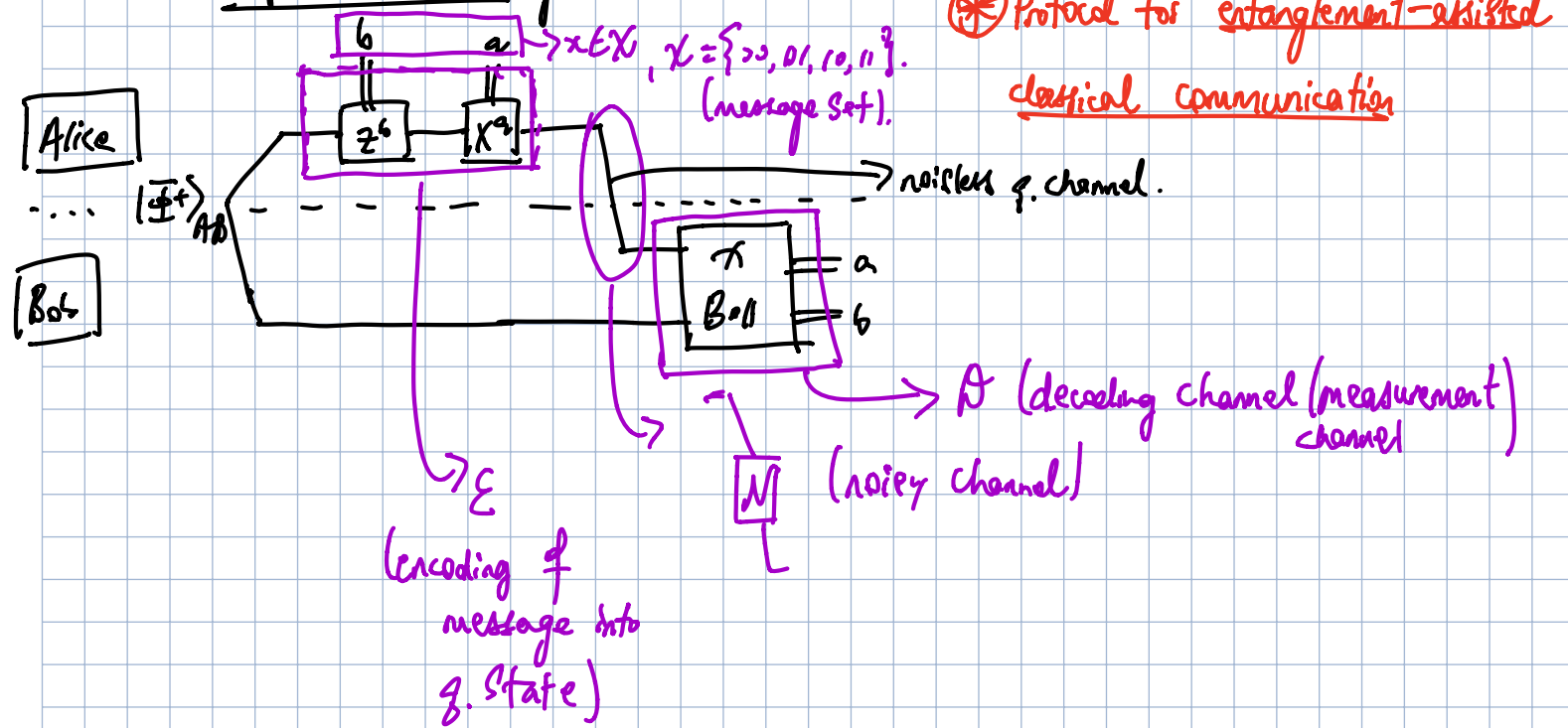
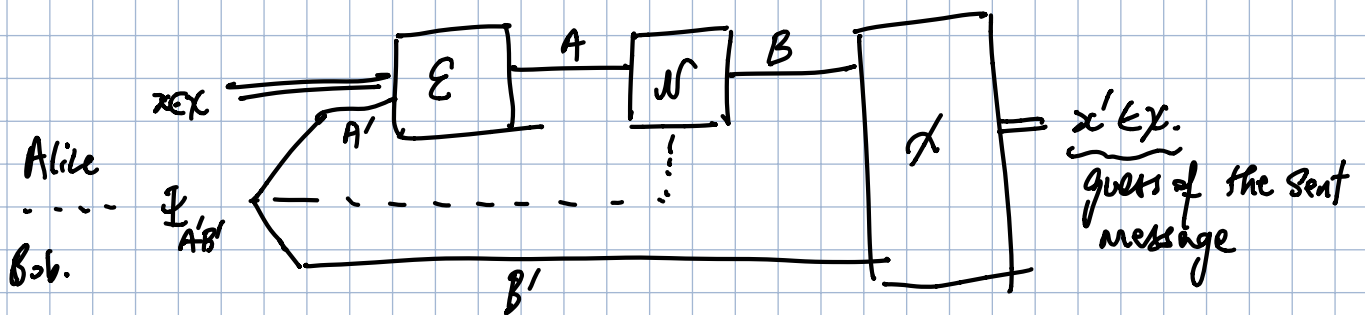


Recall Superdense coding



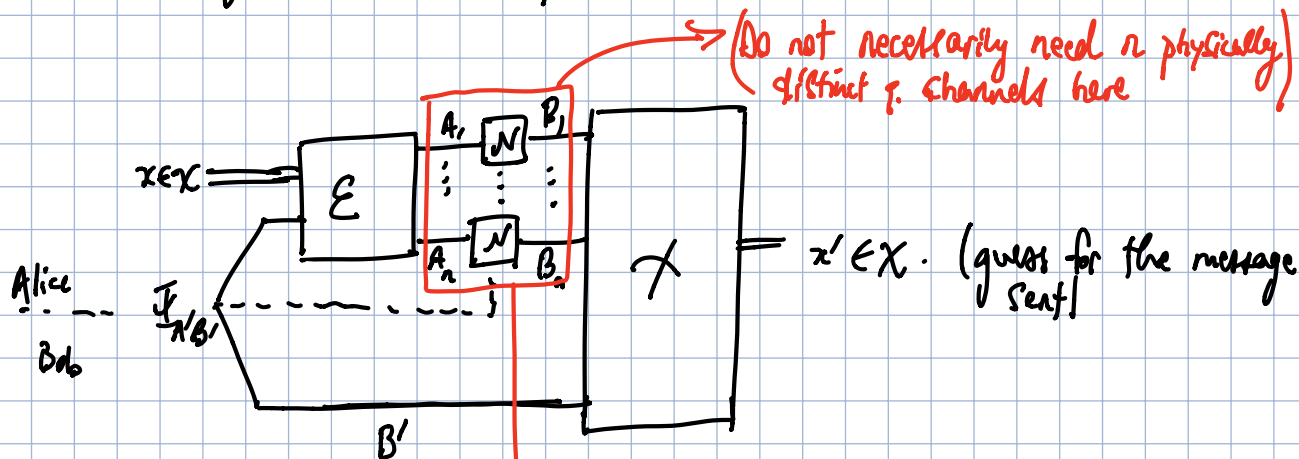
General Classical Communication.

- Messages $X = \{x_1, x_2, \dots, x_m\}$ to be sent by Alice ($\log_2 |X|$ bits of inf.).
 - Prior $p: X \rightarrow [0,1]$ prob. distribution. $\rightarrow$ take uniform: $p(x) = \frac{1}{|X|} \forall x \in X$.
 - Given:
 - Q. channel $N_{A \rightarrow B}$ from Alice to Bob (with many uses).
 - Local encoding and decoding.
 - Entanglement-assistance
- These elements combined are called a code for ent.-assisted classical communication.
- Protocol: "One-shot" Setting: (one channel use only).



⊛ In the One-shot Setting, we allow for some error. In general, it is not possible to get zero with just one channel use.

Asymptotic Setting : unlimited # of channel uses.



⊛ In this setting, we want error to be zero, so we allow an unlimited # of channel uses in order to make this happen.

$$\boxed{\boxed{N^n}}$$

⊛ So n channel uses is equivalent to one use of N^n .

⊛ Main Goal : perfectly send the classical message from Alice to Bob.

$$|\Phi\rangle_{X_A X'_A} = \sum_{x \in X} \sqrt{\frac{1}{|X|}} |x\rangle_{X_A} \otimes |x\rangle_{X'_A} \otimes |\Phi\rangle_{A'B'}$$

↓ measure - picks a message

$$\bar{|\Phi\rangle}_{X_A X'_A} = \sum_{x \in X} \frac{1}{|X|} |x\rangle_{X_A} \otimes |x\rangle_{X'_A} \otimes \Phi_{A'B'}$$

↓ encode

$$\rho_{X_A A} = \sum_{x \in X} \frac{1}{|X|} |x\rangle_{X_A} \otimes \underbrace{E_{X'_A \rightarrow A}(|x\rangle_{X'_A} \otimes \Phi_{A'B'})}_{\tau_{AB'}^x}$$

↓ Send through channel

$$\rho_{X_A B} = \sum_{x \in X} \frac{1}{|X|} |x\rangle_{X_A} \otimes \rho_{B B'}^x, \quad \rho_{B B'}^x = N_{A \rightarrow B}(\tau_{AB'}^x)$$

decode: $\{M_{AB}^x\}_{x \in \mathcal{X}}$

$$\omega_{X_A X_B} = M_{B \rightarrow K_B}(\rho_{X_A B}) = \sum_{x, z \in \mathcal{K}_X} \frac{1}{|\mathcal{X}|} \text{Tr}(M_{AB}^{x'} \rho_{AB}^x) |x, z\rangle \langle x, z|_{K_A K_B}$$

Figure of merit: Registers $X_A + X_B$ should have the same value

$$\Pi_{X_A X_B} = \sum_{x \in \mathcal{X}} |x\rangle \langle x|_{K_A} \otimes |x\rangle \langle x|_{K_B} \quad (\text{projector corresponding to } X_A = X_B).$$

$$\Rightarrow \boxed{P_{\text{err}}(\mathcal{E}, \mathcal{A}, \mathcal{F}_{AB}; \mathcal{N}) = 1 - \text{Tr}[\Pi_{X_A X_B} \omega_{X_A X_B}]} \quad (\text{probability that } K_A + K_B \text{ do not have the same value.})$$

Main Questions:

① What is the smallest possible value of P_{err} over all codes $\mathcal{E}, \mathcal{A}, \mathcal{F}_{AB}$?
(given $\mathcal{X}$)

$$E_{\mathcal{E}}(|\mathcal{X}|, \mathcal{N}) = \min_{(\mathcal{E}, \mathcal{A}, \mathcal{F})} \{P_{\text{err}}(\mathcal{E}, \mathcal{A}, \mathcal{F}; \mathcal{N}) : \mathcal{E}_{X_A A' \rightarrow A}, \mathcal{A}_{B B' \rightarrow K_B} \text{ CTP, } d_{K_A} = d_{K_B} = |\mathcal{X}|\}$$

* ② Given tolerable error ε , what is the largest value of $\log_2 |\mathcal{X}|$?
Main focus.

$$C_{\mathcal{E}}^{\varepsilon}(\mathcal{N}) = \max_{(|\mathcal{X}|, \mathcal{E}, \mathcal{A}, \mathcal{F})} \{\log_2 |\mathcal{X}| : P_{\text{err}}(\mathcal{E}, \mathcal{A}, \mathcal{F}; \mathcal{N}) \leq \varepsilon\}$$

* In the asymptotic setting, what is the largest value of $\frac{1}{n} \log_2 |\mathcal{X}|$ such that $\varepsilon \rightarrow 0$ as $n \rightarrow \infty$ (n : # of channel uses)?

$$\text{Rate} \equiv R = \frac{\log_2 |\mathcal{X}|}{n} \Rightarrow |\mathcal{X}| = 2^{nR} \quad \rightarrow \text{bits per channel use}$$

Entanglement-assisted Classical Capacity of $\mathcal{N}$:

$$C_E(\mathcal{N}) = \max \{R : \lim_{n \rightarrow \infty} E_{\mathcal{E}}(2^{n(R-\delta)}; \mathcal{N}^{\otimes n}) = 0 \quad \forall \delta > 0\}$$

"Highest achievable rate"

* Alternate formula: $C_E(W) = \lim_{\epsilon \rightarrow 0} \liminf_{n \rightarrow \infty} \frac{1}{n} C_E(W^{\otimes n})$

$= \lim_{\epsilon \rightarrow 0} \sup_{n \in \mathbb{N}} \inf \left\{ \frac{1}{m} C_E(W^{\otimes m}) : m \geq n \right\}$

Main Result: $C_E(W) = I(W) = \sup_{\mathcal{P}_A} \inf_{\mathcal{P}_B} D(W_{A \rightarrow B}(\mathcal{P}_A) \| \mathcal{P}_A \otimes \sigma_B)$

⊗ Direct g. analogue of Shannon's noisy channel coding theorem.

$= \sup_{\mathcal{P}_A} I(A; B)_W \quad (W_{AB} = W_{A \rightarrow B}(\mathcal{P}_A))$

$C(W) = \max_{P_X} I(X; Y)$

$D(\rho \| \sigma) = \begin{cases} \text{Tr}[\rho \log \rho - \rho \log \sigma] & \text{if } \text{supp}(\rho) \leq \text{supp}(\sigma) \\ +\infty & \text{otherwise.} \end{cases}$

⊗ Quantum Relative entropy (generalization of KL-divergence for classical prob. distributions: $D(p \| q) = \sum_{x \in X} p(x) \log \left(\frac{p(x)}{q(x)} \right)$)

$I(A; B)_\rho = D(\rho_{AB} \| \rho_A \otimes \rho_B)$

$= \inf_{\sigma_B} D(\rho_{AB} \| \rho_A \otimes \sigma_B)$ is called mutual inf.

(Intuitively: how close the given state is to a product state)

$H(A)_\rho \equiv H(\rho) = -\text{Tr}[\rho \log \rho] = -D(\rho \| \mathbb{1})$ ("von Neumann Entropy")

$I(A; B)_\rho = H(A)_\rho + H(B)_\rho - H(AB)_\rho$

Properties of Q. Relative Entropy:

① Faithful: $D(\rho \| \sigma) = 0$ iff $\rho = \sigma$.

② If $\rho \leq \sigma$, then $D(\rho \| \sigma) \leq 0$.

③ $D(\rho_1 \otimes \rho_2 \| \sigma_1 \otimes \sigma_2) = D(\rho_1 \| \sigma_1) + D(\rho_2 \| \sigma_2)$.

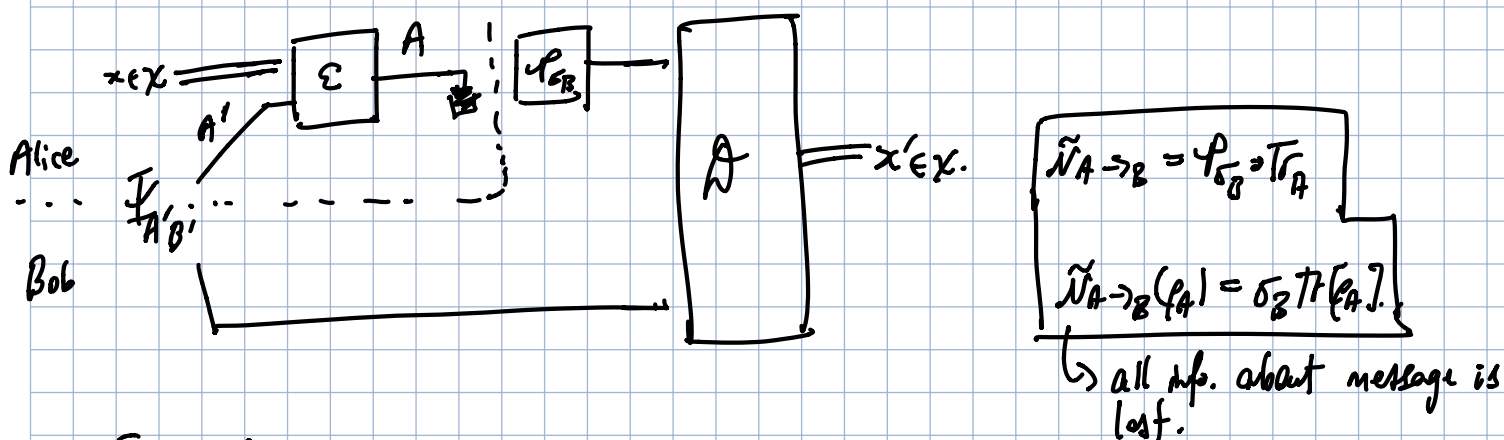
④ $D(W(\rho) \| W(\sigma)) \leq D(\rho \| \sigma) \quad \forall \quad \xi\text{-channels } W \rightarrow$ Data-processing inequality.

Prove in two Parts

① $C_E(W) \leq I(W)$ $\rightarrow$ main focus

② $C_E(W) \geq I(W)$ - requires an explicit protocol construction.

$\rightarrow$ Key idea: Compare the protocol over $\mathcal{N}$ with the protocol over a useless channel (i.e., replacement channel). Intuitively, the useless channel is the one that "cuts the line".



From above:

$$\rho_{XAA} = \sum_{x \in X} \frac{1}{|X|} |x\rangle\langle x|_{X_A} \otimes \underbrace{\mathcal{E}_{X_{A'} \rightarrow A}(|x\rangle\langle x|_{X_{A'}} \otimes \mathbb{I}_{A'B'})}_{\tau_{AB'}^x}$$

$$= \frac{1}{|X|} \sum_{x \in X} |x\rangle\langle x|_{X_A} \otimes \tau_{AB'}^x$$

$$\tilde{\mathcal{N}}_{A \rightarrow B}(\tau_{AB'}^x) = \sigma_B \otimes \text{Tr}_A[\tau_{AB'}^x] = \sigma_B \otimes \mathbb{I}_{B'}, \quad (\mathbb{I}_{B'} = \text{Tr}_A[\mathbb{I}_{AB'}])$$

$$\Rightarrow \text{final state: } \tilde{\omega}_{XAB} = \frac{1}{|X|} \sum_{x \in X} |x\rangle\langle x|_{X_A} \otimes \sum_{x' \in X} \text{Tr}[\mathcal{M}_{BB'}^{x'}(\sigma_B \otimes \mathbb{I}_{B'})] |x'\rangle\langle x'|_{X_B}$$

$$= \frac{1}{|X|} \sum_{x \in X} \text{Tr}[\mathcal{M}_{BB'}^{x'}(\sigma_B \otimes \mathbb{I}_{B'})] |x\rangle\langle x|_{X_A} \otimes |x'\rangle\langle x'|_{X_B}$$

$$= \frac{1}{|X|} \sum_{x \in X} |x\rangle\langle x|_{X_A} \otimes \sum_{x' \in X} \text{Tr}[\mathcal{M}_{BB'}^{x'}(\sigma_B \otimes \mathbb{I}_{B'})] |x'\rangle\langle x'|_{X_B}$$

$$= \frac{1}{|x|} \mathcal{Y}_{K_A} \otimes \tilde{\omega}_{K_B} = \omega_{K_A} \otimes \tilde{\omega}_{K_B}, \tilde{\omega}_{K_B} = A_{BB' \rightarrow K_B} (\sigma_B \otimes \mathbb{I}_{B'})$$

Now, comparator test: $\text{Tr}[\Pi_{K_A K_B} \tilde{\omega}_{K_A K_B}]$

$$= \frac{1}{|x|} \text{Tr}[\Pi_{K_A K_B} (\mathcal{Y}_{K_A} \otimes \tilde{\omega}_{K_B})]$$

$$= \frac{1}{|x|} \text{Tr}[\underbrace{\text{Tr}_{K_A}[\Pi_{K_A K_B}]}_{\mathcal{Y}_{K_B}} \tilde{\omega}_{K_B}]$$

$$= \frac{1}{|x|} \underbrace{\text{Tr}[\tilde{\omega}_{K_B}]}_1$$

$$= \frac{1}{|x|} \Rightarrow \log_2 |x| = -\log_2 \text{Tr}[\Pi_{K_A K_B} (\omega_{K_A} \otimes \tilde{\omega}_{K_B})]$$

For the actual channel $\mathcal{N}_{A \rightarrow B}$, assume that $p_{\text{err}}(\mathcal{E}, \mathcal{A}, \mathbb{P}; \mathcal{N}) \leq \varepsilon$

$$\Rightarrow \text{Tr}[\Pi_{K_A K_B} \omega_{K_A K_B}] \geq 1 - \varepsilon$$

(Protocol is an arbitrary element in the optimization for $\zeta_{\mathcal{E}}(\mathcal{N})$.)

$$\text{Then } \log_2 |x| = -\log_2 \text{Tr}[\Pi_{K_A K_B} (\omega_{K_A} \otimes \tilde{\omega}_{K_B})]$$

$$\leq \max_{\Lambda} \left\{ -\log_2 \text{Tr}[\Lambda (\omega_{K_A} \otimes \tilde{\omega}_{K_B})] : 0 \leq \Lambda \leq \mathbb{I}, \text{Tr}[\Lambda \omega_{K_A K_B}] \geq 1 - \varepsilon \right\}$$

$$= D_H^{\varepsilon}(\omega_{K_A K_B} \| \omega_{K_A} \otimes \tilde{\omega}_{K_B}) \rightarrow \text{Hypothesis testing relative entropy.}$$

$$D_H^{\varepsilon}(\rho \| \sigma) = \max_{\Lambda} \left\{ -\log_2 \text{Tr}[\Lambda \sigma] : 0 \leq \Lambda \leq \mathbb{I}, \text{Tr}[\Lambda \rho] \geq 1 - \varepsilon \right\}$$

(can be computed via Semi-definite programming (SDP))

$$\Rightarrow \log_2 |x| \leq D_H^{\varepsilon}(\omega_{K_A K_B} \| \omega_{K_A} \otimes \tilde{\omega}_{K_B}) \rightarrow \text{depends on the protocol! Get a bound that depends only on } \mathcal{N}!$$

$$\left(\omega_{K_A K_B} = (A_{BB' \rightarrow K_B} \circ \mathcal{N}_{A \rightarrow B} \circ \mathcal{E}_{K_A' \rightarrow A}) (\bar{\mathbb{P}}_{K_A K_A'} \otimes \bar{\mathbb{P}}_{A' B'}) \right)$$

$$\Rightarrow \omega_A = \frac{1}{|A|} \mathbb{I}_A$$

$$= D_H^\varepsilon(\mathcal{N}_{A \rightarrow B}(\sigma_{AB'}) \| \mathcal{N}_{A \rightarrow B}(\omega_A \otimes \sigma_B \otimes \mathbb{I}_{B'}))$$

$$\leq D_H^\varepsilon(\sigma_{AB'} \| \omega_A \otimes \sigma_B \otimes \mathbb{I}_{B'})$$

($D_H^\varepsilon(\mathcal{N}(\rho) \| \mathcal{N}(\sigma)) \leq D_H^\varepsilon(\rho \| \sigma)$
Data-processing inequality).

$$= D_H^\varepsilon(\mathcal{N}_{A \rightarrow B}(\rho_{AB'A}) \| \omega_A \otimes \mathbb{I}_{B'} \otimes \sigma_B)$$

$$\rho_{AB'A} = E_{A'A \rightarrow A}(\bar{\mathbb{I}}_{AX_A'} \otimes \mathbb{I}_{A'B'})$$

$$\text{Tr}_A[\rho_{AB'A}] = \underbrace{\bar{\mathbb{I}}_{AX_A'}}_{\frac{1}{|A|} \mathbb{I}_{X_A}} \otimes \mathbb{I}_{B'} = \omega_{X_A}$$

$$= D_H^\varepsilon(\mathcal{N}_{A \rightarrow B}(\rho_{RA}) \| \rho_R \otimes \sigma_B). \quad (R \equiv X_{AB'})$$

σ_B above was arbitrary $\Rightarrow$ optimize over all states to get the tightest upper bound.

$$\Rightarrow \log_2 |X| \leq \inf_{\sigma_B} D_H^\varepsilon(\mathcal{N}_{A \rightarrow B}(\rho_{RA}) \| \rho_R \otimes \sigma_B)$$

Now optimize over all states ρ_{RA} — like optimizing over all encoding channels.

$$\Rightarrow \log_2 |X| \leq \sup_{\rho_{RA}} \inf_{\sigma_B} D_H^\varepsilon(\mathcal{N}_{A \rightarrow B}(\rho_{RA}) \| \rho_R \otimes \sigma_B) = I_H^\varepsilon(W).$$

$\Rightarrow$ Suffices to optimize over pure states $\rho_{RA}, d_R = d_A$

$$\Rightarrow \log_2 |X| \leq I_H^\varepsilon(W) \text{ — Hypothesis testing mutual information.}$$

$$\Rightarrow \boxed{C_E^\varepsilon(W) \leq I_H^\varepsilon(W)}$$

Now, can get upper bound on $C_\epsilon(W)$. Important fact:

$$\lim_{\epsilon \rightarrow 0} \lim_{n \rightarrow \infty} \frac{1}{n} D_\epsilon^\epsilon(\rho^{\otimes n} \| \sigma^{\otimes n}) = D(\rho \| \sigma). \quad (\text{Quantum Stein's Lemma}).$$

$$D(\rho \| \sigma) = \begin{cases} \text{Tr}[\rho \log \rho - \rho \log \sigma] & \text{if } \text{Supp}(\rho) \leq \text{Supp}(\sigma) \\ +\infty & \text{otherwise.} \end{cases}$$

⊗ Quantum Relative entropy (generalization of KL-divergence for classical prob. distributions: $D(p \| q) = \sum_{x \in \mathcal{X}} p(x) \log \left(\frac{p(x)}{q(x)} \right)$)

$$\Rightarrow C_\epsilon(W) = \lim_{\epsilon \rightarrow 0} \lim_{n \rightarrow \infty} \frac{1}{n} C_\epsilon^\epsilon(W^{\otimes n})$$

$$\leq \lim_{\epsilon \rightarrow 0} \lim_{n \rightarrow \infty} \frac{1}{n} I_\epsilon^\epsilon(W^{\otimes n})$$

$$\leq I(W). \quad \Rightarrow \boxed{C_\epsilon(W) \leq I(W)}.$$

$$I(A:B)_\rho = D(\rho_{AB} \| \rho_A \otimes \rho_B) \\ = \inf_{\sigma_B} D(\rho_{AB} \| \rho_A \otimes \sigma_B) \quad \text{is called } \underline{\text{mutual inf.}}$$

(Intuitively: how close the given state is to a product state)

$$H(A)_\rho \equiv H(\rho) = -\text{Tr}[\rho \log \rho] = -D(\rho \| \mathbb{I}) \quad (\text{"von Neumann Entropy"})$$

$$I(A:B)_\rho = H(A)_\rho + H(B)_\rho - H(AB)_\rho$$

Also, coherent information: $I(A \rightarrow B)_\rho = H(B)_\rho - H(AB)_\rho$

Properties of Relative Entropy:

① Faithful: $D(p||\sigma) = 0$ iff $p = \sigma$.

② If $p \leq \sigma$, then $D(p||\sigma) \leq 0$.

③ $D(p_1 \otimes p_2 || \sigma_1 \otimes \sigma_2) = D(p_1 || \sigma_1) + D(p_2 || \sigma_2)$.

④ $D(N(p)||N(\sigma)) \leq D(p||\sigma) \quad \forall \text{ } \mathcal{E}\text{-channels } N \rightarrow \text{Data-processing inequality.}$