Quantum Information Theory and Applications to Local Decoding

Sayak Chakrabarti

Topics

• Quantum Information Theory

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- Local Decoding

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- Local Decoding
- Random Access Codes

Quantum Information Theory: Mixed States

States given by

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- The concept of Quantum noise
- Implementing a Quantum System,

```
 \text{the device outputs} \left\{ \begin{array}{l} |\psi_1\rangle & \text{with probability } p_1 \\ |\psi_2\rangle & \text{with probability } p_2 \\ & \vdots \\ |\psi_d\rangle & \text{with probability } p_d \end{array} \right.
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- Mixed state through unitary transformation U: $U\rho U^{\dagger}$
- Can be written as probability distribution over orthogonal pure states using SVD

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Mixed states might not always be distinguishable.

Measurements

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- Projective measurements: M_i 's chosen as projector matrices

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- How many bits does a qubit represent?

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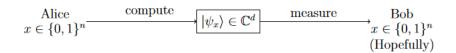
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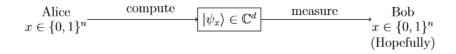
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$$x \in \{0,1\}^n \xrightarrow{\text{compute}} |\psi_x\rangle \in \mathbb{C}^d \xrightarrow{\text{measure}} \text{Bob} \\ x \in \{0,1\}^n \\ \text{(Hopefully)}$$

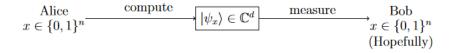
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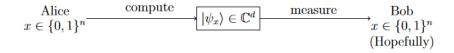
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Theorem ([Hol73])

For measurement given by POVM $E_Y = \{E_1, E_2, \dots E_n\}$ performed on ρ with measurement outcome Y, the amount of information about X possible to find from this measurement is given by

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 χ : Holevo's information

[CvDNT98] gave the following interpretation of Holevo's theorem that is more commonly used.

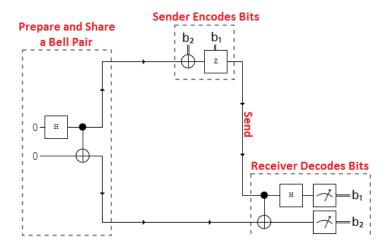
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Theorem ([CvDNT98])

Suppose Alice wants to communicate a string to Bob. Then the following holds.

- If Alice sents m qubits to Bob, and they do not have any prior entangled state, then Bob receives at most m bits of information about x.
- If Alice sents m qubits to Bob, and they do have some prior entangled state, then Bob receives at most 2m bits of information about x.
- If Alice sents m classical bits to Bob, and even if they have some prior entangled state, then Bob receives at most m bits of information about x.

Superdense coding



- Let $X \equiv U_N$, having values from [N]
- Encoding $E: x \in X \mapsto \rho_x$, some *d*-dimensional density matrix.
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• Encode n uniformly random bits into m qubits, success probability after decoding is $\frac{2^m}{2^n}$

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- For each $x \in \{0,1\}^n$, we want $Tr(M_i\rho_x) \ge p$ for $x_i = 1$ and $Tr(M_i\rho_x) \le 1 p$ for $x_i = 0$.

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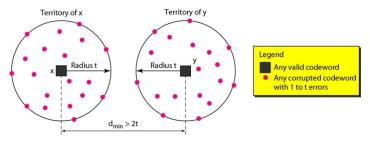
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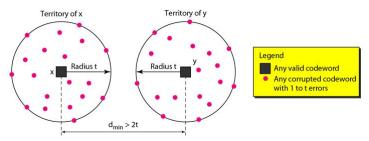
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- Nayak [Nay99] tightened this by showing $m \ge (1 H(p))n$.

• Error correcting codes

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Definition (Local Decoding)

A map $E: \{0,1\}^n \to \{0,1\}^m$ is called an (q,δ,ϵ) -LDC if there exists a classical randomized decoding algorithm D satisfying the properties:

- For each $x \in \{0,1\}^n$ and $\forall y \in \{0,1\}^m$, $Ham(E(x),y) \le \delta n$, we have $\forall i \in [n], Pr[D(y,i) = x_i] \ge \frac{1}{2} + \epsilon$
- D makes at most non-adaptive q queries to y



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Walsh-Hadamard Code

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Example
$$n = 2$$
. For $x = 01$, $E(x) = (x \cdot 00, x \cdot 01, x \cdot 10, x \cdot 11) = 0101$.

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- Now, we have $E(x)_z$ and $E(x)_{z \oplus e_i}$. From this,

$$E(x)_z \oplus E(x)_{z \oplus e_i} = (x \cdot z) \oplus (x \cdot (z \oplus e_i)) = x \cdot e_i = x_i$$

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• WH is $(2, \delta, \frac{1}{2} - 2\delta)$ -LDC.

Local Decoding: Lower Bounds

Theorem ([<mark>KT00</mark>])

For every (q, δ, ϵ) -LDC with encoding $E : \{0,1\}^n \to \{0,1\}^m$, and each $i \in [n]$, there exists a set \mathcal{M}_i of $\Omega(\delta \epsilon m/q^2)$ -many disjoint tuples. Each tuple $t \in \mathcal{M}_i$ consist of q indices from [m], and a bit $a_{i,t}$ such that

$$Pr_{x \in \{0,1\}^n} \left[x_i = a_{i,t} \oplus \sum_{j \in t} E(x)_j \right] \geq \frac{1}{2} + \Omega(\frac{\epsilon}{2^q})$$

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Probability boosted by enumerating over t's randomly.

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- Put them together! $N \ge 2^{\Omega(\delta^2 \epsilon^4 n)}$



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- State $P_{jk} | \phi_x \rangle$ with probability $\frac{2}{m} \forall (j,k) \in \mathcal{M}_i$.

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- State $P_{jk} |\phi_x\rangle$ with probability $\frac{2}{m} \, \forall (j,k) \in \mathcal{M}_i$.
- $|\mathcal{M}_i| = \Omega(\delta \epsilon m)$, probability $|\mathcal{M}_i| \times \frac{2}{m} = \Omega(\delta \epsilon)$.

- How to recover x_i from $|\phi_x\rangle$?
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$$\frac{(-1)^{E(x)_j}}{\sqrt{2}}(\ket{j}+(-1)^{E(x)_j\oplus E(x)_k}\ket{k})$$

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- Success probability $p \geq \frac{r}{2} + \left(\frac{1}{2} + \Omega(\epsilon)\right)(1-r) = \frac{1}{2} + \Omega(\delta\epsilon^2)$



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Thank You!