#### EE7330: Network Information Theory

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# Lecture Notes 2: Convexity

Instructor: Shashank Vatedka Scribe: Ritesh Kumar

**Disclaimer**: These notes have not been subjected to the usual scrutiny reserved for formal publications. Please email the course instructor in case of any errors.

## 2.1 Open set

A set  $\mathcal{A}$  is said to be an open set if for every  $\underline{x} \in \mathbb{R}^n$ , we can find  $\epsilon > 0$  such that,

$$B_{\epsilon}(\underline{x}) \subseteq \mathcal{A} \tag{2.1}$$

Where  $B_{\epsilon}(\underline{x})$  is an n-dimensional ball with radius  $\epsilon$  and center  $\underline{x}$ . No matter which point we are taking, it always lies within the set, open sets are typically denoted as  $(\mathcal{A})$ . In general we denote open interval suing (). For example (a,b) is open interval and  $(4,5) \cup (6,8)$  is an open set.

## 2.2 Closed set

A set A is said to be closed set if:

- If compliment of the set is an open set.
- If every limit point of the set lies inside the set.
- Close interval denoted as [] and  $[a,b] \cup [p,q]$  is closed set. But  $(a,b) \cup [p,q]$  is neither closed set neither open set.

## 2.3 Convex set

 $\mathcal{A}$  is said to be a convex set if,  $\forall \ \underline{x}, \underline{y} \in \mathcal{A}$  and  $\forall \ 0 \leq \alpha \leq 1$ 

$$\alpha \underline{x} + (1 - \alpha)y \in \mathcal{A} \tag{2.2}$$

Geometrically, if we consider any two points inside the set, then every point that lies on the line that joins these two points must lie inside the set.

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Figure 2.1: (a)Convex set

(b) Not a convex set

**Example:** Achievable rate regions are convex: a time-sharing argument: Consider a multi-user system where we have some L-transmitter and L-receiver and they want to communicate through the channel described in figure 2.2. We want to find the achievable rate over this channel.

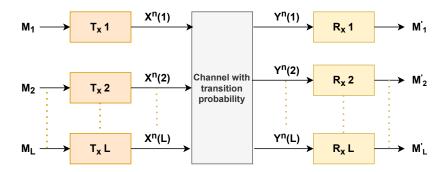


Figure 2.2: (a) Transceiver for L-transmitter and Receiver setting

A rate tuple  $(R_1, R_2 ... R_L)$  is achievable if  $\exists$  pair of encoder and decoder.  $(ENC_n(1), ENC_n(2), ... ENC_n(L), DEC_n(1), DEC_n(2) ..., DEC_n(L))$  pair can be define as :  $ENC_n(i) : \{0,1\}^{K_n(i)} \longrightarrow \mathcal{X}^n(l)$  and  $DEC_n(i) : \mathcal{Y}^n(l) \longrightarrow \{0,1\}^{K_n(l)}$  such that :

$$\lim_{n \to \infty} \sup_{n \to \infty} \frac{K_n(i)}{n} = R_i \tag{2.3}$$

$$\limsup_{n \to \infty} Pr\left[M_i' \neq M_i\right] = P_e = 0 \tag{2.4}$$

And we want to find what kind of rate can be achievable. Consider the tuple  $(R_1, R_2)$  i.e (L = 2) which is achievable. For single  $T_X$  -  $R_X$  setting (i.e L = 1), according to the Shannon coding theorem, any rate less than the capacity of the channel is achievable. So, rate region for the single  $T_X - R_X$  is a closed interval which is [0, C].

#### For 2 user Case:

**Theorm :** Every rate region are convex. i.e if  $(R_1, R_2) \& (R'_1, R'_2)$  are achievable and it lie in the convex set then  $\forall 0 \le \alpha \le 1$ ,  $(\alpha R_1 + (1 - \alpha)R'_1, \alpha R'_2 + (1 - \alpha)R_2)$  is achievable.

**Proof :** Consider a time sharing argument  $(R_1, R_2)$  and  $(R'_1, R'_2)$  for ENC and DEC pair such that :  $K_n(1) \ge n (R_1 - \epsilon)$  and  $K_n(2) \ge n (R_2 - \epsilon)$  for (ENC, DEC).  $K'_n(1) \ge n (R'_1 - \epsilon)$  and  $K'_n(2) \ge n (R'_2 - \epsilon)$  for (ENC', DEC')

We have n-channel uses and consider  $\alpha$ n times channel used by first (ENC - DEC) pair and  $(1 - \alpha)n$  for (ENC' - DEC') pair. Hence

$$\alpha K_n(1) \geqslant \alpha n \left( R_1 - \epsilon \right) \tag{2.5}$$

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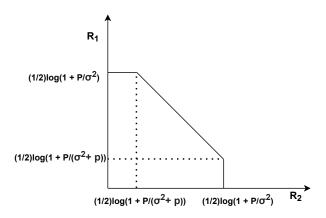


Figure 2.3: 2 user rate tuple case ( 2 user Gaussian multiple-access channel (MAC) only)

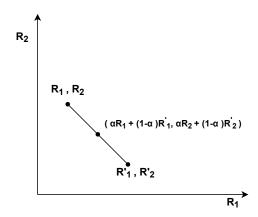


Figure 2.4: 2 user rate case

$$\alpha K_n(2) \geqslant \alpha n \left( R_2 - \epsilon \right) \tag{2.6}$$

and,

$$(1 - \alpha) K'_n(1) \geqslant (1 - \alpha) n \left( R'_1 - \epsilon \right) \tag{2.7}$$

$$(1 - \alpha) K_n'(2) \geqslant (1 - \alpha) n \left( R_2' - \epsilon \right) \tag{2.8}$$

Combining (2.5), (2.7) and (2.6), (2.8), we have

$$\alpha K_n(1) + (1 - \alpha) K'_n(1) \geqslant \alpha n \left( R_1 - \epsilon \right) + (1 - \alpha) n \left( R'_1 - \epsilon \right) \tag{2.9}$$

$$\alpha K_n(2) + (1 - \alpha) K'_n(2) \geqslant \alpha n \left( R_2 - \epsilon \right) + (1 - \alpha) n \left( R'_2 - \epsilon \right)$$

$$(2.10)$$

And it holds  $\forall \epsilon > 0$ .

Using Fano's inequality we can prove the equation (2.4). We can have,

$$K_n \leq H_2(P_e) + P_e \log_2(2^{M^{K_n}}) + I(M^{K_n} : M'^{K_n})$$
 (2.11)

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$$\xrightarrow{\mathsf{M}^{\mathsf{K}_{\mathsf{n}}}} \mathsf{ENC}_{\mathsf{n}} \xrightarrow{\mathsf{X}^{\mathsf{n}}} \mathsf{P}_{\mathsf{Y}|\mathsf{X}} \xrightarrow{\mathsf{Y}^{\mathsf{n}}} \mathsf{DEC}_{\mathsf{n}} \xrightarrow{\mathsf{M}^{\mathsf{r}\mathsf{K}_{\mathsf{n}}}}$$

Figure 2.5: Single user transmission system

Where, 
$$P_e = \limsup_{n \to \infty} Pr\left[M_i' \neq M_i\right]$$

$$K_n \leq H_2(P_e) + P_e log_2(2^{K_n}) + I(X^n : Y^n)$$
 (using data processing inequality) (2.12)

$$K_n \le H_2(P_e) + P_e K_n + I(X^n : Y^n)$$
 (2.13)

$$K_n \leqslant H_2(P_e) + P_e K_n + nC \Rightarrow \frac{H_2(P_e)}{n} \geqslant \frac{K_n(1 - P_e)}{n} - C$$

$$\lim_{n \to \infty} \left( \frac{K_n(1 - P_e)}{n} - C \right) \leqslant \lim_{n \to \infty} \frac{H_2(P_e)}{n}$$

Let  $\frac{K_n}{n} = R$ ,

$$\lim_{n \to \infty} P_e \geqslant \frac{R - C}{R} \tag{2.14}$$

$$\lim_{n \to \infty} \sup P_e \geqslant \frac{R - C}{R} \tag{2.15}$$

If we operate at the rate more than capacity of the channel say R = C +  $\epsilon$ 

$$\limsup_{n \to \infty} P_e \geqslant \frac{\epsilon}{R} \tag{2.16}$$

That is probability or error is always non-zero. But if we operate the channel below the capacity and for large n, in (2.15) we can observe  $P_e$  approaches to 0.