

Homework Set # 4 Principles of Wireless Networks

Problem 1 (Min Cut-Rank Relation)

Consider a linear deterministic layered network. Let \mathbf{G}_{ij} be the $q \times q$ channel matrix from node i to node j , and

$$\mathbf{y}_j[t] = \sum_i \mathbf{G}_{ij} \mathbf{x}_i[t],$$

where $\mathbf{x}_i[t]$ and $\mathbf{y}_j[t]$ are the transmitted vector from node i and received vector at node j , respectively. The relay nodes wait till they receive T vectors and then they perform their own operations \mathbf{F}_j to obtain the transmitting vectors in the next block.

Note that the end-to-end transfer function \mathbf{H} is formed as a combination of channel matrices $\mathbf{I}_T \otimes \mathbf{G}_{ij}$, and the operations performed at the relays \mathbf{F}_j ,

$$\mathbf{Y}_D = \mathbf{H}\mathbf{X}_S.$$

Show that for uniform random choices of the relay operations, we have

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

where $|V|$ is the number of nodes in the network.

Problem 2 (Multi-Sources Deterministic Network)

Consider a linear deterministic network with two sources S_1 and S_2 , where S_i wants to communicate to the destination node at rate R_1 . Assume the network is layered with respect to both sources. Here We will derive an inner bound (achievable region) for $\mathcal{R} = \{(R_1, R_2) : (R_1, R_2) \text{ is achievable}\}$, using the following scheme.

Fix a block length T and an arbitrary product distribution $\prod_{i \in \mathcal{V}} p(x_i)$. The source nodes maps their messages $w_i \in \{1, 2, \dots, 2^{TR_i}\}$ to transmitting sequence of vectors $\mathbf{x}_i[t]$, $t = 1, \dots, T$ chosen uniformly at random according $p(x_{S_i})$ (random codebook). Having T vectors received at the relay node i , it maps its received sequence \mathbf{y}_i to \mathbf{x}_i where \mathbf{x}_i is chosen uniformly at random according to $p(x_i)$ (random mapping operation). In the following we will show that the error probability of this scheme is vanishing as T grows.

(a). Justify the following chain of equality and inequalities.

$$\begin{aligned} \mathbb{E}[\Pr(\text{err})] &\stackrel{(i)}{=} \mathbb{E}[\Pr(\text{error at } D | (w_1, w_2) \text{ is sent})] \\ &\stackrel{(ii)}{\leq} \sum_{(w'_1, w'_2) \neq (w_1, w_2)} \mathbb{E}[\Pr(D \text{ cannot distinguish } (w'_1, w'_2) \text{ and } (w_1, w_2) | (w_1, w_2) \text{ is sent})] \\ &= \sum_{w'_1 \neq w_1} \mathbb{E}[\Pr(D \text{ cannot distinguish } (w'_1, w_2) \text{ and } (w_1, w_2) | (w_1, w_2) \text{ is sent})] \\ &\quad + \sum_{w'_2 \neq w_2} \mathbb{E}[\Pr(D \text{ cannot distinguish } (w_1, w'_2) \text{ and } (w_1, w_2) | (w_1, w_2) \text{ is sent})] \\ &\quad + \sum_{\substack{w'_1 \neq w_1 \\ w'_2 \neq w_2}} \mathbb{E}[\Pr(D \text{ cannot distinguish } (w'_1, w'_2) \text{ and } (w_1, w_2) | (w_1, w_2) \text{ were sent})] \quad (1) \end{aligned}$$

- (b). We can continue by bounding each of the terms in the RHS of above inequality. For each $\emptyset \neq I \subseteq \{1, 2\}$, we define $\mathcal{S}_I = \{S_i : i \in I\}$ and $W_I = \{w_i : i \in I\}$. Show that

$$\begin{aligned} & \mathbb{E}[\Pr(D \text{ cannot distinguish } (W_{I^c}, W'_I) \text{ and } (W_{I^c}, W_I) | (w_1, w_2) \text{ is sent})] \\ & \stackrel{(i)}{\leq} \sum_{\Omega \in \Lambda_I} \mathbb{E} \left[\Pr \left(\begin{array}{c} \Omega \text{ can distinguish } (W_{I^c}, W'_I) \text{ and } (W_{I^c}, W_I) \\ \text{but } \Omega^c \text{ cannot} \end{array} \middle| (w_1, w_2) \text{ were sent} \right) \right] \\ & \stackrel{(ii)}{\leq} \sum_{\Omega \in \tilde{\Lambda}_I} \mathbb{E} \left[\Pr \left(\begin{array}{c} \Omega \text{ can distinguish } (W_{I^c}, W'_I) \text{ and } (W_{I^c}, W_I) \\ \text{but } \Omega^c \text{ cannot} \end{array} \middle| (w_1, w_2) \text{ is sent} \right) \right] \\ & \stackrel{(iii)}{\leq} \sum_{\Omega \in \tilde{\Lambda}_I} \mathbb{E} \left[\Pr \left(\Omega^c \text{ cannot distinguish } (W_{I^c}, W'_I) \text{ and } (W_{I^c}, W_I) \middle| \begin{array}{c} (w_1, w_2) \text{ is sent} \\ \Omega \text{ can distinguish } (W_{I^c}, W'_I) \text{ and } (W_{I^c}, W_I) \end{array} \right) \right] \end{aligned}$$

where $\Lambda_I = \{\Omega : \mathcal{S}_I \subseteq \Omega, D \in \Omega^c\}$, $\tilde{\Lambda}_I = \{\Omega : \mathcal{S}_I \subseteq \Omega, \mathcal{N}(\mathcal{S}_I) \cup \{D\} \subseteq \Omega^c\}$, and $\mathcal{N}(\mathcal{S}_I)$ is the subset of the nodes in the network that are not in the flow of \mathcal{S}_I , *i.e.*, set of all nodes j such that there is no path from any source node in \mathcal{S}_I to j .

- (c). Argue that

$$\mathbb{E} \left[\Pr \left(\Omega^c \text{ cannot distinguish } (W_{I^c}, W'_I) \text{ and } (W_{I^c}, W_I) \middle| \begin{array}{c} (w_1, w_2) \text{ is sent} \\ \Omega \text{ can distinguish } (W_{I^c}, W'_I) \text{ and } (W_{I^c}, W_I) \end{array} \right) \right] \leq 2^{-TH(Y_{\Omega^c} | X_{\Omega^c})}.$$

- (d). Note that the inner term in each of the summations in RHS of (1), does not depend on W'_{I^c} . Count the number of terms appear in each summation.

- (e). By summarizing the above inequality, show that

$$\begin{aligned} \mathbb{E}[\Pr(\text{err})] & \leq |\tilde{\Lambda}_{\{S_1\}}| 2^{TR_1} 2^{-T \min_{\Omega \in \tilde{\Lambda}_{\{S_1\}}} H(Y_{\Omega^c} | X_{\Omega^c})} \\ & \quad + |\tilde{\Lambda}_{\{S_2\}}| 2^{TR_2} 2^{-T \min_{\Omega \in \tilde{\Lambda}_{\{S_2\}}} H(Y_{\Omega^c} | X_{\Omega^c})} \\ & \quad + |\tilde{\Lambda}_{\{S_1, S_2\}}| 2^{T(R_1+R_2)} 2^{-T \min_{\Omega \in \tilde{\Lambda}_{\{S_1, S_2\}}} H(Y_{\Omega^c} | X_{\Omega^c})}. \end{aligned}$$

- (f). Find constraints on (R_1, R_2) such that $\mathbb{E}[\Pr(\text{err})] \rightarrow 0$ as $T \rightarrow \infty$.

- (g). The constraints you found in part (f) involve $\tilde{\Lambda}_I$. In this part we show that $\tilde{\Lambda}_I$ can be replaced by Λ_I by showing

$$\min_{\Omega \in \tilde{\Lambda}_I} H(Y_{\Omega^c} | X_{\Omega^c}) = \min_{\Omega \in \Lambda_I} H(Y_{\Omega^c} | X_{\Omega^c}) \quad (2)$$

for any distribution $p(x_i)$ and all I . First argue that the RHS of (2) does not exceed the LHS. Now, assume that Ω^* is the minimizer of the RHS and $\Omega^* \cap \mathcal{N}(I) \neq \emptyset$. Define $\Omega' = \Omega^* \setminus \mathcal{N}(I)$. Show that $\Omega' \in \Lambda_I$ and it also minimizes the RHS by justifying the following inequalities.

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

Problem 3 (Typicality in deterministic channels)

We define robust typicality as follows.

We define $\underline{x} \in T_\delta$ if

$$|\mathbf{v}_{\underline{x}}(x) - p(x)| \leq \delta p(x), \quad \forall x \in \mathcal{X}$$

where $\mathbf{v}_{\underline{x}}(x) = \frac{1}{T} |\{t : x_t = x\}|$, is the empirical frequency.

We define robust joint typicality in the natural way as follows.

We define $(\underline{x}, \underline{y}) \in T_\delta$ if

$$|\mathbf{v}_{\underline{x}, \underline{y}}(x, y) - p(x, y)| \leq \delta p(x, y), \quad \forall (x, y) \in \mathcal{X} \times \mathcal{Y}$$

where $\mathbf{v}_{\underline{x}, \underline{y}}(x, y) = \frac{1}{T} |\{t : (x_t, y_t) = (x, y)\}|$, is the joint empirical frequency.

Suppose we have a deterministic channel, $Y = f(X)$.

(a). Show that if $\underline{x} \in T_\delta$ and

$$\underline{y} = [f(x_1), \dots, f(x_n)]$$

then $(\underline{x}, \underline{y}) \in T_\delta$ and $\underline{y} \in T_\delta$.

(b). For $Y = f(X)$, *i.e.*, if

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

Show that if $(\underline{x}, \underline{y}) \in T_\delta$ it implies that $\underline{x} \in T_\delta$ and $\underline{y} = [f(x_1), \dots, f(x_n)]$.