

PROBLEM 1.

(a) By the chain rule we have

$$\begin{aligned} H(X^n, Z^n) &= H(X^n) + H(Z^n|X^n) \\ &= H(Z^n) + H(X_1|Z^n) + H(X_2^n|X_1, Z^n). \end{aligned}$$

(where  $X_2^n = (X_2, \dots, X_n)$ ). Since  $(X_1, Z^n)$  determines  $X_2^n$  (because,  $X_1$  determines  $\hat{X}_2$  and, knowing  $(\hat{X}_2, Z_2)$ ,  $X_2$  is known, then  $(X_1, X_2)$  determines  $\hat{X}_3$  which, together with  $Z_3$  determines  $X_3, \dots$ ),  $H(X_2^n|X_1, Z^n) = 0$ . Therefore,

$$H(Z^n) = H(X^n) - H(X_1|Z^n)$$

which (together with the fact that  $H(X_1|Z^n) \leq H(X_1) \leq 1$ ) implies

$$H(X^n) - 1 \leq H(Z^n) \leq H(X^n)$$

Consequently

$$\mathcal{H}(Z) = \lim_{n \rightarrow \infty} \frac{1}{n} H(Z^n) = \lim_{n \rightarrow \infty} \frac{1}{n} H(X^n) = \mathcal{H}(X).$$

(b) Once again using the chain rule,

$$\begin{aligned} H(Z_n, X_n|X^{n-1}) &= H(Z_n|X^{n-1}) + H(X_n|Z_n, X^{n-1}) \\ &= H(X_n|X^{n-1}) + H(Z_n|X^n), \end{aligned}$$

and since  $H(X_n|Z_n, X^{n-1}) = H(Z_n|X^n) = 0$  (for the reasons explained in (a)),

$$h_2(p_n) = H(Z_n) \geq H(Z_n|X^{n-1}) = H(X_n|X^{n-1}).$$

(c) From (b) and since  $X$  is a stationary process we have

$$\liminf_{n \rightarrow \infty} h_2(p_n) \geq \lim_{n \rightarrow \infty} H(X_n|X^{n-1}) = \mathcal{H}(X).$$

Moreover, there exists a subsequence of  $\{p_n : n = 1, 2, \dots\}$ , say  $\{p_{n_i} : n_i \in \mathbb{N}, i = 1, 2, \dots\}$  for which  $\liminf_{n \rightarrow \infty} p_n = \lim_{i \rightarrow \infty} p_{n_i}$ . Thus, by the continuity of  $h_2(\cdot)$ ,

$$\liminf_{n \rightarrow \infty} h_2(p_n) = \lim_{i \rightarrow \infty} h_2(p_{n_i}) = h_2(\lim_{i \rightarrow \infty} p_{n_i}) = h_2(p).$$

PROBLEM 2.

(a) Since the channel is memoryless and feedback-free transmission is assumed, from the code construction, it is obvious that  $(\text{enc}_1(m_1), \text{enc}_2(m_2), Y^n)$  is an i.i.d. length- $n$  sequence of  $(X_1, X_2, Y)$ 's drawn from distribution  $p(x_1, x_2, y) = p_1(x_1)p_2(x_2)p(y|x_1, x_2)$ . Therefore, for sufficiently large  $n$ , the probability of this sequence being  $\epsilon$ -typical is as high as desired.

- (b) Now,  $(\text{enc}_1(\tilde{m}_1), \text{enc}_2(m_2), Y^n)$  is an i.i.d. sequence (of length  $n$ ) whose components are distributed according to  $p_1(x_1)p(y, x_2)$  where  $p(y, x_2) = \sum_{x'_1} p_1(x'_1)p_2(x_2)p(y|x'_1, x_2)$ .
- (c)  $\Pr\{(\text{enc}_1(\tilde{m}_1), \text{enc}_2(m_2), Y^n) \in T\}$  is the probability of a length  $n$  i.i.d. sequence  $X_1^n$  whose elements have distribution  $p_1$  being jointly  $\epsilon$ -typical (with respect to the distribution  $p_1(x_1)p(y, x_2|x_1)$  where  $p(y, x_2|x_1) = p(x_2)p(y|x_1, x_2)$ ) with an independent length  $n$  sequence of  $(X_2, Y)$ 's drawn from distribution  $p(y, x_2)$  (defined in (b)). Thus,

$$\Pr\{(\text{enc}_1(\tilde{m}_1), \text{enc}_2(m_2), Y^n) \in T\} \doteq 2^{-nI(X_1, X_2Y)}.$$

(In the course we have seen this result for two random variables  $X$  and  $Y$ ; it is obvious that we can replace  $X$  by  $X_1$  and  $Y$  by  $(X_2, Y)$  to derive the above result).

- (d) From (a) we know that the probability of the correct message  $m_1$  not being in the list of typical  $m_1$ 's at decoder 2 is small, say at most  $\epsilon/2$ .

From (c), the probability of each incorrect  $\tilde{m}_1$  being on that list (at decoder 2) is equal (up to sub-exponential factors) to  $2^{-nI(X_1; X_2Y)}$ . Since there are  $M - 1 \leq 2^{nR_1}$  such  $\tilde{m}_1$ 's, the probability of having *an* incorrect message on the list is, by the union bound, at most  $2^{n[R_1 - I(X_1; X_2Y)]}$  which is exponentially small in  $n$  provided that  $R_1 < I(X_1; X_2Y)$ . Thus, for large enough  $n$ , this probability is also smaller than  $\epsilon/2$ .

Consequently, the average probability of decoding error at decoder 2 is at most  $\epsilon$  provided that  $R_1 < I(X_1, X_2Y)$ .

By symmetry, the average probability of decoding error at decoder 1 is smaller than  $\epsilon$  if  $R_2 < I(X_2, X_1, Y)$ .

Since the average probability of error (over the generation of codebooks) is small (for rate pairs  $(R_1, R_2)$  satisfying  $R_1 < I(X_1; Y, X_2)$  and  $R_2 < I(X_2; Y, X_1)$ ), there exists a pair of codebooks of rates  $(R_1, R_2)$  in the ensemble for which the average error probability is small, thus such  $(R_1, R_2)$ 's are achievable.

- (e) Firstly note that since  $X_1$  and  $X_2$  are independent,  $I(X_1; YX_2) = I(X_1; Y|X_2)$  (similarly  $I(X_2; YX_1) = I(X_2; Y|X_1)$ ).

Since  $Y = X_1 \times X_2$ , conditioned on  $\{X_2 = 0\}$ ,  $Y$  contains no information about  $X_1$ , whereas conditioned on  $\{X_2 = 1\}$ ,  $Y = X_1$ . Thus, assuming  $\Pr\{X_1 = 1\} = p_1$  and  $\Pr\{X_2 = 1\} = p_2$ ,

$$\begin{aligned} I(X_1; Y|X_2) &= \Pr\{X_2 = 0\}I(X_1; Y|X_2 = 0) + \Pr\{X_2 = 1\}I(X_1; Y|X_2 = 1) \\ &= 0 + p_2h_2(p_1) \end{aligned}$$

where  $h_2(\cdot)$  is the binary entropy function. Similarly it follows that

$$I(X_2; Y|X_1) = p_1h_2(p_2).$$

Suppose  $p_1 = p_2 = p$ , then all rates  $(R_1, R_2)$  satisfying

$$R_1 < ph_2(p) \quad R_2 < ph_2(p)$$

are achievable. In particular,  $ph_2(p) \geq \frac{1}{2}$  for some  $p \geq \frac{1}{2}$  (it evaluates to  $\frac{1}{2}$  at  $p = \frac{1}{2}$  but it is increasing, so it will go above  $\frac{1}{2}$  as  $p$  increases). The set of achievable rate pairs corresponding to such  $p$ 's violate  $R_1 + R_2 < 1$ .

PROBLEM 3.

- (a) For large enough  $n$ ,  $(X^n, Y^n)$  is jointly  $\epsilon$ -typical with respect to the distribution  $p(x, y)$ . Thus, the chance of the true sequence  $X^n$  not appearing on Bob's list vanishes as  $n$  gets large.
- (b) For an  $\epsilon$ -typical sequence  $y^n$ ,  $p(y^n) \leq 2^{-n(1-\epsilon)H(Y)}$ . Similarly, for jointly  $\epsilon$ -typical sequences  $(x^n, y^n)$ ,  $p(x^n, y^n) \geq 2^{-n(1+\epsilon)H(X,Y)}$ . We, then, have

$$\begin{aligned} 2^{-n(1-\epsilon)H(Y)} &\geq p(y^n) \\ &= \sum_{x^n} p(x^n, y^n) \\ &\geq \sum_{\substack{x^n: \\ (x^n, y^n) \in T_\epsilon}} p(x^n, y^n) \\ &\geq |\{x^n : (x^n, y^n) \in T_\epsilon\}| 2^{-n(1+\epsilon)H(X,Y)}, \end{aligned}$$

where  $T_\epsilon$  denotes the set of jointly typical  $(x^n, y^n)$ 's. Therefore, by noticing that  $(1 + \epsilon)H(X, Y) - (1 - \epsilon)H(Y) = (1 + \epsilon)H(X|Y) + 2\epsilon H(Y)$ , for an  $\epsilon$ -typical  $y^n$ ,

$$|\{x^n : (x^n, y^n) \in T_\epsilon\}| \leq 2^{n[(1+\epsilon)H(X|Y)+2\epsilon H(Y)]} \approx 2^{nH(X|Y)}.$$

- (c) Given a typical sequence  $y^n$ ,  $x^n$  appears on Bob's list if  $(x^n, y^n)$  is typical and  $\text{label}(x^n) = \text{label}(x_0^n)$  (where  $x_0^n$  is the true sequence which we assume to be typical as well – otherwise Bob will not receive any label from Alice). The number of wrong sequences is thus

$$N_w(x_0^n, y^n) := \sum_{\substack{x^n: \\ (x^n, y^n) \in T_\epsilon \\ x^n \neq x_0^n}} \mathbf{1}\{\text{label}(x^n) = \text{label}(x_0^n)\}.$$

Since the labels are assigned independently and uniformly from  $\{1, \dots, 2^{nR}\}$ ,

$$E[\mathbf{1}\{\text{label}(x^n) = \text{label}(x_0^n)\}] = \Pr\{\text{label}(x^n) = \text{label}(x_0^n)\} = 2^{-nR} \quad (\text{if } x^n \neq x_0^n).$$

Consequently, using (b) we have:

$$E[N_w(x_0^n, y^n)] = |\{x^n : (x^n, y^n) \in T_\epsilon\} \setminus \{x_0^n\}| 2^{-nR} \leq 2^{-n[R - (H(X|Y) + \delta)]},$$

for some  $\delta = \delta(\epsilon)$  which goes to 0 as  $\epsilon \rightarrow 0$ .

- (d) A decoding error will happen if either  $(X^n, Y^n)$  are atypical or they are typical but Bob's list has more than one element. In other words,

$$\Pr\{\text{error}\} = \Pr\{(X^n, Y^n) \notin T_\epsilon\} + \Pr\{(X^n, Y^n) \in T_\epsilon, N_w(X^n, Y^n) \geq 1\}.$$

The first term on the right-hand-side of the above goes to 0 as  $n$  gets large (independent of  $R$ ). For the second term we have

$$\begin{aligned} &\Pr\{(X^n, Y^n) \in T_\epsilon, N_w(X^n, Y^n) \geq 1\} \\ &= \Pr\{(X^n, Y^n) \in T_\epsilon\} \Pr\{N_w(X^n, Y^n) \geq 1 | (X^n, Y^n) \in T_\epsilon\} \\ &\leq \Pr\{N_w(X^n, Y^n) \geq 1 | (X^n, Y^n) \in T_\epsilon\} \\ &\stackrel{(*)}{\leq} E[N_w(X^n, Y^n) | (X^n, Y^n) \in T_\epsilon] \\ &\leq 2^{-n[R - H(X|Y) - \delta]}, \end{aligned}$$

where (\*) follows from the Markov inequality.

Thus, if  $R > H(X|Y)$  the second term also vanishes as  $n$  gets large which means Bob will decide correctly with high probability.

PROBLEM 4.

- (a) Let  $\mathbf{x}'$  and  $\mathbf{y}'$  be two codewords in  $\mathcal{C}'$  corresponding to information vectors  $\mathbf{u} = (u_0, u_1, \dots, u_{k-1})$  and  $\mathbf{v} = (v_0, \dots, v_{k-1})$  respectively.  $\alpha\mathbf{x}' + \beta\mathbf{y}'$ ,  $\alpha \in \mathbb{F}$ ,  $\beta \in \mathbb{F}$  corresponds to the encoding of  $\alpha\mathbf{u} + \beta\mathbf{v} = (\alpha u_0 + \beta v_0, \alpha u_1 + \beta v_1, \dots, \alpha u_{k-1} + \beta v_{k-1})$ , hence is a codeword of  $\mathcal{C}'$  as well.
- (b) The number of zeros in  $(x_1, \dots, x_n)$  is the number of roots the polynomial  $u(D) = u_0 + u_1D + \dots + u_{k-2}D^{k-2}$  (note that  $u_{k-1} = 0$ ) has among  $\{\alpha_1, \dots, \alpha_n\}$ . A polynomial of degree at most  $k-2$  has at most  $k-2$  roots, and thus a weight of  $\text{weight}(x_1, \dots, x_n) \geq n - (k-2) = n + 2 - k$ . Since  $u_{k-1} = 0$ ,  $\text{weight}(u_{k-1}, x_1, \dots, x_n) = \text{weight}(x_1, \dots, x_n) \geq n + 2 - k$ .
- (c) Since  $u_{k-1} \neq 0$ ,  $\text{weight}(u_{k-1}, x_1, \dots, x_n) = 1 + \text{weight}(x_1, \dots, x_n)$ . Now among  $x_1, \dots, x_n$  at most  $k-1$  elements can be zero (since they are evaluations of a polynomial of degree  $k-1$ ), hence  $\text{weight}(x_1, \dots, x_n) \geq n + 1 - k$ . Thus,  $\text{weight}(\mathbf{x}') \geq n + 2 - k$ .
- (d) From (a), (b) and (c) we have  $d_{\min}(\mathcal{C}') = \min_{\mathbf{x}' \in \mathcal{C}'} \text{weight}(\mathbf{x}') \geq n + 2 - k$ . On the other hand, the Singleton bound states that for any linear code of blocklength  $n + 1$  and dimension  $k$ ,  $d_{\min} \leq n - k + 2$ . This shows the code  $\mathcal{C}'$  has minimum distance exactly equal to  $n - k + 2$ .