

LECTURE 06

2/14/19

TODAY:"CHANNEL CODING"

- DEFINITIONS
- BINARY SYMMETRIC CHANNEL
- GENERAL CHANNELS

Next few lectures: Error-Correction (with "random errors")

General Channel of Communication (Memoryless)



- Given by $P_{Y|X}$ given by $\Omega_x \times \Omega_y$ matrix
- $P_{Y|X}(\alpha, \beta) = P[Y=\beta | X=\alpha]$.

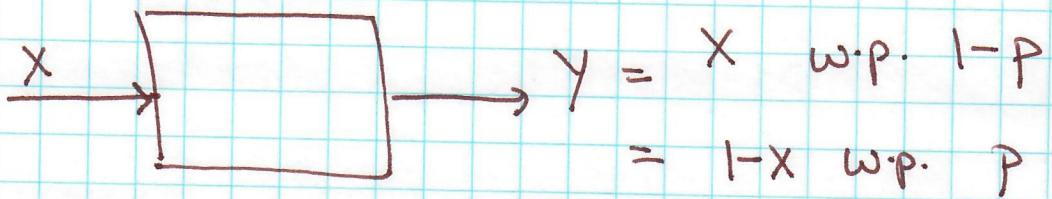
n i.i.d. uses of channel.

How much "information" per use?

(2)

Simple Example

BSC(p) [Binary Symmetric Channel]



"Capacity": rate at which information can be pushed through

Formally: encoding + decoding functions E_n, D_n achieve rate R if ~~no~~ with error ϵ

$$\textcircled{1} \quad E_n: \{0,1\}^{Rn} \rightarrow \mathcal{R}_x^n$$

$$\textcircled{2} \quad D_n: \mathcal{R}_y^n \rightarrow \{0,1\}^{Rn}$$

$$\textcircled{3} \quad \Pr[\text{Decoding failure}] = \Pr_{m \in \text{Unif}(\{0,1\}^{Rn})} [D(Y) \neq m] \quad Y \sim P_{Y|X=E(m)}$$

Capacity of channel $P_{Y|X}$

$$= \sup_R \left\{ \lim_{\epsilon \rightarrow 0} \lim_{n \rightarrow \infty} \left\{ \exists E_n, D_n \text{ of rate } R \right\} \right\} \text{ over } G$$

③ Connections to Information Theory

$$\textcircled{1} \quad \text{Capacity } (P_{Y|X}) = \max_{P_X} \{ I(X; Y) \} !$$

[operational view of Information].

[Theorem ... to be proved later].

② Information Theory gives "best" algorithms + codes !!

Today ①

— X —

Special Case : BSC(p)

$$\text{Capacity} = 1 - h(p) \quad [P_X = \text{Bern}(\frac{1}{2})].$$

$$= H(\text{Bern}(\frac{1}{2})) - I(Y|X)$$

$$= 1 - H(\text{Bern}(p))$$

$$= 1 - h(p).$$

— X —

Proof of ① for BSC(p).

$$\textcircled{1} \quad R \geq 1 - h(p) - \epsilon. \quad ; \quad k = (R \cdot n)$$

Pick $E_n: \{0,1\}^k \rightarrow \{0,1\}^n$ at random

$D_n = \text{max. likelihood decoding.}$

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Lemma:

$$\Pr_{\substack{Y \\ E_n, m, Y \in E_n(m)}} \left[D_n(Y) \neq m \right] \leq \epsilon$$

$$\left[\Rightarrow \Pr_{\substack{\exists E_n \\ m, Y \in E_n(m)}} \left[\dots \right] \leq \epsilon \right]$$

Proof: Error events

$$\textcircled{1} \quad \Pr_{\substack{Y \\ Y \in E_n(m)}} \left[\Delta(Y, E_n(m)) \geq (p + \epsilon)n \right] \leq \exp(-\epsilon^2 n)$$

(E1) ↑
Chernoff Bounds

$$\textcircled{2} \quad \Pr_{\substack{E'_n \\ \exists m' \neq m; \\ \Delta(E'_n(m'), Y) \leq (p + \epsilon)n}} \left[\exists m' \neq m; \Delta(E'_n(m'), Y) \leq (p + \epsilon)n \right]$$

$$\leq 2^k \cdot \binom{n}{(p+\epsilon)n} \cdot \frac{1}{2^n}$$

$$\approx 2^k \cdot 2^{H(p) \cdot n} \cdot 2^{-n}$$

$$= \approx \exp(-\epsilon n) \cdot \otimes$$

if (E1) or (E2) don't happen then decoding right

⊗

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General Channel

Fix P_x

- Pick $E_n: \{0,1\}^{R_n}$ $\rightarrow \mathcal{Z}_x^n$

by picking $E_n(m); q_i \sim P_x$ ind. for all (m, i) .

- Decoding (y)

if \exists unique $\cancel{X \in \mathcal{Z}_x^n}$ $m \in \{0,1\}^{R_n}$ s.t. for $x = E(m)$

s.t. ① X is P_x^n typical

$$[\Pr[X] \approx \frac{1}{2^{-H(P_x)n}}]$$

& ② (X, Y) is P_{XY}^n typical

$$\Pr[X, Y] \approx \frac{1}{2^{H(P_{XY})n}}$$

Output m .

else error.

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Analysis

(1) Two types of errors

(E1) X not typical, Y not typical, (X, Y) not jointly typical

(E2) $(E(m'), Y)$ jointly typical, ~~for some $m' \neq m$~~

— X —

(1) $\Pr[E_1] \rightarrow 0$ by AEP

(2) E2? Key + useful lemma.

— X —

Lemma: Let P, Q be distributions over Σ^* .

$$\Pr_{\substack{\exists \tilde{z} \text{ typical for } Q^n \\ \tilde{z} \sim P^n}} [\exists z \text{ typical for } Q^n] \leq 2^{-D(Q||P) \cdot n}$$

— X —

In our case

$(E(m'), Y)$ drawn from $P_x^n \times P_y^n$

$\Pr[(E(m'), Y) \text{ typical for } P_{xy}^n]$

$$\leq 2^{-D(P_{xy} || P_x \times P_y) \cdot n}$$

$$= 2^{-I(Y; X) \cdot n}$$

⇒ Can take union bound over $2^{m'}$'s. 7

⇒ Rate $\geq I(x; y)$!

Can optimize over P_x

to get

$$\text{Capacity} \geq \sup_{P_x} \{I(x; y)\}$$

————— X —————

Converse Coding Theorem:

V1: if $\Pr[\text{decoding failure}] \rightarrow 0$ then ~~Rate ≤ Capacity~~

$$\text{Capacity} \\ \text{Rate} \leq \sup_{P_x} \{I(x; y)\}$$

V2: for BSC(p): if $\text{Rate} = \sup \{I(x; y)\} + \epsilon$ then

$$\Pr[\text{decoding failure}] \geq 1 - \exp(-n).$$

[V2 much stronger quantitatively; but ~~too~~ being shown
only for BSC(p).]

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Proof of V1 : (uses Fano's Inequality)

have $m \rightarrow X^n \rightarrow Y^n \rightarrow \hat{m}$ - a Markov Chain.

$$① I(X^n; Y^n) \leq \cancel{nR} \cdot n \cdot \sup_{P_X} \{ I(X; Y) \}$$

$$\begin{aligned} ② H(m) &= nR \stackrel{\text{def}}{=} H(m|\hat{m}) + I(m; \hat{m}) \\ &\leq H(m|\hat{m}) + I(X^n; Y^n) \quad [\text{DPI}] \\ &\leq H(m|\hat{m}) + nC \end{aligned}$$

need to bound $H(m|\hat{m})$

$$\begin{aligned} \text{Fano: } H(m|\hat{m}) &\leq h(\Pr[m \neq \hat{m}]) + \Pr[m \neq \hat{m}] \cdot nR \\ &\leq 1 + o(nR) \end{aligned}$$

$$\Rightarrow nR(1-o(1)) \leq nC$$

$$R(1-o(1)) \leq C$$

$$\Rightarrow R \leq C \quad \text{in the limit}$$

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V2: Exercise / Ret.