

Zero-error source-channel coding with entanglement



QIP 2014

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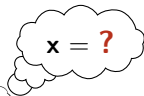
Does entanglement help in zero-error communication between two parties with

- one-way classical noisy **channel**
- side information from a dual **source**



The channel coding problem

$$\mathbf{x} \in \{0, 1\}^m$$



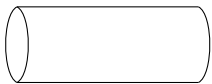
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$$C(\mathbf{x}) = \mathbf{s}$$

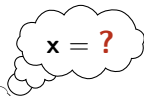
\mathbf{s}



\mathbf{v}



$$D(\mathbf{v}) = \mathbf{y}$$



$$s_1 \bullet \text{-----} \bullet v_1$$

$$s_2 \bullet \text{-----} \bullet v_2$$

$$s_3 \bullet \text{-----} \bullet v_3$$

$$s_4 \bullet \text{-----} \bullet v_4$$

$$s_5 \bullet \text{-----} \bullet v_5$$

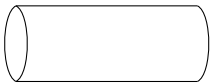
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s

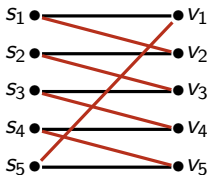


v

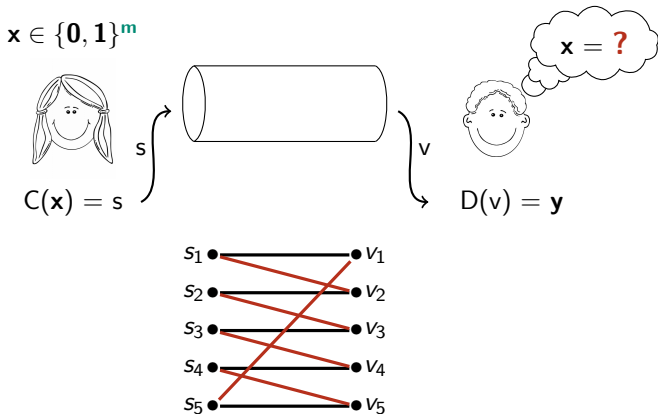


$$D(v) = \mathbf{y}$$

$$\mathbf{x} = ?$$



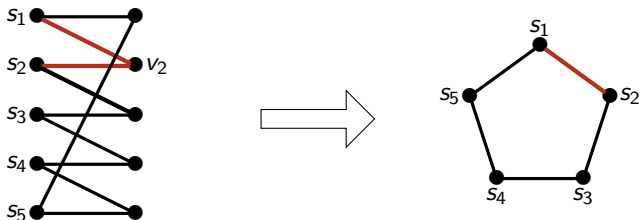
The channel coding problem



- **Goal:** $y = x$ with zero probability of error and **maximize** m

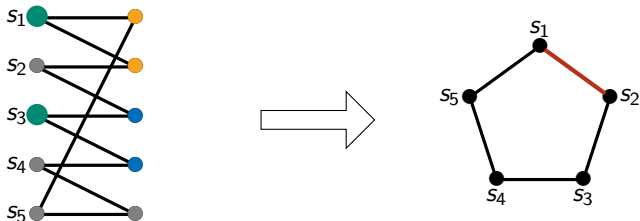
The confusability graph of a channel

- The **confusability graph** of a noisy channel has as vertices the channel's inputs, which are adjacent if they are confusable



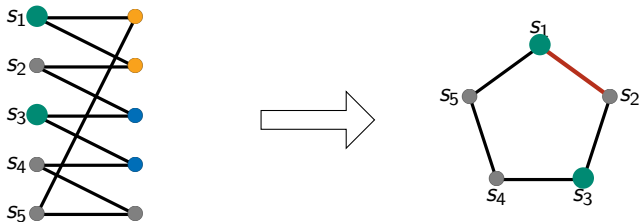
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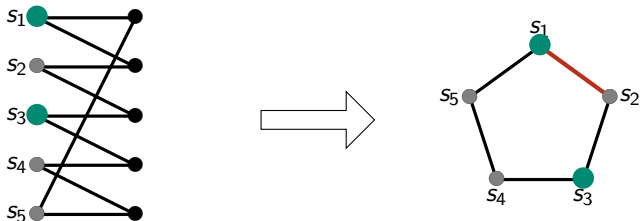
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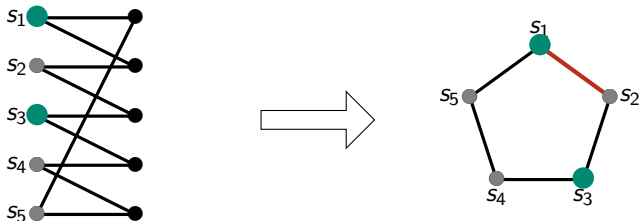
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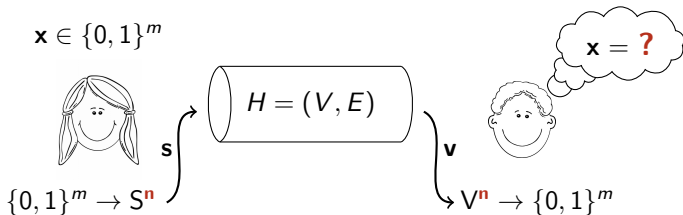
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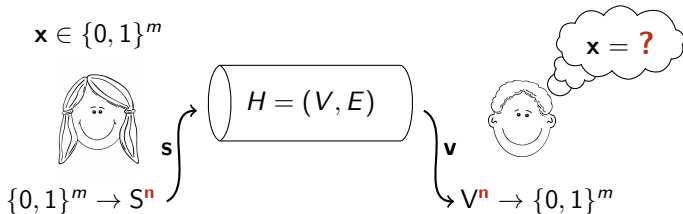
- An independent set is a set of pairwise non-adjacent vertices
- The **independence number** $\alpha(H)$ of a graph H is the size of a largest possible independent set
- $\alpha(H)$ is the maximum number of distinct messages Alice can send to Bob

Block coding and the Shannon capacity



- Encoding \mathbf{x} into a **sequence** of channel inputs can be more efficient

Block coding and the Shannon capacity

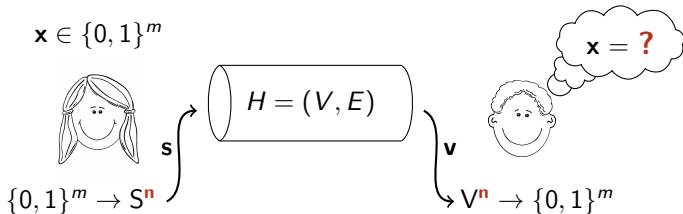


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$$c(H) = \lim_{n \rightarrow \infty} \frac{1}{n} \log \alpha(H^{\boxtimes n})$$

gives the maximum average number of bits that can be sent per channel use

Block coding and the Shannon capacity



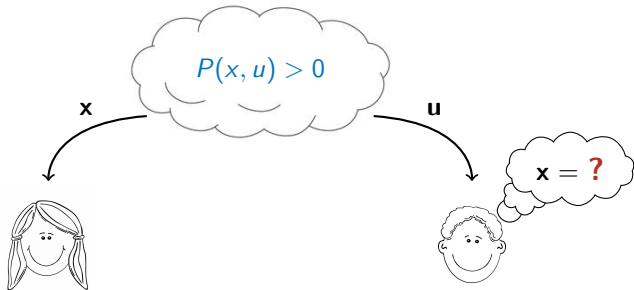
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- The Shannon capacity lead to many interesting developments in combinatorics: e.g., perfect graphs, semidefinite optimization

The source coding problem



$x_1 \bullet$

$x_2 \bullet$

$x_3 \bullet$

$x_4 \bullet$

$x_5 \bullet$

$\bullet u_1$

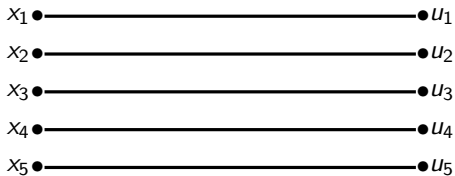
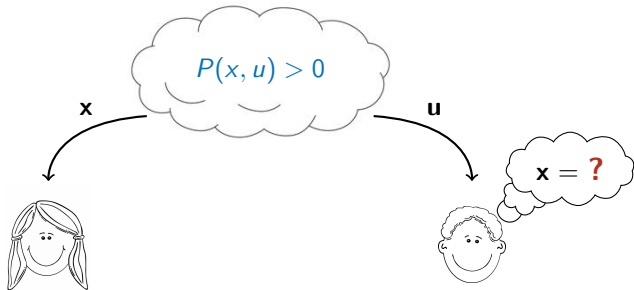
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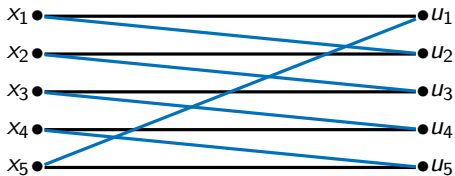
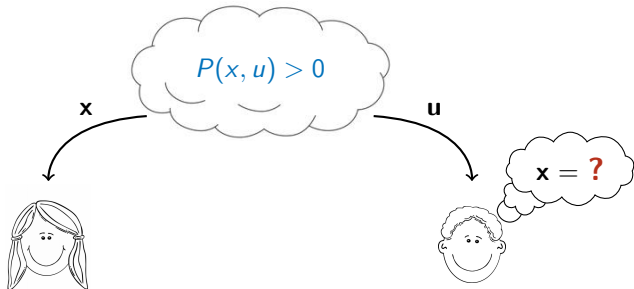
$\bullet u_4$

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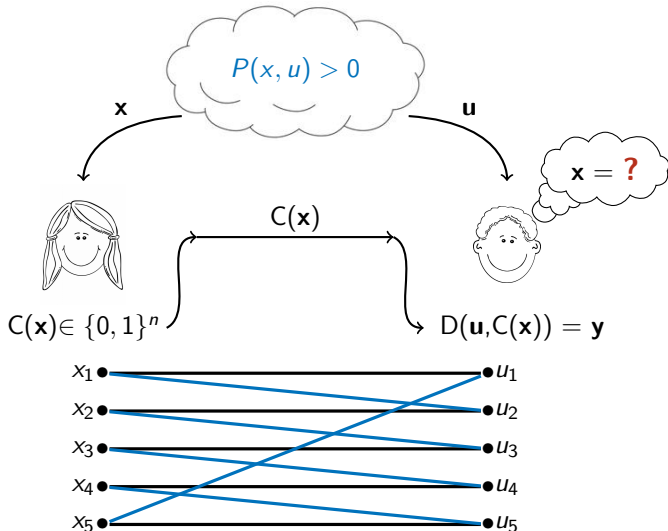
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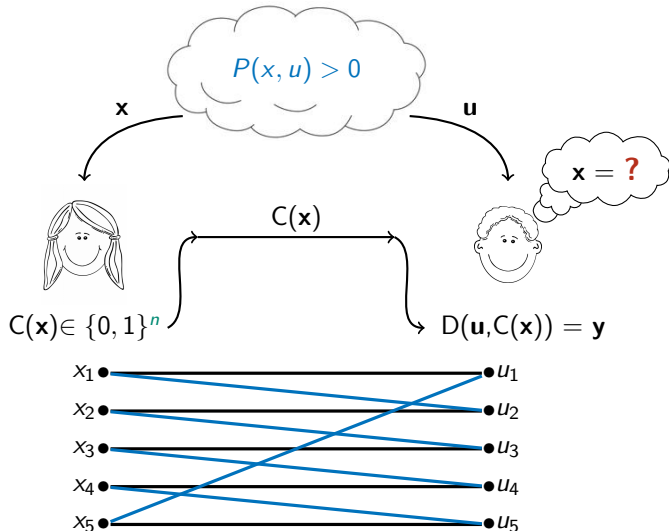
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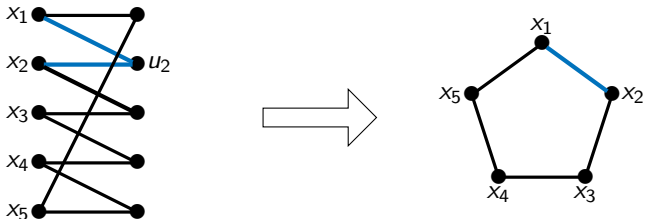
The source coding problem



- **Goal:** $y = x$ with zero probability of error and minimize n

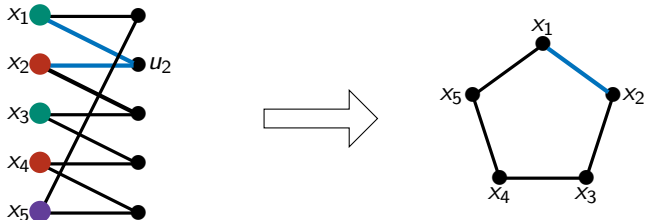
The characteristic graph of a dual source

- The **characteristic graph** of a dual source has as vertices Alice's inputs, which are adjacent if they are confusable to Bob



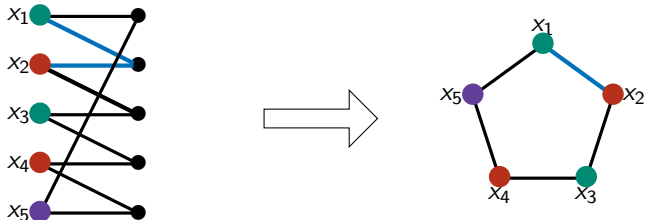
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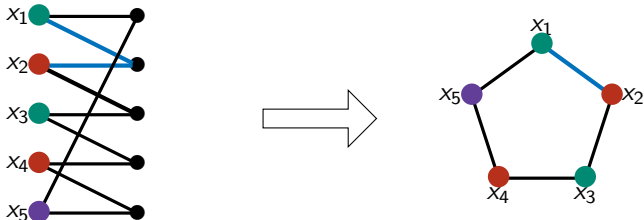
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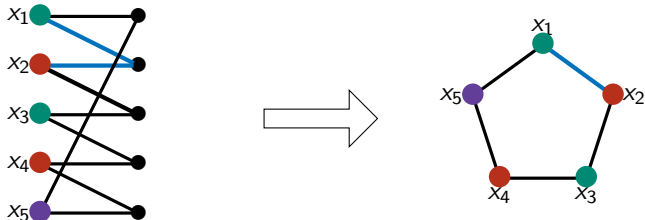
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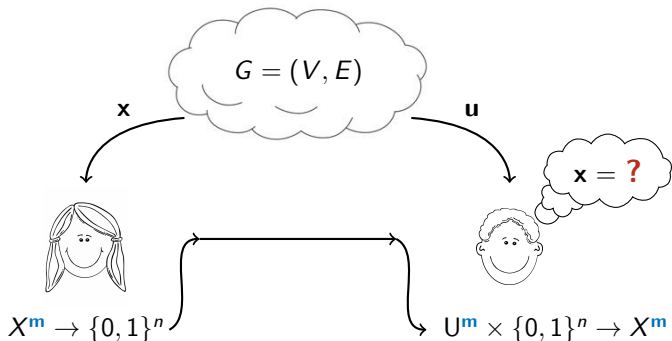
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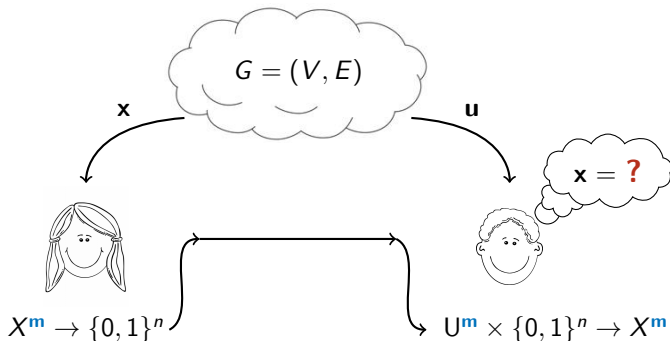
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- $\chi(G)$ is the minimum size of the message set that Alice must use

Block coding and the **Witsenhausen rate**



- Jointly coloring input **sequences** can be more efficient

Block coding and the **Witsenhausen rate**

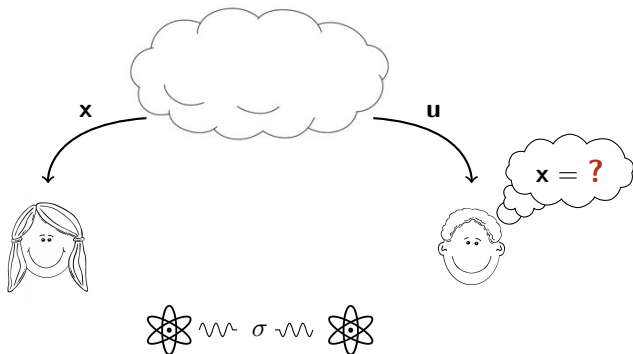


- Jointly coloring input **sequences** can be more efficient
- The **Witsenhausen rate** of G

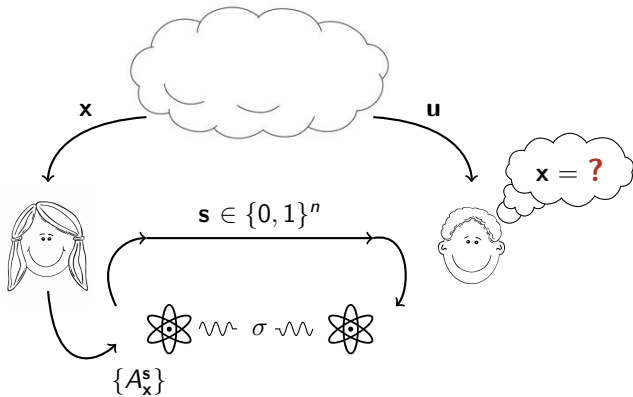
$$R(G) = \lim_{m \rightarrow \infty} \frac{1}{m} \log \chi(G^{\boxtimes m})$$

gives the minimum average number of bits needed per source input

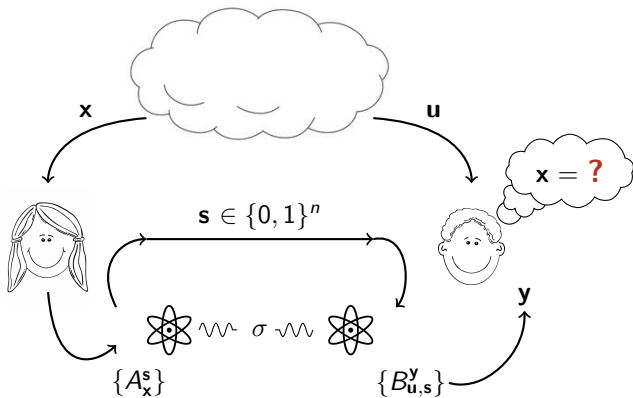
The source coding problem with entanglement



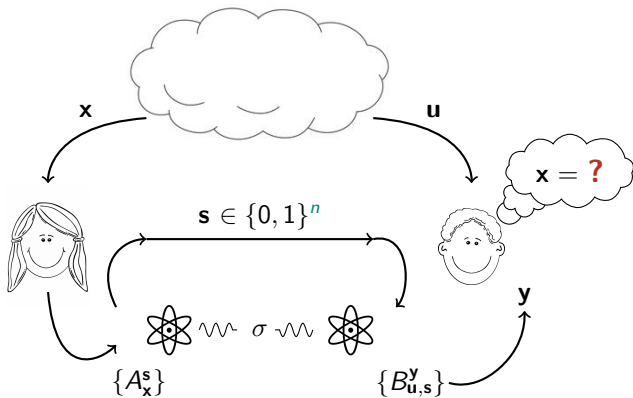
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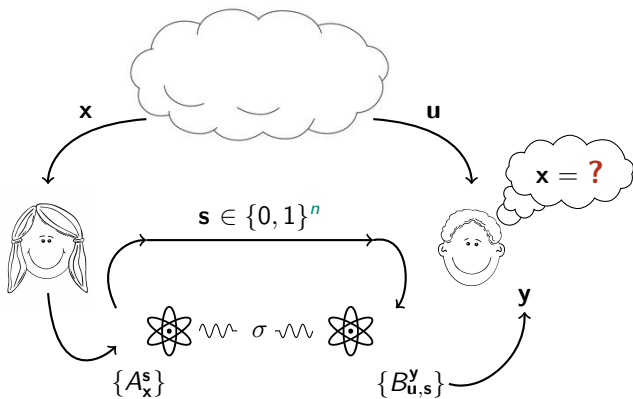


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The source coding problem with entanglement



- As for the classical case, the goal is to have $y = x$ with zero probability of error and **minimize n**
- The **entangled Witsenhausen rate R^*** depends on the characteristic graph G and can be defined by simple constraints on σ , $\{A_x^s\}$

Entangled Shannon capacity

The **entangled Shannon capacity** c^* of a graph H is defined as

$$c^*(H) = \lim_{n \rightarrow \infty} \frac{1}{n} \log \alpha^*(H^{\boxtimes n})$$

- [Cubitt et al. '10] introduced α^* , c^* and showed that entanglement can increase that number of possible messages that can be sent with one use of the **channel** (i.e. $\alpha < \alpha^*$)
- [Leung et al. '12] and [Briët et al. '12] showed that entanglement can increase the capacity of a **channel** (i.e. $c < c^*$ by a constant factor)

Lovász ϑ number

$$\begin{aligned} \vartheta(G) = \min \quad & \lambda \in \mathbb{R} \\ \text{such that} \quad & \exists \text{ PSD matrix } Z \in \mathbb{R}^{V \times V} \\ & Z(u, u) = \lambda - 1 \text{ for all } u \in V \\ & Z(u, v) = -1 \text{ for all } \{u, v\} \notin E \end{aligned}$$

- Introduced by Lovász ['79] to compute $c(C_5)$
- ϑ can be computed efficiently (up to any approximation)
- $c(G) \leq \log \vartheta(G) \leq R(\overline{G})$ ([Lovász '79] and [Nayak et al. '06])
- [Beigi '10] and [Duan et al. '13] proved $c^*(H) \leq \log \vartheta(H)$

ϑ bound on the entangled Witsenhausen rate

Theorem

$$\log \vartheta(G) \leq R^*(\overline{G})$$

- Thus

$$c(G) \leq c^*(G) \leq \log \vartheta(G) \leq R^*(\overline{G}) \leq R(\overline{G})$$

Separation between the classical and entangled Witsenhausen rate

Theorem

There exists an infinite family of graphs H_k such that

$$\frac{R^*(H_k)}{R(H_k)} \leq O\left(\frac{\log k}{k}\right).$$

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- The *orthogonality graph* (a.k.a. Hadamard graph) has vertex set $\{\pm 1\}^k$ and two vertices are adjacent if and only if they are orthogonal
- The *quarter orthogonality graph* H_k is a subgraph of the orthogonality graph induced by the vertices $x \in \{\pm 1\}^k$ with $x_1 = +1$ and an even number of -1 's

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Let $k = 4p^\ell - 1$ where p is an odd prime and $\ell \in \mathbb{N}$, then

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$$\frac{R^*(H_k)}{R(H_k)} \leq O\left(\frac{\log k}{k}\right).$$

- The lower bound on $R(H_k)$ is derived from a technique that upper bounds $c(H_k)$. It is obtained using an instance of the linear algebra method due to [Alon '98] with a construction of certain low-degree polynomials over finite field due to [Barrington et al. '94]

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- The upper bound on $R^*(H_k)$ relies on the construction of a orthogonal representation of the graph H_k (similar idea as used by [Cameron et al. '07])

Separation between the classical and entangled Shannon capacity

Theorem

Let $k = 4p^\ell - 1$ where p is an odd prime and $\ell \in \mathbb{N}$, then

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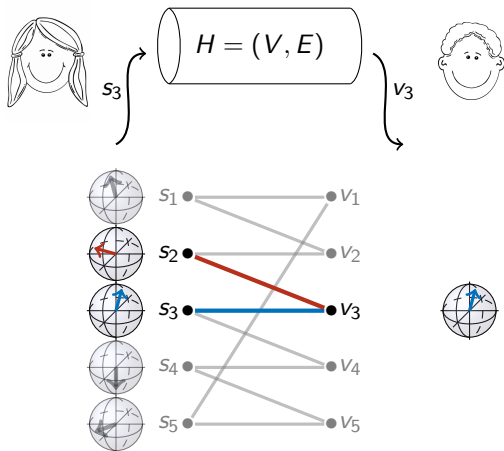
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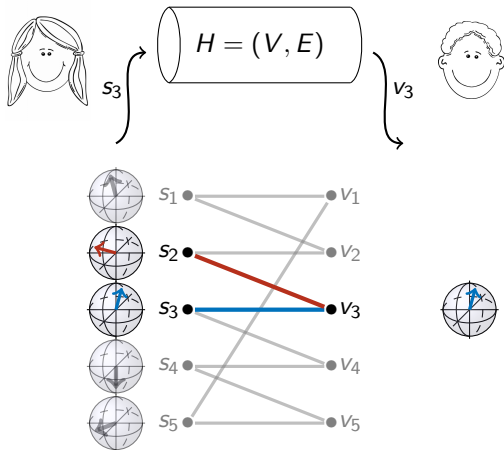
- The upper bound on $c(H_k)$ is obtained using an instance of the linear algebra method due to [Alon '98] (*as before*)
- The lower bound on $c^*(H_k)$ uses a new method based on the teleportation scheme of [Bennet et al. '93]

Lower bound on entangled Shannon capacity



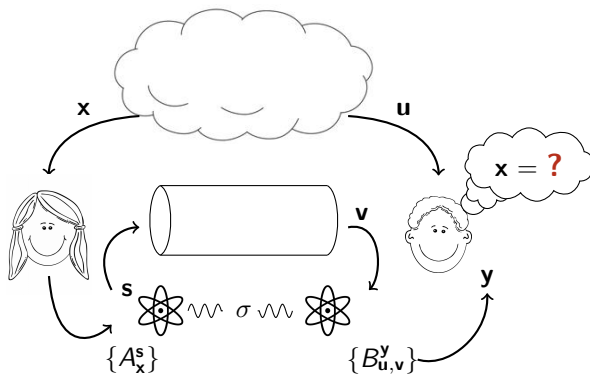
- Main idea: Teleport a state of an orthonormal representation to Bob

Lower bound on entangled Shannon capacity



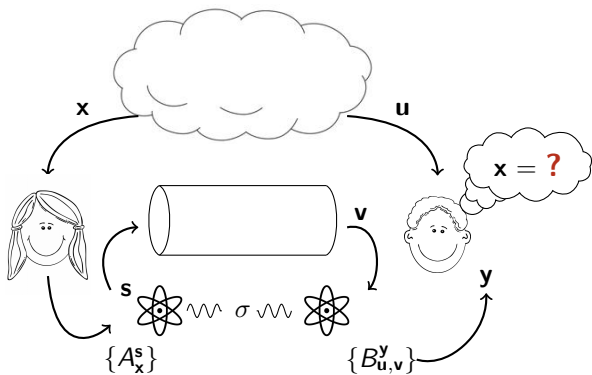
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- Send $|V|^t$ messages in $t + 1$ steps if $\log \alpha(H) \geq 2t \log \text{orthdim}(H)$

Further results



- We study the zero-error **source-channel** coding problem with entanglement, a generalization of the zero-error **channel** coding and of the **source** coding with entanglement

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- We study the zero-error **source-channel** coding problem with entanglement, a generalization of the zero-error **channel** coding and of the **source** coding with entanglement
- We present an infinite family of source and channels combinations for which entanglement allows an exponential saving in communication in zero-error **source-channel** coding

Conclusions and open questions

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arXiv:1308.4283

Thank you!