

Lecture 20

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(1)

Error Analysis:

$$\overline{P_e} \equiv \frac{1}{|M|} \sum_{m=1}^{|M|} \text{Tr} \left\{ (I - \Lambda_m^{B^n}) \sigma_{x^n(m)}^{B^n} \right\}$$

$$\text{Since } \text{Tr} \left\{ \Pi_{\sigma, s}^{B^n} \sigma_{x^n(m)}^{B^n} \right\} \geq 1 - \epsilon$$

$$\begin{aligned} \overline{P_e} &\leq \frac{1}{|M|} \sum_{m=1}^{|M|} \text{Tr} \left\{ (I - \Lambda_m^{B^n}) \Pi_{\sigma, s}^{B^n} \sigma_{x^n(m)}^{B^n} \Pi_{\sigma, s}^{B^n} \right\} \\ &\quad + 2\sqrt{\epsilon} \end{aligned} \quad (*)$$

Now we need Hayashi-Nagaoka :

$$I - (S+T)^{-1/2} S (S+T)^{-1/2} \leq$$

$$2(I-S) + 4T \quad \text{for } 0 \leq S \leq I \quad T \geq 0$$

(can prove this)

$$\text{Set } S = \Pi_{\sigma_{x^n(m)}, s}^{B^n/x^n}$$

$$T = \sum_{m' \neq m} \Pi_{\sigma_{x^n(m')}, s}^{B^n/x^n}$$

Since $\Pi_{\sigma, s}^{B^n} \sigma_{x^n(m)}^{B^n} \Pi_{\sigma, s}^{B^n} \geq 0$, we have

$$\begin{aligned} (*) &\leq \frac{1}{|M|} \sum_{m=1}^{|M|} \left[2 \text{Tr} \left\{ (I - \Pi_{\sigma_{x^n(m)}, s}^{B^n/x^n}) \Pi_{\sigma, s}^{B^n} \sigma_{x^n(m)}^{B^n} \Pi_{\sigma, s}^{B^n} \right\} \right. \\ &\quad \left. + 4 \text{Tr} \left\{ \sum_{m' \neq m} \Pi_{\sigma_{x^n(m')}, s}^{B^n/x^n} \Pi_{\sigma, s}^{B^n} \sigma_{x^n(m)}^{B^n} \Pi_{\sigma, s}^{B^n} \right\} \right] \\ &\quad + 2\sqrt{\epsilon} \end{aligned} \quad (**)$$

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Consider 1st term:

$$\text{Tr} \left\{ \left(I - \Pi_{\sigma_{x^n(m)}, s}^{B^n | X^n} \right) \Pi_{\sigma_s, s}^{B^n} \sigma_{x^n(m)}^{B^n} \Pi_{\sigma_s, s}^{B^n} \right\}$$

Again, since $\text{Tr} \left\{ \Pi_{\sigma_s, s}^{B^n} \sigma_{x^n(m)}^{B^n} \right\} \geq 1 - \epsilon$,

the above is less than

$$\text{Tr} \left\{ \left(I - \Pi_{\sigma_{x^n(m)}, s}^{B^n | X^n} \right) \sigma_{x^n(m)}^{B^n} \right\} + 2\sqrt{\epsilon}$$

$$\leq \epsilon + 2\sqrt{\epsilon}$$

↑ by conditional typicality

so,

$$(\star\star) \leq 2(\epsilon + 2\sqrt{\epsilon}) +$$

$$\frac{4}{|M|} \sum_{m=1}^{|M|} \sum_{m' \neq m} \text{Tr} \left\{ \Pi_{\sigma_{x^n(m')}, s}^{B^n | X^n} \Pi_{\sigma_s, s}^{B^n} \sigma_{x^n(m)}^{B^n} \Pi_{\sigma_s, s}^{B^n} \right\}$$

can't really do much of
anything w/ this term
in the general case

invoke the random coding
argument!

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Consider the expectation of the average error probability over the random choice of code

$$\mathbb{E}_{X^n} \{ \bar{P}_e \} \leq 2(\epsilon + 2\sqrt{\epsilon})$$

$$+ \frac{4}{|M|} \mathbb{E}_{X^n} \left\{ \sum_{m=1}^{|M|} \sum_{m' \neq m} \text{Tr} \left\{ \Pi_{\sigma_{X^n(m)}, S}^{B^n | X^n} \Pi_{\sigma_{X^n(m')}, S}^{B^n} \sigma_{X^n(m')}^{B^n} \Pi_{S, S}^{B^n} \right\} \right\}$$

$m' \neq m$ one difference

By the way we chose
the code, it means
that $X^n(m') \neq X^n(m)$
are independent

$$\text{then } = 2(\epsilon + 2\sqrt{\epsilon}) +$$

$$\frac{4}{|M|} \sum_{m=1}^{|M|} \sum_{m' \neq m} \text{Tr} \left\{ \mathbb{E}_{X^n} \left\{ \Pi_{\sigma_{X^n(m')}, S}^{B^n | X^n} \right\} \right. \\ \left. \Pi_{S, S}^{B^n} \mathbb{E}_{X^n} \left\{ \sigma_{X^n(m)}^{B^n} \right\} \Pi_{S, S}^{B^n} \right\}$$

Note that $\mathbb{E}_{X^n} \left\{ \sigma_{X^n}^{B^n} \right\} \leq [1-\epsilon]^{-1} \sigma^{\otimes n}$

$$\text{then } \leq 2(\epsilon + 2\sqrt{\epsilon}) +$$

$$\left[1 - \epsilon \right]^{-1} \frac{4}{|M|} \sum_{m=1}^{|M|} \sum_{m' \neq m} \text{Tr} \left\{ \mathbb{E}_{X^n} \left\{ \Pi_{\sigma_{X^n(m')}, S}^{B^n | X^n} \right\} \right. \\ \left. \Pi_{S, S}^{B^n} \sigma^{\otimes n} \Pi_{S, S}^{B^n} \right\}$$

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$$\leq 2(\epsilon + 2\sqrt{\epsilon}) +$$

$$\{1-\epsilon\}^{-1} \frac{4}{|\mathcal{U}|} \sum_{m=1}^{|\mathcal{U}|} \sum_{m' \neq m} \mathbb{E}_{x^n} \left\{ \text{Tr} \left\{ \Pi_{\sigma_{x^n(m')}, s}^{B^n/x^n} \right. \right. \\ \left. \left. \Pi_{\sigma, s}^{B^n} \right\} \right\} 2^{-n \{H(B)-s\}}$$

Since $\Pi_{\sigma, s}^{B^n} \leq I$

$$\Rightarrow \Pi_{\sigma_{x^n(m')}, s}^{B^n/x^n} \Pi_{\sigma, s}^{B^n} \Pi_{\sigma_{x^n(m')}, s}^{B^n/x^n} \leq \Pi_{\sigma_{x^n(m')}, s}^{B^n/x^n}$$

∴

$$\leq 2(\epsilon + 2\sqrt{\epsilon}) +$$

$$\{1-\epsilon\}^{-1} \frac{4}{|\mathcal{U}|} \sum_{m=1}^{|\mathcal{U}|} \sum_{m' \neq m} \mathbb{E}_{x^n} \left\{ \text{Tr} \left\{ \Pi_{\sigma_{x^n(m')}, s}^{B^n/x^n} \right\} \right\} 2^{-n \{H(B)-s\}}$$

$$\leq 2(\epsilon + 2\sqrt{\epsilon}) +$$

$$\{1-\epsilon\}^{-1} \frac{4}{|\mathcal{U}|} \sum_{m=1}^{|\mathcal{U}|} \sum_{m' \neq m} 2^{-n \{H(B)-s\}} 2^{n \{H(B)x\} + s}$$

$$\leq 2(\epsilon + 2\sqrt{\epsilon}) +$$

$$\{1-\epsilon\}^{-1} 4/|\mathcal{U}| 2^{-n \{I(x; B) - 2s\}}$$

— choose $|\mathcal{U}| = 2^{n(I(x; B) - 3s)}$ + bound becomes

$$2(\epsilon + 2\sqrt{\epsilon}) + \{1-\epsilon\}^{-1} 4 \cdot 2^{-ns}$$

which becomes arbitrarily small as $n \rightarrow \infty$

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- Since we have a good bound on expectation of average error prob., there must exist a particular code w/ its average error prob $\leq \epsilon'$
- By throwing away the worst half of the codewords, we have a code w/ maximal error prob $\leq 2\epsilon'$ w/ an asymptotically negligible loss in rate
- The rate is $\frac{\log |\mathcal{U}|}{n} = I(X; B) - 3\delta$

Thus, the rate $I(X; B)$ is achievable.

We can get $X(N)$ by ~~choosing~~ making codes from the ensemble that attain the maximum in

$$X(N) = \max_{P(X)} I(X; B)$$

To achieve $\frac{1}{k} X(N^{(k)})$, build codes for the channel $X(N^{(k)})$ instead. This shows that $\lim_{k \rightarrow \infty} \frac{1}{k} X(N^{(k)})$ is achievable.

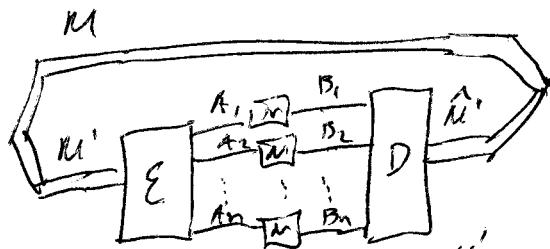
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Converse Theorem

consider the task of common randomness generation instead.

CR capacity has to be an upper bound on classical comm. capacity b/c a classical bit channel can always generate common randomness.



$$\|D^{B^n \rightarrow \hat{u}}(N^{\otimes n}(E^{u' \rightarrow A^n}(\Phi^{uu'}))) - \Phi^{u\hat{u}}\|_1 \leq \epsilon$$

where $\Phi^{u\hat{u}} = \frac{1}{m} \cdot \sum_{m=1}^M m \otimes m \otimes m$ {If this holds for a good code,

rate of common randomness is

$$\log \frac{|M|}{n} = C$$

What is limit on rate?

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$$nC = \log |\mu|$$

$$= H(\mu)_{\overline{\Phi}} - H(\mu|\hat{\mu})_{\overline{\Phi}}$$

$$= I(\mu; \hat{\mu})_{\overline{\Phi}}$$

$$\leq I(\mu; \hat{\mu})_{w^*} + n\epsilon'$$

$$\leq I(\mu; B^n)_{w^*} + n\epsilon' \quad (\text{QDP})$$

Alicki-Fannes'

state w is a classical quantum state of
the form

$$\sum_m p(m) |m\rangle\langle m| \otimes N_{pm}$$

must have mutual info less than optimal rate

$$\rightarrow \leq \chi(N^{\otimes n}) + n\epsilon'$$

$$\therefore C \leq \frac{1}{n} \chi(N^{\otimes n}) + \epsilon'$$

statement of classical capacity theorem

$$\sup \{C : C \text{ is achievable}\} = \lim_{k \rightarrow \infty} \frac{1}{k} \chi(N^{\otimes k})$$

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In general, cannot compute capacity.

But suppose N is entanglement breaking so that

$$N(\rho^{A'A}) = \sum_y p(y) \sigma_y^{A'} \otimes w_y^B$$

for any entangled state $\rho^{A'A}$.

- Then we can show that $\chi(N)$ is capacity
- would like to prove additivity of χ in this case:

$$\chi(N_1 \otimes N_2) = \chi(N_1) + \chi(N_2)$$

$$\chi(N_1 \otimes N_2) \geq \chi(N_1) + \chi(N_2) \text{ always holds}$$

Let's prove $\chi(N_1 \otimes N_2) \leq \chi(N_1) + \chi(N_2)$
consider arbitrary state

$$\rho^{XA_1 A_2} = \sum_x p(x) |x\rangle\langle x|^X \otimes (\rho_x^{A_1 A_2})$$

Inputting to tensor product channel gives

$$\rho^{XB_1 B_2} = (N_1^{A_1 \rightarrow B_1} \otimes N_2^{A_2 \rightarrow B_2})(\rho^{XA_1 A_2})$$

Suppose N_2 is entanglement breaking - Then

$$\rho^{XB_1 B_2} = \sum_x p(x) |x\rangle\langle x|^X \otimes \sum_y p(y|x) [N_1(\sigma_y, x) \otimes \cancel{w_y^B}]$$

can amalgamate this state as

$$\rho^{XYB_1 B_2} = \sum_{x,y} p(x)p(y|x) |x\rangle\langle x|^X \otimes |y\rangle\langle y|^Y \otimes N_1(\sigma_y, x) \otimes w_{y,x}^{B_2}$$

Then

$$I(X; B_1 B_2) = H(B_1 B_2) - H(B_1 B_2 | X)$$

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$$\leq H(B_1) + H(B_2) - H(B_1 B_2 | X)$$

$$\leq H(B_1) + H(B_2) - H(B_1 B_2 | XY)$$

$$= H(B_1) + H(B_2) - H(B_1 | XY) - H(B_2 | XY)$$

$$= I(XY; B_1) + I(XY; B_2)$$

$$\leq \chi(n_1) + \chi(n_2)$$