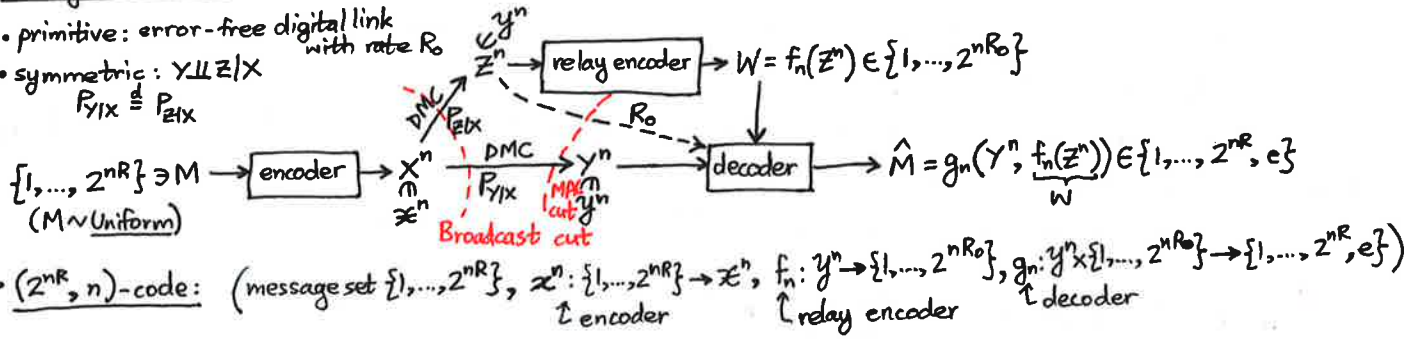


INFORMATION INEQUALITY:

① Relay Channel:

- primitive: error-free digital link with rate R_0
- symmetric: $Y \perp\!\!\!\perp Z | X$
 $P_{Y|X} \triangleq P_{Z|X}$



- $(2^{nR}, n)$ -code: (message set $\{1, \dots, 2^{nR}\}$, $x^n: \{1, \dots, 2^{nR}\} \rightarrow \mathcal{X}^n$, $f_n: \mathcal{Y}^n \rightarrow \{1, \dots, 2^{nR_0}\}$, $g_n: \mathcal{Y}^n \times \{1, \dots, 2^{nR_0}\} \rightarrow \{1, \dots, 2^{nR}, e\}$)
- Probability of error: $P_e^{(n)} = P(\hat{M} \neq M)$ ← average
- Rate R is achievable if \exists sequence of $(2^{nR}, n)$ -codes with $P_e^{(n)} \rightarrow 0$ as $n \rightarrow \infty$.
- Capacity: $C(R_0) \triangleq \sup \{R \geq 0 : R \text{ achievable}\}$.

② Cutset Bound: (Cover-ElGamal '79)

For a primitive relay channel, if rate R is achievable, then $\exists P_X$ such that:

$$R \leq I(X; Y, Z), \quad \text{[Broadcast bound]}$$

$$R \leq I(X; Y) + R_0. \quad \text{[MAC bound]}$$

choice of codes determine P_X^n from P_M via $x^n(\cdot)$
construct from P_X 's by averaging

Remarks:

1. $C(0) = \max_{P_X} I(X; Y)$ and $C(\infty) = \max_{P_X} I(X; Y, Z)$.
 ↑ MAC equality ↑ Broadcast equality (true for $R_0 \geq \log(1/|D|)$)
 [Cover: Find min R_0 s.t. $C(R_0) = \max_{P_X} I(X; Y, Z)$.]
2. Reverse "Physically Degraded Primitive Channel": (not symmetric)

$$P_{Y, Z|X} = P_{Y|X} P_{Z|Y} \quad [Z \perp\!\!\!\perp X | Y] \quad X \rightarrow Y \rightarrow Z$$

$$I(X; Y, Z) = H(X) - H(X|Y, Z) = H(X) - H(X|Y) = I(X; Y)$$

$$\Rightarrow R \leq I(X; Y) \quad \text{[Broadcast bound]}$$

So, $R = \max_{P_X} I(X; Y)$ is the relay channel capacity [for achievability, don't use relay].

3. Zhang '88: Primitive "Stochastically Degraded Channel" (not symmetric)

Assume $Y \perp\!\!\!\perp Z | X$, $P_{Y|X} = P_{Z|X} \circ Q_{Y|Z}$ for some $Q_{Y|Z}$. Then:

$$R_0 > \max_{P_X: I(X; Y) = C_{XY}} I(X; Z) - C_{XY} \Rightarrow C(R_0) < C_{XY} + R_0, \text{ where } C_{XY} = \max_{P_X} I(X; Y).$$

Note: Known that: $R_0 \leq \max_{P_X: I(X; Y) = C_{XY}} I(X; Z) - C_{XY} \Rightarrow C(R_0) = C_{XY} + R_0$. [Intuition: We can send $C_{XY} + R_0$ rate to Z , which will send the R_0 part to decoder.]

Note: $\max_{P_X: I(X; Y) = C_{XY}} I(X; Z) - C_{XY} < R_0 < \max_{P_X: I(X; Y) = C_{XY}} I(X; Y, Z) - C_{XY} \Rightarrow$ Cutset Bound is $R \leq C_{XY} + R_0$ BUT no equality due to Zhang's result.

Proof: $nR = H(M) = I(M; \hat{M}) + H(M | \hat{M}) \leq I(M; \hat{M}) + nJ_n, J_n \rightarrow 0 \text{ as } n \rightarrow \infty$

DPI $\Rightarrow nR \leq I(X^n; Y^n, W) + nJ_n$

↓ Broadcast bound
 $nR \leq I(X^n; Y^n, Z^n) + nJ_n$ [DPI]
 $\leq \sum_{i=1}^n I(X_i; Y_i, Z_i) + nJ_n$ [memoryless]

$= nI(X_Q; Y_Q, Z_Q | Q) + nJ_n$ [$Q \sim \text{Unif}\{1, \dots, n\} \perp\!\!\!\perp (X^n, Y^n, Z^n)$ (time sharing)]
 $\leq nI(X_Q; Y_Q, Z_Q) + nJ_n$ [$Q \rightarrow X_Q \rightarrow (Y_Q, Z_Q)$]
 $\Rightarrow R \leq I(X_Q; Y_Q, Z_Q) + J_n$

MAC bound
 $nR \leq I(X^n; Y^n) + I(W; X^n | Y^n) + nJ_n$
 $\leq \sum_{i=1}^n I(X_i; Y_i) + \frac{H(W|Y^n) - H(W|X^n, Y^n)}{\leq nR_0} + nJ_n$ [memoryless]
 $\leq \sum_{i=1}^n I(X_i; Y_i) + nR_0 + nJ_n$
 $= nI(X_Q; Y_Q | Q) + nR_0 + nJ_n$
 $\leq nI(X_Q; Y_Q) + nR_0 + nJ_n$
 $\Rightarrow R \leq I(X_Q; Y_Q) + R_0 + J_n$

Take $n \rightarrow \infty$, $P_X^{(n)} \rightarrow P_X$ wlog by compactness. $P_{XY} \rightarrow I(X; Y)$ cont. in disc. case.

Tightness
First non-tight example

③ Improving Cutset Bound: (Wu-Özgür-Xie '16)

From MAC bound:

$$nR \leq nI(X_Q; Y_Q) + H(W|Y^n) - H(W|X^n) + nJ_n \quad [\text{Symmetry: } W|Y^n|X^n]$$

Let $H(W|X^n) = n\epsilon_n$. First, observe that:

Intuition

1. If $H(W|X^n) \approx 0$, then all Z^n jointly typical with X^n map to W given X^n . Since $P_{Y|X} = P_{Z|X}$, we also expect $H(W|Y^n) \approx 0$. So, $H(W|Y^n) \leq nR_0$ is a loose bound.
2. If $\epsilon_n \geq \epsilon$ for large n , then $H(W|X^n) \geq 0$ is a loose bound.

Information inequality: $Y^n \xleftarrow{\text{DMC}} X^n \xrightarrow{\text{DMC}} Z^n \xrightarrow{f_n} W$, $\epsilon_n = \frac{H(W|X^n)}{n}$. ← Do not need reliable code information.

$\forall n \in \mathbb{N}$, $H(W|Y^n) \leq n g(\epsilon_n)$ for some function $g: \mathbb{R}^+ \rightarrow \mathbb{R}^+$ continuous with $g(0) = 0$.
↑ explicit examples in [Wu-Özgür-Xie '16]

Then, we have: $nR \leq nI(X_Q; Y_Q) + H(W|Y^n) - n\epsilon_n + nJ_n$
 $\swarrow H(W|Y^n) \leq nR_0$ [card. bound] $\searrow H(W|Y^n) \leq n g(\epsilon_n)$ [info. ineq.]

$$R \leq I(X_Q; Y_Q) + R_0 - \epsilon_n + J_n, \quad R \leq I(X_Q; Y_Q) + g(\epsilon_n) - \epsilon_n + J_n.$$

So, we get: $R \leq I(X_Q; Y_Q, Z_Q) + J_n$, where $0 \leq \epsilon_n \leq R_0$.
 $R \leq I(X_Q; Y_Q) + R_0 - \epsilon_n + J_n$
 $R \leq I(X_Q; Y_Q) + g(\epsilon_n) - \epsilon_n + J_n$

By compactness, $\epsilon_n \rightarrow \epsilon \in [0, R_0]$ because $P_{X_Q}^{(n)} \rightarrow P_X$ WLOG as $n \rightarrow \infty$. So, letting $n \rightarrow \infty$, we derive the following result: ↳ continuity of cond. entropy

Thm: For the symmetric primitive relay channel, if R is achievable, then $\exists P_X, \exists \epsilon \in [0, R_0]$ s.t.:

$$\begin{aligned} R &\leq I(X; Y, Z), \\ R &\leq I(X; Y) + R_0 - \epsilon, \\ R &\leq I(X; Y) + g(\epsilon) - \epsilon. \end{aligned} \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} \begin{array}{l} \text{better than MAC bound:} \\ R \leq I(X; Y) + R_0 \end{array}$$

Remarks:

1. Improvement is not obtained via tensorization.
2. Strictly tighter than cutset bound for $R_0 > 0$. [$\epsilon > 0 \Rightarrow$ obvious, $\epsilon = 0 \Rightarrow$ Ineq. (3) is $R \leq I(X; Y)$]
3. As in cutset bound, all 3 inequalities are coupled by P_X .

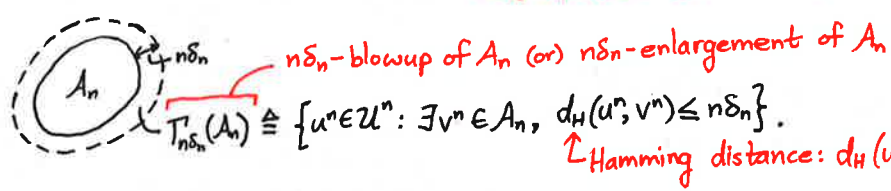
④ Proof of Information Inequality: (special case)

Suppose $Y^n \xleftarrow{\text{DMC}} X^n \xrightarrow{\text{DMC}} Z^n \rightarrow W = f_n(Z^n) \in \{1, \dots, 2^{nR_0}\}$, and $\frac{1}{n} H(W|X^n) \triangleq \epsilon_n \xrightarrow{n \rightarrow \infty} 0$.

Thm: $H(W|Y^n) \leq n g(\epsilon_n)$ for some function $g: \mathbb{R}^+ \rightarrow \mathbb{R}^+$ s.t. $0 = g(0) = \lim_{\epsilon \rightarrow 0^+} g(\epsilon)$, and for sufficiently large n . ↳ continuous at zero

Blowing-Up Lemma: (Ahlsvede-Gács-Körner '76, Marton '86)

Let $U_1, U_2, \dots, U_n \in \mathcal{U}$ be independent random variables. Suppose $A_n \subseteq \mathcal{U}^n$ satisfies $P(U^n \in A_n) \geq 2^{-n\epsilon_n}$ for $\epsilon_n \xrightarrow{n \rightarrow \infty} 0$. Then, \exists sequences $\delta_n \xrightarrow{n \rightarrow \infty} 0$ and $\eta_n \xrightarrow{n \rightarrow \infty} 0$ such that $P(U^n \in T_{n\delta_n}(A_n)) \geq 1 - \eta_n$.
↳ only depend on ϵ_n , not on $A_n \Rightarrow$ can use any A_n with same sequences



Marton's proof: $\eta_n = \frac{\sqrt{\epsilon_n}}{\delta_n}$ and choose δ_n s.t. $\eta_n \xrightarrow{n \rightarrow \infty} 0$.
 Eg: $\delta_n = \epsilon_n^{1/4}, \eta_n = \epsilon_n^{3/4}$

Proof: (Polyanskiy '16)

① Lower Bound $P_{W|X^n}$ using Markov's Inequality:

$$\mathbb{P}(P_{W|X^n}(W|X^n) < 2^{-n\sqrt{\epsilon_n}}) = \mathbb{P}(\log_2 \left(\frac{1}{P_{W|X^n}(W|X^n)} \right) > n\sqrt{\epsilon_n}) \leq \frac{H(W|X^n)}{n\sqrt{\epsilon_n}} = \sqrt{\epsilon_n} \quad [\text{Markov inequality}]$$

$H(W|X^n) = n\epsilon_n$

Let $S = \{(w, x^n) \in \{1, \dots, 2^{nR_0}\} \times \mathcal{X}^n : P_{W|X^n}(w|x^n) \geq 2^{-n\sqrt{\epsilon_n}}\}$, and note that:

$$\mathbb{P}((W, X^n) \in S) \geq 1 - \sqrt{\epsilon_n}. \quad [\text{large } n]$$

② Let $f_n^{-1}(w) \triangleq \{z^n \in \mathcal{Y}^n : f_n(z^n) = w\}$ for any $w \in \{1, \dots, 2^{nR_0}\}$.

Show $\mathbb{P}(Y^n \in T_{n\delta_n}(f_n^{-1}(W)))$ is large using Blowing-Up Lemma:

For $(w, x^n) \in S$, $P_{W|X^n}(w|x^n) = \mathbb{P}(Z^n \in f_n^{-1}(w) | X^n = x^n) \geq 2^{-n\sqrt{\epsilon_n}}$.

Blow-up Lemma \Rightarrow $\mathbb{P}(Z^n \in T_{n\delta_n}(f_n^{-1}(w)) | X^n = x^n) \geq 1 - \eta_n$, where $\delta_n = \eta_n = \epsilon_n^{\frac{1}{8}}$.

$P_{Y|X} \stackrel{\text{Symmetry}}{\approx} P_{X|Y} \Rightarrow \mathbb{P}(Y^n \in T_{n\delta_n}(f_n^{-1}(w)) | X^n = x^n) \geq 1 - \eta_n$

$$\begin{aligned} \mathbb{P}(Y^n \in T_{n\delta_n}(f_n^{-1}(W))) &\geq \sum_{(w, x^n) \in S} \underbrace{\mathbb{P}(Y^n \in T_{n\delta_n}(f_n^{-1}(w)) | X^n = x^n, W = w)}_{\geq 1 - \eta_n} P_{W, X^n}(w, x^n) \\ &\geq (1 - \eta_n) \mathbb{P}((W, X^n) \in S) \\ &\geq (1 - \eta_n)(1 - \sqrt{\epsilon_n}) \leftarrow \eta_n = \epsilon_n^{\frac{1}{8}} \\ &\geq 1 - 2\epsilon_n^{\frac{1}{8}}. \quad [\text{large } n] \end{aligned}$$

③ Bound $H(W|Y^n)$ using List Decoding + Fano's Inequality:
↳ (Ahlsweede - Dueck '76)

Let $E = \mathbb{1}\{Y^n \in T_{n\delta_n}(f_n^{-1}(W))\}$.
↑ indicator r.v.

$$\begin{aligned} H(W|Y^n) &\leq H(W, E|Y^n) = H(E|Y^n) + H(W|E, Y^n) \\ &\leq \underbrace{H(E)}_{\leq h(2\epsilon_n^{\frac{1}{8}})} + \underbrace{P_E(0)H(W|E=0, Y^n)}_{\leq 2\epsilon_n^{\frac{1}{8}} \leq nR_0} + \underbrace{P_E(1)H(W|E=1, Y^n)}_{\approx 1} \end{aligned}$$

[large n] $\leftarrow \leq h(2\epsilon_n^{\frac{1}{8}}) \leq 2\epsilon_n^{\frac{1}{8}} \leq nR_0$ (binary entropy)

$$\begin{aligned} \Rightarrow H(W|Y^n) &\leq h(2\epsilon_n^{\frac{1}{8}}) + n2\epsilon_n^{\frac{1}{8}}R_0 + n(\delta_n \log_2(|\mathcal{Y}|-1) + h(\delta_n)) \\ \Rightarrow H(W|Y^n) &\leq n(h(2\epsilon_n^{\frac{1}{8}}) + 2\epsilon_n^{\frac{1}{8}}R_0 + h(\epsilon_n^{\frac{1}{8}}) + \epsilon_n^{\frac{1}{8}} \log_2(|\mathcal{Y}|-1)) \\ &\triangleq g(\epsilon_n) \rightarrow 0 \text{ as } \epsilon_n \rightarrow 0 \end{aligned}$$

$\therefore H(W|Y^n) \leq ng(\epsilon_n)$.

IF $Y^n \in T_{n\delta_n}(f_n^{-1}(W))$, then $\exists y^n \in f_n^{-1}(W) \cap \{z^n \in \mathcal{Y}^n : d_H(z^n, Y^n) \leq n\delta_n\}$.

Hamming Ball $(n\delta_n)$

$$\Rightarrow H(W|Y^n, E=1) \leq \log_2(|\{z^n \in \mathcal{Y}^n : d_H(z^n, Y^n) \leq n\delta_n\}|) \leq n(\delta_n \log_2(|\mathcal{Y}|-1) + h(\delta_n))$$

↑ bound on volume of Hamming ball (requires $n\delta_n \geq 1$)

Remarks:

- Bound better for larger n .
- Such bound does not hold without factoring $X^n \rightarrow Z^n \rightarrow W$. [Eq: $Y^n \leftarrow X^n \rightarrow W$ let $W = X^n, Y^n \perp\!\!\!\perp X^n$. Then, $H(W|X^n) = 0, H(W|Y^n) = H(X^n)$.] THE END

Volume of Hamming Ball:

Consider the set \mathcal{Y} with $2 \leq |\mathcal{Y}| < +\infty$. For $y^n, z^n \in \mathcal{Y}^n$, let $d_H(y^n, z^n) \triangleq \sum_{i=1}^n \mathbb{1}\{y_i \neq z_i\}$ denote Hamming distance.

Let $\text{Ball}_{\mathcal{Y}^n}(nr) \triangleq \{z^n \in \mathcal{Y}^n : d_H(y^n, z^n) \leq nr\}$ denote the Hamming ball of radius nr around $y^n \in \mathcal{Y}^n$.

Prop: For $0 \leq r \leq 1 - \frac{1}{|\mathcal{Y}|}$, and sufficiently large n (s.t. $nr \geq 1$), we have:

$$\frac{1}{n} \log(|\text{Ball}_{\mathcal{Y}^n}(nr)|) \leq r \log(|\mathcal{Y}| - 1) + h(r). \quad [\text{Note: This is asymptotically tight.}]$$

↑ any base ↑

Proof:

$$1 = (r + (1-r))^n$$

$$= \sum_{i=0}^n \binom{n}{i} r^i (1-r)^{n-i}$$

$$\geq \sum_{i=0}^{nr} \binom{n}{i} r^i (1-r)^{n-i}$$

$$= \sum_{i=0}^{nr} \binom{n}{i} (|\mathcal{Y}| - 1)^i (1-r)^n \left(\frac{r}{(|\mathcal{Y}| - 1)(1-r)}\right)^i$$

$$\geq (1-r)^n \left(\frac{r}{(|\mathcal{Y}| - 1)(1-r)}\right)^{nr} \underbrace{\sum_{i=0}^{nr} \binom{n}{i} (|\mathcal{Y}| - 1)^i}_{= |\text{Ball}_{\mathcal{Y}^n}(nr)|}$$

$$= |\text{Ball}_{\mathcal{Y}^n}(nr)| (1-r)^n \left(\frac{r}{(|\mathcal{Y}| - 1)(1-r)}\right)^{nr} = |\text{Ball}_{\mathcal{Y}^n}(nr)| r^{nr} (1-r)^{n(1-r)} (|\mathcal{Y}| - 1)^{-nr}$$

$$\Rightarrow 0 \geq \frac{1}{n} \log(|\text{Ball}_{\mathcal{Y}^n}(nr)|) + r \log(r) + (1-r) \log(1-r) - r \log(|\mathcal{Y}| - 1)$$

$$\Rightarrow \frac{1}{n} \log(|\text{Ball}_{\mathcal{Y}^n}(nr)|) \leq r \log(|\mathcal{Y}| - 1) + h(r)$$

↑ binary entropy

$$\left[\frac{r}{(|\mathcal{Y}| - 1)(1-r)} \leq 1\right] \Leftrightarrow r \leq (|\mathcal{Y}| - 1)(1-r) \Leftrightarrow r|\mathcal{Y}| \leq |\mathcal{Y}| - 1$$

this is assumed

Intuition: $|\text{Ball}_{\mathcal{Y}^n}(nr)| \leq \underbrace{(|\mathcal{Y}| - 1)^{nr}}_{\text{no. of choices of replacement at } nr \text{ locations}} \cdot \underbrace{\exp(nh(r))}_{\text{no. of binary strings of length } n \text{ with type } (nr \text{ 1's, } n(1-r) \text{ 0's)}} \leftarrow \boxed{\text{nr distance dominates lower distance strings}}$

↑ encode changed letters