## Lecture 4: Multiple Access Channels

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## **1** Discrete Multiple Access Channel

**Definition 1** (Discrete Memoryless Multiple Access Channel (DM-MAC)). A DM-MAC is defined by a tuple  $(\mathcal{X}_1, \mathcal{X}_2, \mathcal{Y}, P_{Y|X_1X_2})$ , where  $\mathcal{X}_1$  and  $\mathcal{X}_2$  are finite input alphabets,  $\mathcal{Y}$  is a finite output alphabet and  $P_{Y|X_1X_2}$  is a conditional probability mass function such that for all  $(x_1, x_2, y) \in \mathcal{X}_1 \times \mathcal{X}_2 \times \mathcal{Y}$ ,  $P_{X_1X_2Y}(x_1, x_2, y) = P_{Y|X_1X_2}(y|x_1, x_2)P_{X_1}(x_1)P_{X_2}(x_2)$ , where  $P_{X_1}(x_1)$  and  $P_{X_2}(x_2)$  are the corresponding probabilities of the input symbols  $x_1$  and  $x_2$ .

**Definition 2** (Codes for Multiple Access Channel (DM-MAC)). A code for a DM-MAC  $(\mathcal{X}_1, \mathcal{X}_2, \mathcal{Y}, P_{Y|X_1X_2})$  is determined by the tuple  $(M_1, M_2, n, f_1^{(n)}, f_2^{(n)}, \phi^{(n)})$ , where  $f_1^{(n)}$  and  $f_2^{(n)}$  are the encoding functions, such that for all  $i \in \{1, 2\}$ ,

$$f_i^{(n)} \colon \{1, 2, \dots, M_i\} \to \mathcal{X}_i^n \tag{1}$$

and  $\phi^{(n)}$  is the decoding function, such that

$$\phi^{(n)}: \mathcal{Y}^n \to \{1, 2, \dots, M_1\} \times \{1, 2, \dots, M_2\}.$$
 (2)

**Definition 3** (Rate of a Code in the DM-MAC). The rate pair  $(R_1, R_2)$  associated to a code  $(M_1, M_2, n, f_1^{(n)}, f_2^{(n)}, \phi^{(n)})$  satisfies for all  $i \in \{1, 2\}$ ,

$$R_i = \frac{\log_2(M_i)}{n}.$$
(3)

**Definition 4** (Probability of Error). The probability of error  $P_{e}^{(n)}$  of a code  $\left(M_{1}, M_{2}, n, f_{1}^{(n)}, f_{2}^{(n)}, \phi^{(n)}\right)$  in the DM-MAC  $\left(\mathcal{X}_{1}, \mathcal{X}_{2}, \mathcal{Y}, P_{Y|X_{1}X_{2}}\right)$  is

$$P_{\mathbf{e}} = \Pr\left[\phi^{(n)}\left(\mathbf{Y}\right) \neq (W_1, W_2)\right]. \tag{4}$$

**Definition 5** (Achievable Rates). The pair  $(R_1, R_2) \in \mathbb{R}^2_+$  is achievable in the DM-MAC  $(\mathcal{X}_1, \mathcal{X}_2, \mathcal{Y}, P_{Y|X_1X_2})$  if there exists a sequence of tuples  $\{(2^{nR_1}, 2^{nR_2}, n, f_1^{(n)}, f_2^{(n)}, \phi^{(n)})\}_{n=1}^{\infty}$  such that the error probability tends to zero as the blocklength n tends to infinity. That is,

$$\limsup_{n \to \infty} P_{\mathbf{e}}^{(n)} = 0.$$
 (5)

**Definition 6** (Capacity Region of the DM-MAC). The information capacity region  $C_{\text{DM-MAC}} \subseteq \mathbb{R}^2_+$  of the DM-MAC  $(\mathcal{X}_1, \mathcal{X}_2, \mathcal{Y}, P_{Y|X_1X_2})$  is the closure of all achievable information rate pairs  $(R_1, R_2)$ .

**Theorem 1** (Capacity Region of the DM-MAC). The capacity region  $C_{\text{DM-MAC}}$  of the DM-MAC  $(\mathcal{X}_1, \mathcal{X}_2, \mathcal{Y}, P_{Y|X_1X_2})$  is

$$R_1 < I(X_1; Y | X_2),$$
 (6)

$$R_2 < I(X_2; Y|X_1), and$$
 (7)

$$R_1 + R_2 < I(X_1, X_2; Y),$$
 (8)

for some  $P_{X_1} \in \triangle(\mathcal{X}_1)$  and  $P_{X_2} \in \triangle(\mathcal{X}_2)$ .

**Theorem 2** (Capacity Region of the DM-MAC). The capacity region  $C_{\text{DM-MAC}}$  of the DM-MAC  $(\mathcal{X}_1, \mathcal{X}_2, \mathcal{Y}, P_{Y|X_1X_2})$  is a convex set of  $\mathbb{R}^2$ .

**Theorem 3** (Achievable Region of the DM-MAC). The achievable region  $\underline{C}_{DM-MAC}$  of the DM-MAC  $(\mathcal{X}_1, \mathcal{X}_2, \mathcal{Y}, P_{Y|X_1X_2})$  is

$$R_1 < I(X_1; Y | X_2, Q),$$
 (9)

$$R_2 < I(X_2; Y | X_1, Q), and$$
 (10)

$$R_1 + R_2 < I(X_1, X_2; Y|Q),$$
 (11)

for some joint distribution  $P_{QX_1X_2} = P_Q P_{X_2|Q} P_{X_2|Q}$  and  $|Q| \leq 4$ .

**Theorem 4** (Converse Region of the DM-MAC). The converse region  $\overline{C}_{DM-MAC}$  of the DM-MAC  $(\mathcal{X}_1, \mathcal{X}_2, \mathcal{Y}, P_{Y|X_1X_2})$  is

$$R_1 < I(X_1; Y | X_2, Q),$$
 (12)

$$R_2 < I(X_2; Y | X_1, Q), and$$
 (13)

$$R_1 + R_2 < I(X_1, X_2; Y|Q),$$
 (14)

for some joint distribution  $P_{QX_1X_2} = P_Q P_{X_2|Q} P_{X_2|Q}$  and  $|Q| \leq 4$ .

**Corollary 1** (Capacity Region of the DM-MAC). The achievable region  $\underline{C}_{DM-MAC}$ , the converse region  $\overline{C}_{DM-MAC}$  and the capacity region  $C_{DM-MAC}$  of the DM-MAC satisfy

$$\underline{\mathcal{C}}_{\mathrm{DM-MAC}} = \mathcal{C}_{\mathrm{DM-MAC}} = \mathcal{C}_{\mathrm{DM-MAC}}.$$
 (15)

## 2 Gaussian Multiple Access Channel

Figure 1: Two-User Discret Memoryless Multiple Access Channel

Consider the two-user memoryless Gaussian MAC (G-MAC) in Fig. 1. The goal of the communication is to convey the independent messages  $M_1$  and  $M_2$  from transmitters I and 2 to the common receiver using a blocklength of  $n \in \mathbb{N}$  channel uses. The messages  $W_1$  and  $W_2$  are mutually independent and uniformly distributed over the sets  $\mathcal{W}_1 \triangleq \{1, \ldots, 2^{nR_1}\}$  and  $\mathcal{W}_2 \triangleq \{1, 2, \ldots, 2^{nR_2}\}$ , where  $R_1 = \frac{\log_2 |\mathcal{W}_1|}{n}$  and  $R_2 = \frac{\log_2 |\mathcal{W}_2|}{n}$  denote the transmission rates. For all  $i \in \{1, 2\}$ , the channel inputs  $X_i = (X_{i,1}, X_{i,2}, \ldots, X_{i,n}) \in \mathbb{R}^n$  are generated at the beginning of the transmission using the encoding functions

$$f_i^{(n)} \colon \mathcal{W}_i \to \mathbb{R}^n. \tag{16}$$

That is  $X_i = f_i^{(n)}(W_i)$  and moreover, the channel inputs  $X_{i,1}, X_{i,2}, \ldots, X_{i,n}$  satisfy an expected average *input power constraint*:

$$\frac{1}{n}\sum_{t=1}^{n} \mathbb{E}\left[X_{i,t}^{2}\right] \leqslant P_{i},\tag{17}$$

where  $P_i$  denotes the energy rate constraint of transmitter *i*. For all  $t \in \{1, 2, ..., n\}$ , the channel inputs  $X_{1,t}$  and  $X_{2,t}$  generate the channel output  $Y_t$  according to a probability density function  $f_{Y|X_1X_2}$  for which the following holds

$$Y_t = h_1 X_{1,t} + h_2 X_{2,t} + Z_t, (18)$$

where  $h_1$  and  $h_2$  are the corresponding constant non-negative channel coefficients from transmitter *i* to the receiver. The noise terms  $Z_1, Z_2, \ldots, Z_n$  are zero-mean unitvariance real Gaussian random variables.

The receiver produces an estimate  $(\hat{W}_1^{(n)}, \hat{W}_2^{(n)}) = \Phi^{(n)}(\mathbf{Y})$  of the message-pair  $(W_1, W_2)$  via a decoding function  $\Phi^{(n)} \colon \mathbb{R}^n \to \mathcal{W}_1 \times \mathcal{W}_2$ , and the average probability of error is

$$P_{\mathbf{e}}^{(n)} = \Pr\left[(\hat{W}_1, \hat{W}_2) \neq (W_1, W_2)\right].$$
(19)

The G-MAC is fully described by two parameters: the signal to noise ratios with respect to transmitter 1, denoted by  $SNR_1$  and to transmitter 2, denoted by  $SNR_2$ . Note that given that the variance of the noise is 1, these parameters are defined as follows:

$$SNR_1 = P_1$$
 and (20)

$$SNR_2 = P_2.$$
 (21)

**Theorem 5** (Capacity Region of the G-MAC). The capacity region  $C_{G-MAC}$  of the G-MAC with parameters  $SNR_1$  and  $SNR_2$  is

$$R_1 < \frac{1}{2} \log (1 + \mathrm{SNR}_1),$$
 (22)

$$R_2 < \frac{1}{2}\log(1 + \mathrm{SNR}_2), \text{ and}$$
 (23)

$$R_1 + R_2 < \frac{1}{2} \log \left( 1 + \text{SNR}_1 + \text{SNR}_2 \right).$$
 (24)