

EE 231A Information Theory
Lecture 9
Converse to the Channel Coding Theorem

- Fano's Inequality
- Considering possibly dependent inputs
- Proof of the converse

Part A: Fano's Inequality

Fano's inequality

- Stated in section 7.9, but proof in section 2.10.
- For a DMC and a $(2^{nR}, n)$ code with input messages uniformly distributed.

$$H(W | \hat{W}) \leq 1 + P(\hat{W} \neq W)nR$$

An upper bound on $H(W | \hat{W})$

$$E = \begin{cases} 1 & \text{if } \hat{W} \neq W \\ 0 & \text{if } \hat{W} = W \end{cases} \quad \begin{aligned} H(E, W | \hat{W}) &= H(W | \hat{W}) + H(E | W, \hat{W}) \\ &= H(E | \hat{W}) + H(W | E, \hat{W}) \end{aligned}$$

$$\begin{aligned} H(W | \hat{W}) &= \underbrace{H(E | \hat{W})}_{\leq 1} + H(W | E, \hat{W}) - \cancel{H(E | W, \hat{W})} \rightarrow 0 \\ &\leq 1 + H(W | E, \hat{W}) \end{aligned}$$

$$H(W | \hat{W}) \leq 1 + H(W | E, \hat{W})$$

An upper bound on $H(W | E, \hat{W})$

$$\begin{aligned}
 H(W | E, \hat{W}) &= P(E=0) \cancel{H(W | \hat{W}, E=0)}^0 \\
 &\quad + P(E=1)H(W | \hat{W}, E=1) \\
 &= P(\hat{W} \neq W)H(W | \hat{W}, E=1) \\
 &\leq P(\hat{W} \neq W) \log |\mathcal{W}| \\
 &= P(\hat{W} \neq W)nR
 \end{aligned}$$

Thus $H(W | \hat{W}) \leq 1 + P(\hat{W} \neq W)nR$

Part B: Considering Possibly Dependent Inputs

Lemma 7.9.2

Let Y^n be the result of passing X^n through a discrete memoryless channel of capacity C .

Then $I(X^n; Y^n) \leq nC$ for all $p(x^n)$.

Proof of Lemma 7.9.2

$$\begin{aligned}
 I(X^n; Y^n) &= H(Y^n) - H(Y^n | X^n) \\
 &= H(Y^n) - \sum_{i=1}^n H(Y_i | Y_1, \dots, Y_{i-1}, X^n) \\
 &= H(Y^n) - \sum_{i=1}^n H(Y_i | X_i) \\
 &\leq \sum_{i=1}^n H(Y_i) - \sum_{i=1}^n H(Y_i | X_i) \\
 &\leq \sum_{i=1}^n I(X_i; Y_i) \leq nC
 \end{aligned}$$

Part C: Converse to the Channel Coding Theorem

Converse to the channel coding theorem

- No rate $R > C$ is achievable.
- Suppose a $(2^{nR}, n)$ code and a uniform distribution on \mathcal{W} .

Proof

$$\begin{aligned}
 nR = H(W) &= \underbrace{H(W | \hat{W}) + H(W) - H(W | \hat{W})}_{= H(W | \hat{W}) + I(W; \hat{W})} \\
 &= H(W | \hat{W}) + I(W; \hat{W}) \\
 &\leq 1 + P(\hat{W} \neq W)nR + \underbrace{I(W; \hat{W})}_{\text{FANO}} \\
 &\quad \boxed{W \rightarrow X^n \rightarrow Y^n \rightarrow \hat{W} \Rightarrow I(W; \hat{W}) \leq I(X^n; Y^n)} \\
 &\leq 1 + P(\hat{W} \neq W)nR + I(X^n; Y^n) \\
 &\leq 1 + P(\hat{W} \neq W)nR + nC
 \end{aligned}$$

Conclusion

$$\boxed{nR \leq 1 + P(\hat{W} \neq W)nR + nC}$$

– dividing by n $R \leq \frac{1}{n} + P(\hat{W} \neq W)R + C$

– Dividing by R and rearranging

$$P(\hat{W} \neq W) \geq 1 - \frac{C}{R} - \frac{1}{nR}$$

– Hence, for large n , if $R > C$, $P(\hat{W} \neq W) > 0$ strictly.
Then it must be true for all n .

$P(\hat{W} \neq W) = \text{average of } \lambda_i \text{'s}$ so $\lambda^{(n)} = \max \lambda_i > 0$ as well

Converse to the channel coding theorem

- No rate $R > C$ is achievable.