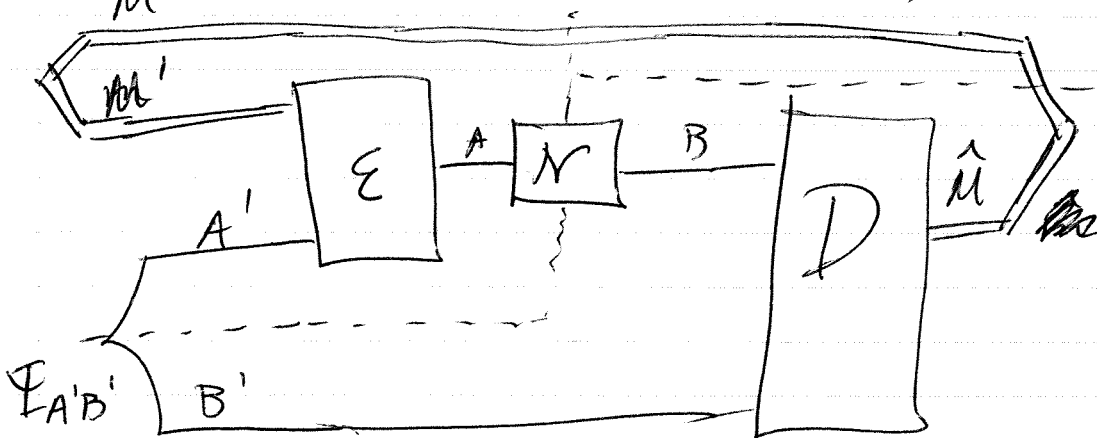


Lecture 25

1

Entanglement-assisted communication

Begin w/ one-shot setting



Initial state is

$$\bar{\Phi}_{MM'}^P \otimes \Psi_{A'B'}$$

where
$$\bar{\Phi}_{MM'}^P = \sum_{m \in M} p(m) |m\rangle\langle m|_M \otimes |m\rangle\langle m|_{M'}$$

Then encoding and decoding lead to final state

$$\left(D_{BB' \rightarrow \hat{M}} \circ N_{A \rightarrow B} \circ E_{M'A \rightarrow A} \right) \left(\bar{\Phi}_{MM'}^P \otimes \Psi_{A'B'} \right)$$

(2)

Note that decoder is a measurement channel

$$D_{BB' \rightarrow \hat{M}}(\tau_{BB'}) = \sum_{\hat{m} \in \mathcal{M}} \text{Tr} \left[\Lambda_{BB'}^{\hat{m}} \tau_{BB'} \right] \times \begin{matrix} | \hat{m} \rangle \langle \hat{m} |_{\hat{M}} \end{matrix}$$

we can also define

$$E_{A' \rightarrow A}^m(\cdot) = E_{M'A'}(|m\rangle \langle m|_{M'} \otimes (\cdot))$$

write the final state as

$$w_{MM}^P = \sum_{m, \hat{m} \in \mathcal{M}} p(m) |m\rangle \langle m|_M \otimes$$

$$\text{Tr} \left[\Lambda_{BB'}^{\hat{m}} N_{A \rightarrow B} (E_{A' \rightarrow A}^m(\Phi_{A'B'})) \right] \cdot \begin{matrix} | \hat{m} \rangle \langle \hat{m} |_{\hat{M}} \end{matrix}$$

Defining

$$q(\hat{m}|m) =$$

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error prob. when sending
message $m \in \mathcal{M}$ then

$$P_{\text{err}}(m) = 1 - q(m|m)$$

(prob. that decoded
message does not
equal transmitted
one)

maximal error prob. of a code
is then

$$P_{\text{err}}^* = \max_{m \in \mathcal{M}} P_{\text{err}}(m)$$

A code is an (M, ϵ) code

if M is # messages

$$P_{\text{err}}^* \leq \epsilon.$$

- can prove that $P_{\text{err}}^* =$

$$\max_{p: \mathcal{M} \rightarrow [0,1]} \frac{1}{2} \left\| \sum_{m \in \mathcal{M}} p_m \bar{\Phi}_{mm}^p - \sum_{m \in \mathcal{M}} p_m \bar{\Phi}_{mm} \right\|_1$$

(4)

one-shot EA capacity:

$$C_{EA}^{\epsilon}(N)$$

$$= \sup_{(M, \Psi, \epsilon, D)} \left\{ \log_2 |M| : p_{\text{err}}^*(\Psi, \epsilon, D); N \leq \epsilon \right\}$$

our 1st goal is to establish

an upper bound on $C_{EA}^{\epsilon}(N)$

Main idea is to ~~substitute~~ substitute

the actual channel $N_{A \rightarrow B}$ a

useless channel $R_{A \rightarrow B}^{\sigma}(\cdot) = \text{Tr}[\cdot] \sigma_B$

† then compare these different protocols using hypothesis testing relative entropy.

(5)

We begin w/ a lemma

$$\text{Let } \bar{\Phi}_{m, m'} = \frac{1}{M} \sum_{m \in M} |m\rangle \langle m|_m \otimes |m\rangle \langle m|_{m'}$$

Define the comparator test

$$\{\Pi_{m, m'}, I_{m, m'} - \Pi_{m, m'}\}$$

where

$$\Pi_{m, m'} = \sum_{m \in M} |m\rangle \langle m|_m \otimes |m\rangle \langle m|_{m'}$$

Suppose that $\omega_{m, m'}$ is a state

~~such~~ such that $\omega_m = \frac{I_m}{M} = \Pi_m$

$$\& \text{Tr}[\Pi_{m, m'} \omega_{m, m'}] \geq 1 - \epsilon.$$

Then

$$\log_2 M \leq I_H^\epsilon(m; m')_\omega$$

where I_H^ϵ is the hypothesis

testing mutual information,

(6)

defined for ρ_{AB} as

$$I_H^\epsilon(A; B)_\rho = \min_{\sigma_B} D_H^\epsilon(\rho_{AB} \| \rho_A \otimes \sigma_B)$$

Proof: By assumption,

$$\text{Tr}[\Pi_{mm'} \omega_{mm'}] \geq 1 - \epsilon$$

Now consider

$$t_{mm'} = \omega_{mM} \otimes \sigma_{m'} = \Pi_{mM} \otimes \sigma_{m'}$$

then

$$\begin{aligned} & \text{Tr}[\Pi_{mm'} t_{mm'}] \\ &= \text{Tr}[\Pi_{mm'} (\omega_{mM} \otimes \sigma_{m'})] \\ &= \frac{1}{|M|} \text{Tr}[\Pi_{mm'} (\mathbb{I}_M \otimes \sigma_{m'})] \\ &= \frac{1}{|M|} \text{Tr}[\text{Tr}_M[\Pi_{mm'}] \sigma_{m'}] \\ &= \frac{1}{|M|} \text{Tr}[\mathbb{I}_{M'} \sigma_{m'}] = \frac{1}{|M|} \end{aligned}$$

(7)

By using the definition of hypothesis testing relative entropy,

$$\begin{aligned}\log_2 |M| &= -\log_2 \text{Tr}[\tau_{M|U}] \\ &\leq D_H^\epsilon(w_{M|U} \| \tau_{M|U}) \\ &= D_H^\epsilon(w_{M|U} \| w_M \otimes \sigma_U)\end{aligned}$$

Inequality holds for arbitrary σ_U

\Rightarrow take infimum over σ_U & apply definition of HTR

to arrive @ claim.

Assumption for an (M, ϵ) protocol:

$$P_{\text{err}}^* \leq \epsilon$$

$$\Rightarrow 1 - \frac{1}{|M|} \sum_m q(m|m) \leq \epsilon$$

\uparrow is ϵ_0 , average error probability

(8)

can show that this implies
that

$$\text{Tr}[\pi_{\mu\hat{\mu}} \omega_{\mu\hat{\mu}}] \geq 1 - \epsilon$$

† by the lemma \uparrow Anal state of protocol.

$$\Rightarrow \log_2 |\mathcal{M}| \leq I_H^\epsilon(\mathcal{M}; \hat{\mu})_\omega$$

establishes a close link

between communication
† hypothesis testing.

Now data processing under
decoding channel to get

$$\begin{aligned} \log_2 |\mathcal{M}| &\leq I_H^\epsilon(\mathcal{M}; \hat{\mu})_\omega \\ &\leq I_H^\epsilon(\mathcal{M}; BB')_\theta \end{aligned}$$

where $\theta_{\mathcal{M}BB'} = \left(\mathcal{N}_{A \rightarrow B} \circ \mathcal{E}_{\mathcal{M}'A \rightarrow A} \right)$
 $\left(\mathbb{P}_{\mathcal{M}'B} \otimes \mathbb{P}_{A'B'} \right)$

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Consider that

$$\begin{aligned}\Theta_{MB'} &= \text{Tr}_B \left[(N_{A \rightarrow B} \circ E_{M'A \rightarrow A}) (\bar{\Phi}_{MM'} \otimes \Phi_{A'B'}) \right] \\ &= \text{Tr}_{M'A'} \left[\bar{\Phi}_{MM'} \otimes \Phi_{A'B'} \right] \\ &= \bar{\Phi}_M \otimes \Phi_{B'} \\ &= \Theta_M \otimes \Theta_{B'}\end{aligned}$$

$$\Rightarrow I_H^\epsilon(M; BB')_\Theta$$

$$= \inf_{\sigma_{BB'}} D_H^\epsilon(\Theta_{M BB'} \| \Theta_M \otimes \sigma_{BB'})$$

$$\leq \inf_{\sigma_B} D_H^\epsilon(\Theta_{M BB'} \| \Theta_M \otimes \sigma_B \otimes \Theta_{B'})$$

$$= \text{" } D_H^\epsilon(\Theta_{M BB'} \| \Theta_{MB'} \otimes \sigma_B)$$

$$= I_H^\epsilon(MB'; B)_\Theta$$

$$\leq \sup_{P_{SA}} I_H^\epsilon(S; B)_{\Theta_S}$$

where $\Theta_S = N_{A \rightarrow B}(P_{SA})$

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subtrees to optimize over
pure states ψ_{SA} w/ $S \simeq A$

$$\Rightarrow \log_2 |M| \leq I_H^E(N) := \sup_{\psi_{SA}} I_H^E(S; B)_{\mathcal{G}}$$

where $\mathcal{G}_{SB} = \mathcal{N}_{A \rightarrow B}(\psi_{SA})$

\Rightarrow final bound ^{on} one-shot EA capacity

$$C_{EA}^E(N) \leq I_H^E(N)$$

\uparrow
one-shot mutual information
of channel.

By applying

$$D_H^E(p||q) \leq D_\alpha(p||q) + \frac{\alpha}{\alpha-1} \log_2 \left(\frac{1}{1-\epsilon} \right)$$

\Rightarrow

$$C_{EA}^E(N) \leq \tilde{I}_\alpha(N) + \frac{\alpha}{\alpha-1} \log_2 \left(\frac{1}{1-\epsilon} \right)$$

(11)

By using

$$D_H^\epsilon(p||\sigma) \leq \frac{1}{1-\epsilon} (D(p||\sigma) + h_2(\epsilon) + \epsilon \log_2 \text{Tr}[\sigma])$$

we also get

$$C_{EA}^\epsilon(N) \leq \frac{1}{1-\epsilon} (I(N) + h_2(\epsilon))$$

where $I(N) = \sup_{\Psi_{RA}} I(R; B)_\omega$

w/ $W_{FB} = N_{A \rightarrow B}(\Psi_{RA})$

Asymptotic capacity is defined as

$$C_{EA}(N) = \inf_{\epsilon \in (0,1)} \liminf_{n \rightarrow \infty} \frac{1}{n} C_{EA}^\epsilon(N^{*n})$$

& strong converse capacity as

$$\tilde{C}_{EA}(N) = \sup_{\epsilon \in (0,1)} \limsup_{n \rightarrow \infty} \frac{1}{n} C_{EA}^\epsilon(N^{*n})$$

(12)

$C_{EA}(N) \leq \tilde{C}_{EA}(N)$ always holds

Let us 1st bound

$C_{EA}(N)$ from above by $I(N)$

using the weak-converse bound we have that

$$\frac{1}{n} C_{EA}^{\epsilon}(N_{\otimes n}) \leq \frac{1}{1-\epsilon} \left(\frac{I(N_{\otimes n})}{n} + \frac{\ln 2(\epsilon)}{n} \right)$$

Let us prove that $\frac{I(N_{\otimes n})}{n} = I(N)$

Follows from

$$I(N_{\otimes n}) = I(N) + I(n)$$

+ induction

(13)

We always have $I(N \otimes M) \geq I(N) + I(M)$

Now for opposite inequality

$$I(N \otimes M) = \sup_{\Psi_{R, A_1, A_2}} I(R; B_1, B_2)_\omega$$

$$\omega_{R, B_1, B_2} = (N_{A_1 \rightarrow B_1} \otimes M_{A_2 \rightarrow B_2}) (\Psi_{R, A_1, A_2})$$

Consider that

$$I(R; B_1, B_2)_\omega =$$

$$I(R; B_1)_\omega + I(R; B_2 | B_1)_\omega$$

$$\leq I(R; B_1)_\omega + I(R, B_1; B_2)_\omega$$

$$\leq I(N) + I(M)$$

Finishing off we get

$$\frac{1}{n} \left(\sum_{\Psi_{R, A_1, A_2}} I(R; B_1, B_2)_\omega \right) \leq \frac{1}{1-\epsilon} \left(I(N) + \frac{h_2(\epsilon)}{n} \right)$$

then ~~as~~ $n \rightarrow \infty$ & $\epsilon \rightarrow 0$ gives upper bound

(14)

can also get the strong converse
by using previous upper bound
& the fact that

$$\tilde{I}_\alpha(N \otimes M) = \tilde{I}_\alpha(N) + \tilde{I}_\alpha(M)$$

$$\forall \alpha > 1$$