

**Solutions: Homework Set # 4**  
**Principles of Wireless Networks**

**Problem 1 (Min Cut-Rank Relation)**

This proof is essentially based on the capacity characterization of the coherent network, where the channel matrix is known at the receiver. It is shown that in a coherent deterministic network, using random linear mapping at the relays, for any  $\epsilon > 0$ , the rate  $R = \min_{\Omega} \text{rank}(\mathbf{G}_{\Omega, \Omega^c}) - \epsilon$  is achievable with probability at least  $1 - 2^{-T\epsilon 2^{|V|}}$ . Therefore, there exist at least  $2^{TR}$  vectors in the range space of  $\mathbf{H}$  with probability more than  $1 - 2^{-T\epsilon 2^{|V|}}$ . The fact that the dimension of the range of a matrix is upper bounded by its rank, implies

$$\Pr[\text{rank}(\mathbf{H}) < T(\min_{\Omega} \text{rank}(\mathbf{G}_{\Omega, \Omega^c}) - \epsilon)] < 2^{|V|} 2^{-T\epsilon},$$

**Problem 2 (Multi-Sources Deterministic Network)**

This problem is a part of a published paper “On Noise Strategies for Wireless Network Secrecy,” by E. Perron, S. Diggavi, and E. Telatar.

- (a). (i) is true by the symmetry of the code construction.
  - (ii) holds since if the decoder cannot distinguish between  $(w_1, w_2)$  and  $(w'_1, w'_2)$ , it can only choose one of them randomly as the decoded message, and there is a non-zero probability to choose the wrong one. However, if it can distinguish, it never makes a mistake.
- (b). (i) is just a union bound (it can be replaced by equality, since the events are disjoint).
  - (ii) holds because the terms for which  $\mathcal{N}(\mathcal{S}_I) \not\subseteq \Omega^c$  are zero. That is because a node in  $\mathcal{N}(\mathcal{S}_I)$  does not receive anything from  $\mathcal{S}_I$  and therefore cannot distinguish between  $(W_{I^c}, W_I)$  and  $(W_{I^c}, W'_I)$ .
  - (iii) is due to  $\Pr[A \& B | C] = \Pr[A | C] \Pr[B | A \& C] \leq \Pr[B | A \& C]$ .
- (c). The proof of this inequality is similar to the error probability analysis for deterministic network we have seen in the class.
- (d). The number of terms in  $\sum_{w'_1 \neq w_1}$  is the number of different messages can be transmitted by the first source which is  $2^{TR_1} - 1$ . Similarly, we can show that the number of terms in the second and third summations are  $2^{TR_2} - 1$ , and  $(2^{TR_1} - 1)(2^{TR_2} - 1) < 2^{T(R_1 + R_2)}$ .
- (e). It is shown in part (c) that any inner expectation in (b) is upper bounded by  $2^{-TH(Y_{\Omega^c} | X_{\Omega^c})} \leq 2^{-T \min_{\Omega \in \Lambda} H(Y_{\Omega^c} | X_{\Omega^c})}$ , where the last term does not depend on the particular cut. Therefore the whole summation can be bounded by the number of terms  $(|\Lambda|)$  times the maximum one. Putting all together we obtain (e).

- (f). Note that the upper bound on  $\mathbb{E}[\text{Pr}(\text{err})]$  goes to zero, if and only if all three terms go to zero. The number of cuts are constant, and do not increase as  $T \rightarrow 0$ . Here are the constraints on  $(R_1, R_2)$  to have vanishing error probability.

$$R_1 < \min_{\Omega \in \tilde{\Lambda}_{\{s_1\}}} H(Y_{\Omega^c} | X_{\Omega^c}) \quad (1)$$

$$R_2 < \min_{\Omega \in \tilde{\Lambda}_{\{s_2\}}} H(Y_{\Omega^c} | X_{\Omega^c}) \quad (2)$$

$$R_1 + R_2 < \min_{\Omega \in \tilde{\Lambda}_{\{s_1, s_2\}}} H(Y_{\Omega^c} | X_{\Omega^c}). \quad (3)$$

- (g). Since  $\tilde{\Lambda}_I \subseteq \Lambda_I$ , the minimization in the LHS is more restricting, and its value can not be less than the value of the RHS. The other inequality follows from the chain as follows.

$$\begin{aligned} H(Y_{\Omega'^c} | X_{\Omega'^c}) &\stackrel{(i)}{=} H(Y_{\Omega^{*c}} | X_{\Omega'^c}) + H(Y_{\mathcal{N}(I) \setminus \Omega^{*c}} | Y_{\Omega^{*c}} X_{\Omega'^c}) \\ &\stackrel{(ii)}{=} H(Y_{\Omega^{*c}} | X_{\Omega'^c}) \\ &\stackrel{(iii)}{\leq} H(Y_{\Omega^{*c}} | X_{\Omega^{*c}}) \\ &\stackrel{(iv)}{=} \min_{\Omega \in \Lambda_I} H(Y_{\Omega^c} | X_{\Omega^c}), \end{aligned}$$

where

(i) is by using chain rule.

(ii) is true because  $\Omega'^c$  contains  $\mathcal{N}(I)$ , and since  $\mathcal{N}(I)$  is in flow of itself,  $Y_{\mathcal{N}(I)}$  is a function of  $X_{\mathcal{N}(I)}$ .

(iii) holds since  $\Omega^{*c} \subseteq \Omega'^c$ , and conditioning reduces entropy.

(iv) is due to the assumption that  $\Omega^*$  is the minimizer cut.

### Problem 3 (Typicality in deterministic channels)

- (a). Note that we have

$$p(y|x=a) = \begin{cases} 1 & \text{if } y = f(a) \\ 0 & \text{else.} \end{cases}$$

Therefore,  $p(x, y) = p(x)p(y|x) = p(x)\mathbf{1}_{[y=f(x)]}$ , where  $\mathbf{1}_{[\cdot]}$  is the indicator function. For  $\underline{x}$  and  $y = f(\underline{x})$ , we have

$$\mathbf{v}_{\underline{x}, y}(x, y) = \frac{1}{T} |\{t : (x_t, y_t) = (x, f(x))\}| = \frac{1}{T} |\{t : x_t = x\}| = \mathbf{v}_{\underline{x}}(x).$$

Therefore we have

$$\begin{aligned} |\mathbf{v}_{\underline{x}, y}(x, y) - p(x, y)| &\leq |\mathbf{v}_{\underline{x}}(x) - p(x)| \\ &\leq \delta p(x) \\ &= \delta p(x, y), \end{aligned}$$

where the second inequality holds since  $\underline{x} \in T_\delta$ .

- (b). Let  $(\underline{x}, y)$  be jointly typical. For a pair of  $(x, y)$  with  $y \neq f(x)$ , we have  $p(x, y) = p(x)p(y|x) = 0$ . Hence definition of jointly typical sequences implies

$$|\mathbf{v}_{\underline{x}, y}(x, y) - 0| = |\mathbf{v}_{\underline{x}, y}(x, y) - p(x, y)| \leq \delta p(x, y) = 0,$$

and therefore  $v_{\underline{x}, \underline{y}}(x, y) = 0$ . We also know that

$$v_{\underline{x}}(x) = \sum_{y \in \mathcal{Y}} v_{\underline{x}, \underline{y}}(x, y) = v_{\underline{x}, \underline{y}}(x, f(x)),$$

where the last equality holds since there is only one non-zero term in the summation. This implies  $\underline{y} = f(\underline{x})$ . Moreover,

$$\begin{aligned} |v_{\underline{x}}(x) - p(x)| &= |v_{\underline{x}, \underline{y}}(x, f(x)) - p(x)p(f(x)|y)| \\ &\leq \delta p(x)p(f(x)|y) \\ &= \delta p(x), \quad \forall x \in \mathcal{X} \end{aligned}$$

which is the definition of typicality for  $\underline{x}$ . Note that the inequality comes from the definition of typicality for  $(x, y) = (x, f(x))$ .