$$

Therefore,

$$\begin{aligned} C &= \max_{p(x)} I(X; \tilde{Y}) \\ &= \max_{p(x)} \{H(\tilde{Y}) - H(\tilde{Y}|X)\} \\ &= \max_{p(x)} \{H(\tilde{Y})\} - H(\tilde{Y}|X) \\ &= \max_{p(x)} \{H((1 - \alpha)(\pi(1 - p) + p(1 - \pi)), \alpha, (1 - \alpha)(p\pi + (1 - p)(1 - \pi)))\} \\ &\quad - H((1 - \alpha)(1 - p), \alpha, (1 - \alpha)p) \end{aligned} \tag{4}$$

Note that the maximum value of $H(\tilde{Y})$ occurs when $\pi = 1/2$ by the concavity and symmetry of $H(\cdot)$. (We can check this also by differentiating (4) with respect to π .) Substituting the value $\pi = 1/2$ in the expression for the capacity yields

$$\begin{aligned} C &= H((1-\alpha)/2, \alpha, (1-\alpha)/2) - H((1-p)(1-\alpha), \alpha, p(1-\alpha)) \\ &= (1-\alpha)(1 + p \log p + (1-p) \log(1-p)) \\ &= C_1 C_2 \end{aligned}$$

- (b) For the cascade of an arbitrary discrete memoryless channel (with capacity C) with the erasure channel (with the erasure probability α), we will show that

$$I(X; \tilde{Y}) = (1-\alpha)I(X; Y) \quad (5)$$

Then, by taking suprema of both sides over all input distributions $p(x)$, we can conclude the capacity of the cascaded channel is $(1-\alpha)C$.

Proof of (5):

Let

$$E = \begin{cases} 1, & \tilde{Y} = e \\ 0, & \tilde{Y} = Y \end{cases}$$

Then, since E is a function of Y ,

$$\begin{aligned} H(\tilde{Y}) &= H(\tilde{Y}, E) \\ &= H(E) + H(\tilde{Y}|E) \\ &= H(\alpha) + \alpha H(\tilde{Y}|E=1) + (1-\alpha)H(\tilde{Y}|E=0) \\ &= H(\alpha) + (1-\alpha)H(Y), \end{aligned}$$

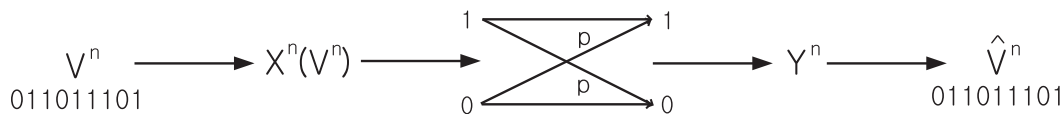
where the last equality comes directly from the construction of E . Similarly,

$$\begin{aligned} H(\tilde{Y}|X) &= H(\tilde{Y}, E|X) \\ &= H(E|X) + H(\tilde{Y}|X, E) \\ &= H(E) + \alpha H(\tilde{Y}|X, E=1) + (1-\alpha)H(\tilde{Y}|X, E=0) \\ &= H(\alpha) + (1-\alpha)H(Y|X), \end{aligned}$$

whence

$$I(X; \tilde{Y}) = H(\tilde{Y}) - H(\tilde{Y}|X) = (1-\alpha)I(X; Y)$$

8. We wish to encode a Bernoulli(α) process V_1, V_2, \dots for transmission over a binary symmetric channel with error probability p .



Find conditions on α and p so that the probability of error $p(\hat{V}^n \neq V^n)$ can be made to go to zero as $n \rightarrow \infty$.

Solution :

Suppose we want to send a binary i.i.d. Bern(α) source over a binary symmetric channel with error

probability p . By the source-channel separation theorem, in order to achieve the probability of error that vanishes asymptotically, i.e. $P(\hat{V}^n \neq V^n) \rightarrow 0$, we need the entropy of the source to be less than the capacity of the channel. Hence,

$$H(\alpha) + H(p) < 1,$$

or, equivalently,

$$\alpha^\alpha (1 - \alpha)^{1-\alpha} p^p (1 - p)^{1-p} < \frac{1}{2}.$$

9. Let (X_i, Y_i, Z_i) be i.i.d. according to $p(x, y, z)$. We will say that (x^n, y^n, z^n) is jointly typical [written $(x^n, y^n, z^n) \in A_\epsilon^{(n)}$] if

- $2^{-n(H(X)+\epsilon)} \leq p(x^n) \leq 2^{-n(H(X)-\epsilon)}$
- $2^{-n(H(Y)+\epsilon)} \leq p(y^n) \leq 2^{-n(H(Y)-\epsilon)}$
- $2^{-n(H(Z)+\epsilon)} \leq p(z^n) \leq 2^{-n(H(Z)-\epsilon)}$
- $2^{-n(H(X,Y)+\epsilon)} \leq p(x^n, y^n) \leq 2^{-n(H(X,Y)-\epsilon)}$
- $2^{-n(H(X,Z)+\epsilon)} \leq p(x^n, z^n) \leq 2^{-n(H(X,Z)-\epsilon)}$
- $2^{-n(H(Y,Z)+\epsilon)} \leq p(y^n, z^n) \leq 2^{-n(H(Y,Z)-\epsilon)}$
- $2^{-n(H(X,Y,Z)+\epsilon)} \leq p(x^n, y^n, z^n) \leq 2^{-n(H(X,Y,Z)-\epsilon)}$

Now suppose that $(\tilde{X}^n, \tilde{Y}^n, \tilde{Z}^n)$ is drawn according to $p(x^n)p(y^n)p(z^n)$. Thus, $\tilde{X}^n, \tilde{Y}^n, \tilde{Z}^n$ have the same marginals as $p(x^n, y^n, z^n)$ but are independent. Find (bounds on) $\Pr\{(\tilde{X}^n, \tilde{Y}^n, \tilde{Z}^n) \in A_\epsilon^{(n)}\}$ in terms of the entropies $H(X)$, $H(Y)$, $H(Z)$, $H(X, Y)$, $H(X, Z)$, $H(Y, Z)$ and $H(X, Y, Z)$.

Solution :

$$\begin{aligned} \Pr\{(\tilde{X}^n, \tilde{Y}^n, \tilde{Z}^n) \in A_\epsilon^{(n)}\} &= \sum_{(x^n, y^n, z^n) \in A_\epsilon^{(n)}} p(x^n)p(y^n)p(z^n) \\ &\leq \sum_{(x^n, y^n, z^n) \in A_\epsilon^{(n)}} 2^{-n(H(X)+H(Y)+H(Z)-3\epsilon)} \\ &\leq |A_\epsilon^{(n)}| 2^{-n(H(X)+H(Y)+H(Z)-3\epsilon)} \\ &\leq 2^{n(H(X,Y,Z)+\epsilon)} 2^{-n(H(X)+H(Y)+H(Z)-3\epsilon)} \\ &\leq 2^{n(H(X,Y,Z)-H(X)-H(Y)-H(Z)+4\epsilon)} \end{aligned}$$

Also,

$$\begin{aligned} \Pr\{(\tilde{X}^n, \tilde{Y}^n, \tilde{Z}^n) \in A_\epsilon^{(n)}\} &= \sum_{(x^n, y^n, z^n) \in A_\epsilon^{(n)}} p(x^n)p(y^n)p(z^n) \\ &\geq \sum_{(x^n, y^n, z^n) \in A_\epsilon^{(n)}} 2^{-n(H(X)+H(Y)+H(Z)+3\epsilon)} \\ &\geq |A_\epsilon^{(n)}| 2^{-n(H(X)+H(Y)+H(Z)+3\epsilon)} \\ &\geq (1 - \epsilon) 2^{n(H(X,Y,Z)-\epsilon)} 2^{-n(H(X)+H(Y)+H(Z)-3\epsilon)} \\ &\geq (1 - \epsilon) 2^{n(H(X,Y,Z)-H(X)-H(Y)-H(Z)-4\epsilon)} \end{aligned}$$

Note that the upper bound is true for all n , but the lower bound only hold for n large.

10. Twenty questions.

- (a) Player A chooses some object in the universe, and player B attempts to identify the object with a series of yes-no questions. Suppose that player B is clever enough to use the code achieving the minimal expected length with respect to player A 's distribution. We observe that player B requires an average 38.5 questions to determine the object. Find a rough lower bound to the number of objects in the universe.
- (b) Let X be uniformly distributed over $\{1, 2, \dots, m\}$. Assume that $m = 2^n$. We ask random questions: Is $X \in S_1$? Is $X \in S_2$? \dots until only one integer remains. All 2^m subsets S of $\{1, 2, \dots, m\}$ are equally likely.
- How many deterministic questions are needed to determine X ?
 - Without loss of generality, suppose that $X = 1$ is the random object. What is the probability that object 2 yields the same answers as object 1 for k questions?
 - What is the expected number of objects in $\{2, 3, \dots, m\}$ that have the same answers to the questions as those of the correct object 1?

Solution :

(a)

$$37.5 = L^* - 1 < H(X) \leq \log |\mathcal{X}|$$

and hence number of objects in the universe $> 2^{37.5} = 1.94 \times 10^{11}$.

- (b) i. Obviously, Huffman codewords for X are all of length n . Hence, with n deterministic questions, we can identify an object out of 2^n candidates.
- ii. Observe that the total number of subsets which include both object 1 and object 2 or neither of them is 2^{m-1} . Hence, the probability that object 2 yields the same answers for k questions as object 1 is $(2^{m-1}/2^m)^k = 2^{-k}$.
- iii. Let

$$1_j = \begin{cases} 1, & \text{object } j \text{ yields the same answers for } k \text{ questions as object 1} \\ 0, & \text{otherwise.} \end{cases} \quad \text{for } j = 2, \dots, m$$

Then

$$\begin{aligned} E[N] &= E \left[\sum_{j=2}^m 1_j \right] \\ &= \sum_{j=2}^m E[1_j] \\ &= \sum_{j=2}^m 2^{-k} \\ &= (m-1)2^{-k} \\ &= (2^n - 1)2^{-k} \end{aligned}$$

where N is the number of objects in $\{2, 3, \dots, m\}$ with the same answers.