

# The Wire-Tap Channel and Secrecy in Communication

Group 2

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# Outline

- 1 Formal Statement and Main Results
- 2 A Special Case
- 3 Direct Half of Main Theorem
- 4 Converse of Main Theorem

# The communication system

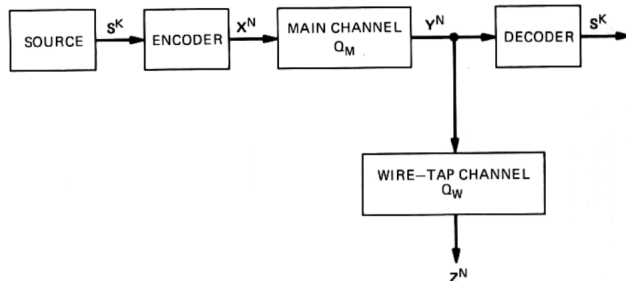


Figure: Wire-tap channel (from [1])

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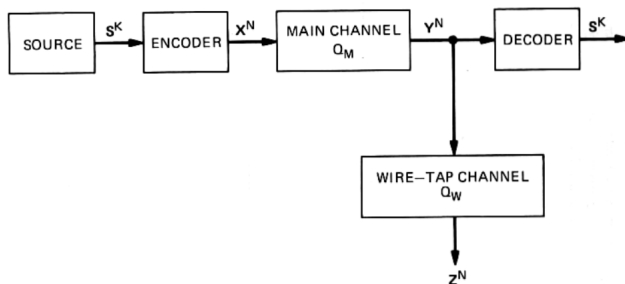


Figure: Wire-tap channel (from [1])

- **Source:**

Defined by  $\{S_k\}_1^\infty$  where  $S_k$  are iid random variables that take values in the finite set  $\mathcal{S}$ .

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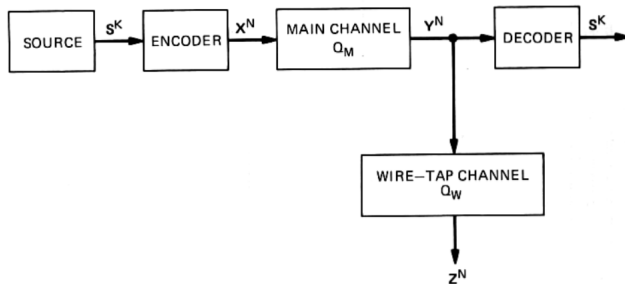


Figure: Wire-tap channel (from [1])

## ● Main channel:

Discrete memoryless channel with finite input alphabet  $\mathcal{X}$ , finite output alphabet  $\mathcal{Y}$  and transition probability  $Q_M(y|x)$ . Denote the channel capacity by  $C_M$ .

# The communication system

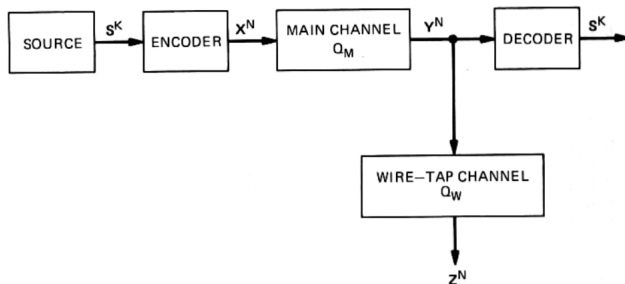


Figure: Wire-tap channel (from [1])

- **Wire-tap channel:**

Discrete memoryless channel with input alphabet  $\mathcal{Y}$ , finite output alphabet  $\mathcal{Z}$ , and transition probability  $Q_W(z|y)$ .

# Encoder-decoder $(K, N, \Delta, P_e)$

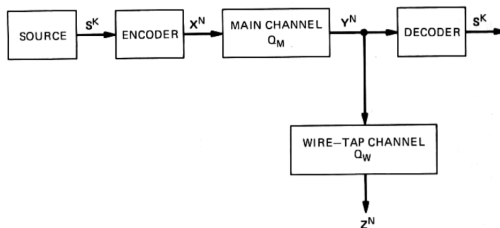


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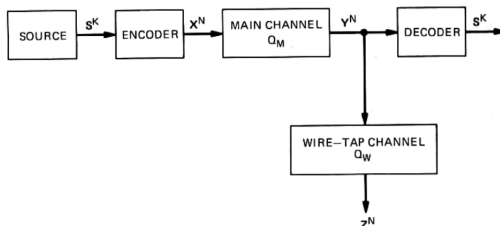


Figure: Wire-tap channel (from [1])

- **Encoder:**

A  $(K, N)$  encoder takes in  $s^K$  and outputs the random vector  $x^N$ .



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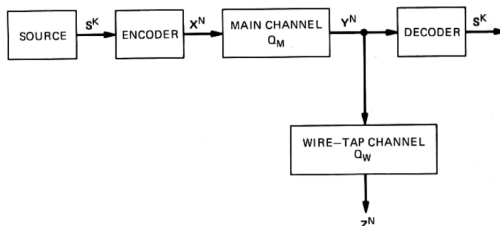


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- **Decoder:**

A mapping  $f_D : \mathcal{Y}^N \rightarrow \mathcal{S}^K$  such that  $\hat{\mathbf{S}} = f_D(\mathbf{Y})$ .

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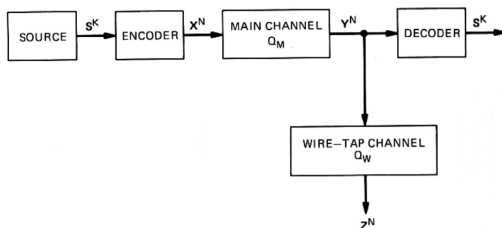


Figure: Wire-tap channel (from [1])

- For a given encoder and decoder, the **error rate** is

$$P_e = \frac{1}{K} \sum_{k=1}^K \mathbb{P}\{S_k \neq \hat{S}_k\}$$

## Encoder-decoder $(K, N, \Delta, P_e)$

- The **equivocation** of the source at the output of the wire-tap channel for a particular encoder is

$$\Delta \triangleq \frac{1}{K} H(\mathbf{S}^K | \mathbf{Z}^N)$$

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$$\Delta \leq h(P_{ew}) + P_{ew} \log |\mathcal{S}|$$

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Thus a large  $\Delta$  corresponds to a large  $P_{ew}$ .

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A pair  $(R, d)$  is **achievable** if,  $\forall \epsilon > 0$ ,  $\exists$  an encoder-decoder  $(N, K, \Delta, P_e)$  such that

$$\frac{H_S K}{N} \geq R - \epsilon \quad (1a)$$

$$\Delta \geq d - \epsilon \quad (1b)$$

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The problem is to characterize the set  $\mathcal{R}$  of achievable  $(R, d)$  pairs.

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- Let  $X \sim p_X(x), x \in \mathcal{X}$ .  
For  $R \geq 0$ , let  $\mathcal{P}(R) = \{p_X : I(X; Y) \geq R\}$ .

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For  $R \geq 0$ , let  $\mathcal{P}(R) = \{p_X : I(X; Y) \geq R\}$ .
- For  $0 \leq R \leq C_M$ , define

$$\Gamma(R) \triangleq \sup_{p_X \in \mathcal{P}(R)} I(X; Y|Z) \quad (2)$$

# The main theorem

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## Theorem 1

*The set  $\mathcal{R}$  defined above is equal to  $\overline{\mathcal{R}}$ , where*

$$\overline{\mathcal{R}} \triangleq \{(R, d) : 0 \leq R \leq C_M, 0 \leq d \leq H_S, Rd \leq H_S \Gamma(R)\} \quad (3)$$

# Setting for the special case

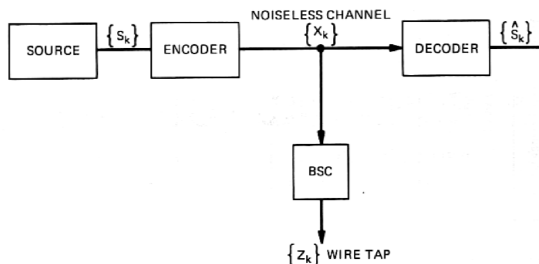


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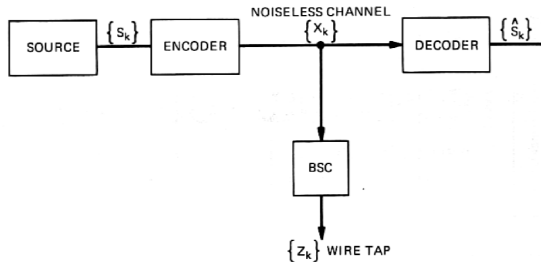


Figure: Wire-tap channel (from [1])

- **Source:**

A simple equiprobable binary source emitting a data sequence  $S_1, S_2, \dots$ , where  $\Pr\{S_i = 0\} = \Pr\{S_i = 1\} = \frac{1}{2}$  for all  $i$ .



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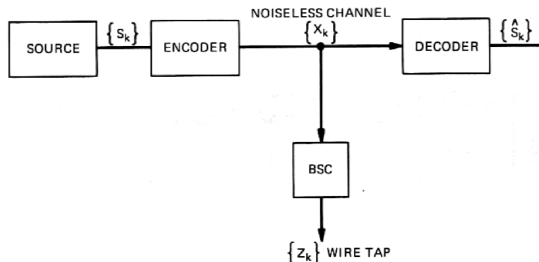


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- **Main channel:**

A perfectly noiseless channel, i.e.

$$Q_M(y|x) = \delta_{x,y}$$

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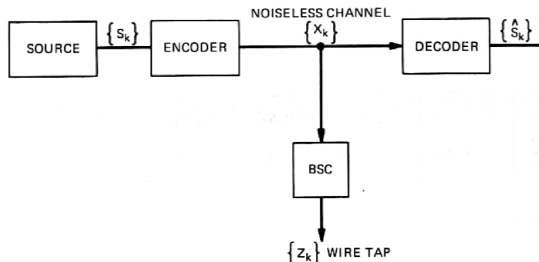


Figure: Wire-tap channel (from [1])

- **Wire-tap channel:**

A Binary Symmetric Channel with crossover probability  $p_0$ , i.e.

$$Q_W(z|y) = (1 - p_0)\delta_{y,z} + p_0(1 - \delta_{y,z})$$

# Theorem

For this example, observe that  $C_M = 1$ ,  $H_S = 1$ , and  $\Gamma(R) = h(p_0)$  for all  $0 \leq R \leq 1$ .

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## Theorem 2 (for the special case)

*The pair  $(R, d)$  is achievable if and only if*

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We first prove the converse, then the direct half.

$R \leq C_M$  being necessary is immediate from the converse to the ordinary coding theorem, and  $d \leq H_S$  follows directly from

$$\Delta = \frac{1}{K} H(\mathbf{S}^K | \mathbf{Z}^N) \leq \frac{1}{K} H(\mathbf{S}^K) = H_S$$



# Closer look at $\overline{\mathcal{R}}$

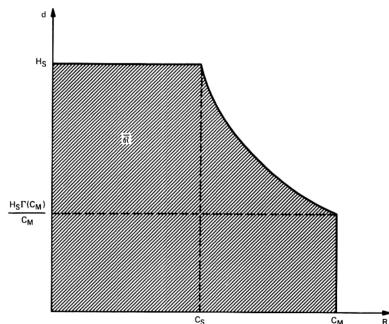


Figure:  $\overline{\mathcal{R}}$  when  $Q_M$  is noiseless and  $Q_W$  a BSC (from [1])

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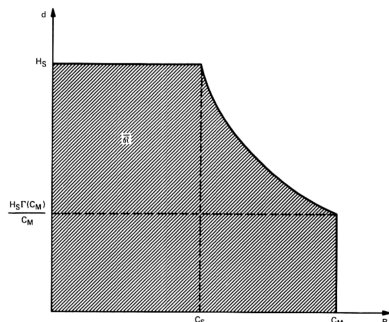


Figure:  $\overline{\mathcal{R}}$  when  $Q_M$  is noiseless and  $Q_W$  a BSC (from [1])

Note that  $\overline{\mathcal{R}}$  is not convex!



# Converse

$$\begin{aligned} K\Delta &= H(\mathbf{S}^K | \mathbf{Z}^N) = H(\mathbf{S}^K, \mathbf{Z}^N) - H(\mathbf{Z}^N) \\ &= H(\mathbf{S}^K, \mathbf{X}^N, \mathbf{Z}^N) - H(\mathbf{X}^N | \mathbf{S}^K, \mathbf{Z}^N) - H(\mathbf{Z}^N) \\ &= H(\mathbf{Z}^N | \mathbf{X}^N, \mathbf{S}^K) + H(\mathbf{X}^N, \mathbf{S}^K) - H(\mathbf{X}^N | \mathbf{S}^K, \mathbf{Z}^N) - H(\mathbf{Z}^N) \\ &= H(\mathbf{Z}^N | \mathbf{X}^N) + H(\mathbf{S}^K | \mathbf{X}^N) + H(\mathbf{X}^N) - H(\mathbf{X}^N | \mathbf{S}^K, \mathbf{Z}^N) - H(\mathbf{Z}^N) \end{aligned}$$

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Since  $(R, d)$  is achievable, this gives us

$$(R - \epsilon)[(d - \epsilon) - h(\epsilon)] \leq h(p_0) \text{ for any } \epsilon > 0$$



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$$\begin{aligned} (R - \epsilon)[(d - \epsilon) - h(\epsilon)] &\leq h(p_0) \text{ for any } \epsilon > 0 \\ \implies Rd &\leq h(p_0) \quad \square \end{aligned}$$

## Direct half

Now we show that any pair  $(R, d)$  satisfying

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is achievable, i.e. the existence of an encoder-decoder  $(K, N, \Delta, P_e)$  that gives arbitrarily low error and high equivocation.

## Direct half

Let  $G$  be a group code, with block length  $N$  and  $|G| = 2^{N-K}$ . This has  $2^K$  cosets,  $\{C_i\}_1^{2^K}$ . (with  $C_1 = G$ )

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- **Encoder:**

When  $\mathbf{S}^K = i$ , we let  $\mathbf{X}^N$  be any random member of the coset  $C_i$ , i.e.

$$\Pr\{\mathbf{X}^N = \mathbf{x} | \mathbf{S} = i\} = \begin{cases} \frac{1}{|G|} = 2^{-(N-K)} & \mathbf{x} \in C_i \\ 0 & \text{else} \end{cases}$$

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- **Decoder:**

Simply letting  $f_D(\mathbf{y}) = i$  when  $\mathbf{y} \in C_i$ , we have  $P_e = 0$ .

## Direct half

We use the following equality from the converse proof,

$$K\Delta = Nh(p_0) + H(\mathbf{S}^K|\mathbf{X}^N) + H(\mathbf{X}^N) - H(\mathbf{X}^N|\mathbf{S}^K, \mathbf{Z}^N) - H(\mathbf{Z}^N)$$



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$$\begin{aligned} \Rightarrow \Delta &\geq \frac{N}{K}h(p_0) - \frac{h(\lambda)}{K} - \lambda\left(\frac{N}{K} - 1\right) \\ &\geq d - \left[\frac{h(\lambda)}{K} + \lambda\left(\frac{1}{R} - 1\right)\right] \end{aligned}$$

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Since  $Rd \leq h(p_0)$  and  $R = \frac{K}{N}$ , we have

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Thus  $\lambda$  can be made arbitrarily small, and we have that  $\Delta \geq d - \epsilon$ , and the proof is complete. □



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is achievable, i.e.  $\overline{\mathcal{R}} \subseteq \mathcal{R}$ .

## Some preliminaries: Typical sequences

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Define the set of typical sequences as

$$T^*(N) = \left\{ \mathbf{x} \in \mathcal{X}^N : \left| \frac{\#(i, \mathbf{x})}{N} - p_X^*(i) \right| \leq N^{-\frac{1}{4}}, 1 \leq i \leq A \right\}$$

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### Lemma 4

Let  $X^N, Z^N$  correspond to an arbitrary encoder and let  $X^*, Z^*, T^*$  correspond to an arbitrary  $p_X^*$ . Then

$$\frac{1}{N} I(\mathbf{X}^N; \mathbf{Z}^N) \leq I(\mathbf{X}^*; \mathbf{Z}^*) + (\log A) \mathbb{P}\{\mathbf{X}^N \notin T^*(N)\} + f_1(N)$$

where  $f_1(N) \rightarrow 0$  as  $N \rightarrow \infty$ .

## Some preliminaries: Ad-hoc encoder-decoder

We now define an ad-hoc encoder-decoder to prove the direct half.



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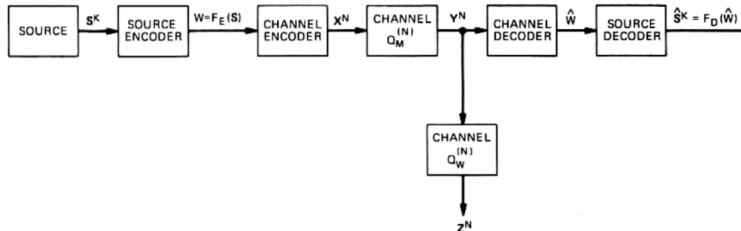


Figure: Ad-hoc encoder-decoder (from [1])

## Some preliminaries: Source encoder-decoder

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- Let the **source encoder** be the mapping  $F_E : \mathcal{S}^K \rightarrow \{1, \dots, M\}$  where

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- Let the **source decoder** be the mapping  $F_D : \{1, \dots, M\} \rightarrow \mathcal{S}^K$ . Let

$$P_{es}^{(K)} = \mathbb{P}\{F_D \circ F_E(S^K) \neq S^K\}$$

be the error probability.

## Some preliminaries: Channel encoder

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$$C_i = \{\mathbf{x}_{(i-1)M_2+1}, \dots, \mathbf{x}_{iM_2}\}, \quad 1 \leq i \leq M$$

- When  $W = i$ , the channel encoder output  $\mathbf{X}^N$  is a random uniformly chosen member of  $C_i$ , i.e,

$$\mathbb{P}\{\mathbf{X}^N = \mathbf{x}_{(i-1)M_2+j}\} = \frac{q_i}{M_2}, \quad 1 \leq i \leq M, 1 \leq j \leq M_2$$

where

$$q_i \triangleq \mathbb{P}\{W = F_E(S^K) = i\}$$



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- The source decoder outputs  $\hat{\mathbf{S}}^K = F_D(\hat{\mathbf{W}})$  where  $F_D$  is chosen as mentioned earlier.

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### Lemma 5

*For the ad-hoc encoder-decoder defined above*

$$I(\mathbf{X}^N; \mathbf{Z}^N | \mathbf{S}^K) \geq \log M_2 - h(\bar{\lambda}) - \bar{\lambda} \log M_2$$

## Some preliminaries

Combining all the above preliminary results,



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## Corollary 6

Let  $p_X^*$  be an arbitrary probability distribution on  $\mathcal{X}$  and  $X^*, Y^*, Z^*, T_X^*(N)$  be as defined earlier.  $\mathbf{S}^K, \mathbf{X}^N, \mathbf{Y}^N, \mathbf{Z}^N$  correspond to the ad-hoc encoder-decoder. Then

$$P_e \leq P_{es}^{(K)} + \lambda \quad (7a)$$

$$\begin{aligned} \frac{K}{N} \Delta \geq & \frac{K}{N} H_S + \frac{1}{N} \log M_2 - I(X^*; Z^*) - \frac{h(\bar{\lambda})}{N} - \frac{\bar{\lambda} \log M_2}{N} \\ & - (\log A) \mathbb{P}\{\mathbf{X}^N \notin T_X^*(N)\} - f_1(N) \end{aligned} \quad (7b)$$

where  $f_1(N) \rightarrow 0$  as  $N \rightarrow \infty$ .

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$$M_1 = 2^{N \left[ I(\mathbf{X}^*; \mathbf{Y}^*) - \frac{\epsilon R}{2H_S} \right]}$$

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$$M_1 = 2^{N \left[ I(\mathbf{X}^*; \mathbf{Y}^*) - \frac{\epsilon R}{2H_S} \right]}$$

and  $M_2$  follows as

$$M_2 = \frac{M_1}{M} = 2^{N \left[ I(\mathbf{X}^*; \mathbf{Y}^*) - \frac{\epsilon R}{2H_S} - \frac{K}{N} H_S - \frac{K}{N} H_S \delta_K \right]}$$

# Proof

Substituting for  $M_2$  in (7b) yields

$$\frac{R\Delta}{H_S} \geq \Gamma(R) - f_2(N)$$

where

$$f_2(N) = \frac{\epsilon R}{2H_S} + \frac{h(\bar{\lambda})}{N} + \frac{\bar{\lambda} \log M_2}{N} + (\log A) \mathbb{P}\{\mathbf{X}^N \notin \mathcal{T}^*(N)\} + f_1(N)$$



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Note that we have

$$Rd \leq H_S \Gamma(R)$$

Hence it is enough to bound  $f_2(N)$ .

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Similarly, we have

$$P_e \leq P_{es}^{(K)} + \lambda$$

Hence it is enough to bound  $\lambda$ .

## Proof: Choosing $\{\mathbf{x}_m\}_1^{M_1}$

The following lemma gives us the existence of a  $\{\mathbf{x}_m\}_1^{M_1}$  such that  $\lambda$  and  $f_2(N)$  are small.

## Proof: Choosing $\{\mathbf{x}_m\}_1^{M_1}$

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### Lemma 7

*With  $p_X^*$ ,  $M_1$ ,  $M_2$  as given above, there exists for arbitrary  $N$  a set  $\{\mathbf{x}_m\}_1^{M_1}$  such that*

$$f_3(N) \geq \begin{cases} \mathbb{P}\{\mathbf{X}^N \notin T^*(N)\}, \\ \lambda, \\ \bar{\lambda} \end{cases}$$

*where  $f_3(N) \rightarrow 0$  as  $N \rightarrow \infty$ .*

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Thus, any  $(R, d) \in \overline{\mathcal{R}}$  is achievable.





## Converse of main theorem

We now show that the family of achievable rates  $\mathcal{R}$  is contained in  $\overline{\mathcal{R}}$ , i.e. every achievable  $(R, d)$  satisfies

$$0 \leq R \leq C_M, \quad 0 \leq d \leq H_S, \quad Rd \leq H_S \Gamma(R) \quad (8)$$

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We now prove the third part.

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- For  $x \in \mathcal{X}$ ,  $p_{n,\mathbf{y}}(x) = \Pr\{X_n = x | \mathbf{Y}^{n-1} = \mathbf{y}\}$

It is clear that  $p_{n,\mathbf{y}} \in \mathcal{P}[\alpha_n(\mathbf{y})]$  and hence

$$\Gamma[\alpha_n(\mathbf{y})] \geq I(X_n; Y_n | Z_n, \mathbf{Y}^{n-1} = \mathbf{y})$$

# Proof

We now make use of the following lemma to relate  $\Delta$  with  $H_S$ .

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## Lemma 8

For any encoder-decoder  $(N, K, \Delta, P_e)$ ,

$$\frac{K}{N}[\Delta - \delta(P_e)] \leq \frac{1}{N} \sum_{n=1}^N I(X_n; Y_n | Z_n, \mathbf{Y}^{n-1}), \quad (10a)$$

$$\frac{K}{N}[H_S - \delta(P_e)] \leq \frac{1}{N} \sum_{n=1}^N I(X_n; Y_n | \mathbf{Y}^{n-1}) \quad (10b)$$



# Proof

$$\frac{K}{N}[\Delta - \delta(P_e)] \leq \frac{1}{N} \sum_{n=1}^N I(X_n; Y_n | Z_n, \mathbf{Y}^{n-1})$$

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# Proof

Hence we have

$$\frac{H_S K}{N} [\Delta - \delta(P_e)] \leq H_S \Gamma \left( \frac{K}{N} [H_S - \delta(P_e)] \right)$$

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$$(R - \epsilon)[(d - \epsilon) - \delta(\epsilon)] \leq H_S \Gamma \left( R - \frac{K}{N} \delta(\epsilon) \right) \text{ for any } \epsilon > 0$$



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Hence we have

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# References

- [1] A. D. Wyner.  
The wire-tap channel.  
*Bell System Technical Journal*, 54(8):1355–1387, 1975.
- [2] Abbas El Gamal and Young-Han Kim.  
*Network information theory*.  
Cambridge University Press, 2018.
- [3] Robert Gallager.  
*Information theory and reliable communication*.  
John Wiley, 1968.

Thank you