

# Constant Compositions in the Sphere Packing Bound for Classical-Quantum Channels

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- $\mathcal{X}$  discrete set of input symbols
- Associated (finite dimensional) density operators  $S_x, x \in \mathcal{X}$
- $n$ -fold product channel:

$$\begin{array}{ccc} \text{Codeword} & & \text{Product State} \\ \mathbf{x} = (x_1, x_2, \dots, x_n) & \mapsto & \mathbf{S}_{\mathbf{x}} = S_{x_1} \otimes S_{x_2} \cdots \otimes S_{x_n} \end{array}$$

- Code: a set of  $M$  codewords  $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_M\} \subset \mathcal{X}^n$
- Quantum decision scheme: a set of  $M$  positive operators  $\mathbf{\Pi} = \{\Pi_1, \Pi_2, \dots, \Pi_M\}$  with  $\sum_i \Pi_i \leq \mathbb{1}$
- Rate

$$R = \frac{\log M}{n}$$

- Probability of error

$$P_{e|m} = 1 - \text{Tr}(\Pi_m \mathbf{S} \mathbf{x}_m), \quad P_{e,\max} = \max_m P_{e|m}$$

- Optimal probability of error

$$P_{e,\max}^{(n)}(R) = \min_{\{\mathbf{x}_m\}, \Pi} P_{e,\max}$$

(min. over coding schemes of block-length  $n$  and rate at least  $R$ ).

- Reliability function

$$E(R) = \limsup_{n \rightarrow \infty} -\frac{1}{n} \log P_{e,\max}^{(n)}(R) \quad (\text{i.e., } P_e \approx e^{-nE(R)})$$

- Zero-error capacity

$$C_0 = \sup\{R : P_{e,\max}^{(n)}(R) = 0 \text{ for some } n\} \quad (1)$$

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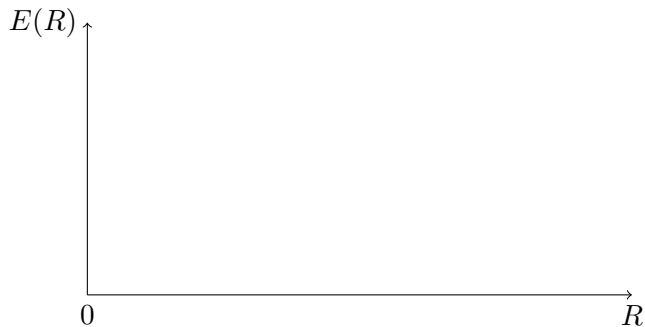
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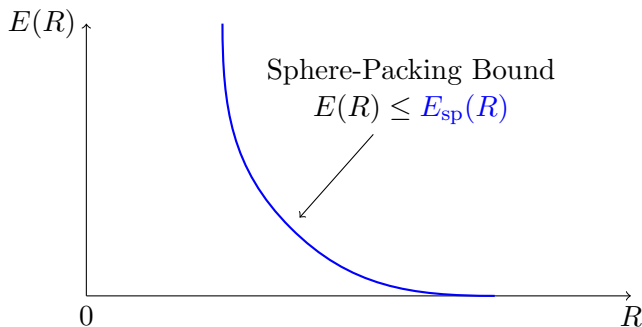
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# Classical-Quantum Channels: Reliability Function



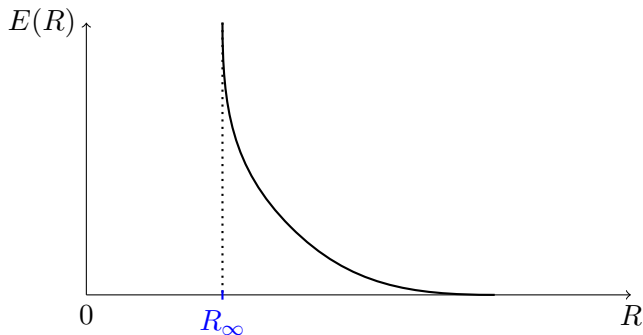
# Classical-Quantum Channels: Reliability Function



Dalai, ISIT 2012 (extending S-G-B 1967)

$$E_{sp}(R) = \sup_{\rho \geq 0} \max_P \left( \overbrace{\left[ -\log \text{Tr} \left( \sum_x P(x) S_x^{1/(1+\rho)} \right)^{1+\rho} \right]}^{E_0(\rho, P)} - \rho R \right)$$





Upper bound on the zero-error capacity  $C_0$

$$C_0 \leq R_\infty$$

where

$$R_\infty = \min_F \max_x \log \frac{1}{\text{Tr}(S_x^0 F)}$$

## Sphere Packing and the Lovász theta Function

- Given a confusability graph  $G$ , we can optimize over channels

$$\vartheta_{\text{sp}}(G) = \min_{\{S_x\}} R_{\infty}(\{S_x\})$$

over  $\{S_x\}$  with confusability graph  $G$ .

- Note: consider pure  $S_x$  and  $F$

$$\begin{aligned}\vartheta_{\text{sp}}(G) &= \min_{\{S_x\}} \min_F \max_x \log \frac{1}{\text{Tr}(S_x^0 F)} \\ &\leq \min_{\{u_x\}} \min_f \max_x \log \frac{1}{|\langle u_x | f \rangle|^2} \\ &= \vartheta(G)\end{aligned}$$

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- All codewords with the same composition  $P_n$
- $P_{e,\max}^{(n)}(R, P_n)$  as before but with c.c.  $P_n$
- Reliability function with c.c.

$$E(R, P) = \limsup_{n \rightarrow \infty} -\frac{1}{n} \log P_{e,\max}^{(n)}(R, P_n) \quad (\text{where } P_n \rightarrow P)$$

- Zero-error capacity

$$C_0(P) = \inf\{R : E(R, P) < \infty\}$$

- Questions
  - Bounds on  $E(R, P)$ ?
  - Bounds on  $C_0(P)$ ?

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- For classical channels (commuting  $S_x$ )

$$E_{\text{sp}}(R) = \max_P \sup_{\rho \geq 0} \left[ -\log \text{Tr} \left( \sum_x P(x) S_x^{1/(1+\rho)} \right)^{1+\rho} - \rho R \right]$$

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$$= \max_P \min_{V: I(P,V) \leq R} \sum_x P(x) \text{Tr}[V_x (\log V_x - \log S_x)]$$

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 &= \max_P \sup_{\rho \geq 0} \left[ \min_F -(1+\rho) \sum_x P(x) \log \text{Tr} (S_x^{\frac{1}{1+\rho}} F^{\frac{\rho}{1+\rho}}) - \rho R \right] \\
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Only for commuting  $S_x$

$E_{\text{sp}}^{\text{cc}}(R, P)$   
for non commuting  $S_x$

## Theorem

For  $R > 0$ ,  $0 < \varepsilon < R$ ,

$$E(R, P) \leq E_{sp}^{cc}(R - \varepsilon, P),$$

where

$$E_{sp}^{cc}(R, P) = \sup_{\rho \geq 0} [E_0^{cc}(\rho, P) - \rho R],$$

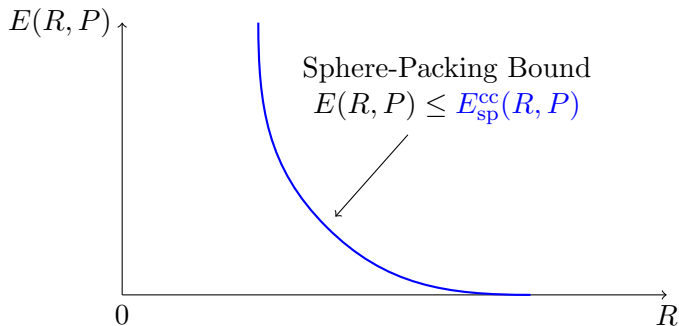
$$E_0^{cc}(\rho, P) = \min_F \left[ -(1 + \rho) \sum_x P(x) \log \text{Tr}(S_x^{\frac{1}{1+\rho}} F^{\frac{\rho}{1+\rho}}) \right].$$

the minimum being over all density operators  $F$ .

# Sphere Packing for Constant Composition Codes



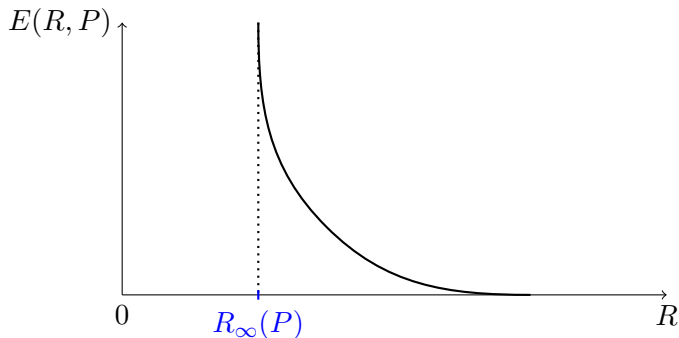
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**This paper**

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# Sphere Packing for Constant Composition Codes



Then

$$C_0(P) \leq R_\infty(P)$$

where

$$R_\infty(P) = \min_F \left[ - \sum_x P(x) \log \text{Tr}(S_x^0 F) \right]$$

- Optimize over channels with confusability graph  $G$

$$\begin{aligned} \vartheta_{\text{sp}}(G, P) &= \min_{\{S_x\}} R_\infty(\{S_x\}, P) \\ &= \min_{\{S_x\}} \min_F \sum_x P(x) \log \frac{1}{\text{Tr}(S_x^0 F)}, \end{aligned}$$

- Using only pure states we get

$$\begin{aligned} \vartheta_{\text{sp}}(G, P) &\leq \min_{\{u_x\}} \min_f \sum_x P(x) \log \frac{1}{|\langle u_x | f \rangle|^2} \\ &= \vartheta(G, P) \end{aligned}$$

where  $\vartheta(G, P)$  is Marton's variation of the Lovász theta function.

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- Bounding  $C_0$ :

$$C_0 \leq \vartheta(G) \quad (\text{Lovász}) \quad (3)$$

and

$$\begin{aligned} C_0 &= \max_P C_0(P) \\ &\leq \max_P \vartheta(G, P) \quad (\text{Marton}) \end{aligned}$$

- Marton: “it is easily seen that”  $\max_P \vartheta(G, P) = \vartheta(G)$   
... perhaps not really easily

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$$\max_P \vartheta(G, P) = \vartheta(G) \quad (4)$$

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# Conditional Constant Composition

- Assume we have a finite set of states  $\mathcal{A}$  and different channels  $\{\mathcal{C}_a\}_{a \in \mathcal{A}}$
- Transmission governed by a sequence of states  $\mathbf{a} = (a_1, \dots, a_n)$  with composition  $P$
- In a state  $a \in \mathcal{A}$ , channel  $\mathcal{C}_a$  is used (hence, channel  $a$  is used  $nP(a)$  times)
- Codewords with conditional composition  $V$  given  $\mathbf{a}$  (each codeword has symbol  $x$  in a fraction  $V(x|a)$  of the  $nP(a)$  positions where  $a_i = a$ )
- For this scenario we can define an error exponent

$$E(\{\mathcal{C}_a\}, R, P, V)$$



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## Theorem

*We have*

$$E(\{\mathcal{C}_a\}, R, P, V) \leq E_{sp}^{cc}(\{\mathcal{C}_a\}, R - \varepsilon, P, V), \quad (5)$$

*where  $E_{sp}^{cc}(\{\mathcal{C}_a\}, R, P, V)$  is defined by*

$$E_{sp}^{cc}(\{\mathcal{C}_a\}, R, P, V) = \sup_{\rho \geq 0} [E_0^{cc}(\{\mathcal{C}_a\}, \rho, P, V) - \rho R], \quad (6)$$

$$E_0^{cc}(\{\mathcal{C}_a\}, \rho, P, V) = \sum_a P(a) E_0^{cc}(\mathcal{C}_a, \rho, V(\cdot|a)). \quad (7)$$

*and  $E_0^{cc}(\mathcal{C}_a, \rho, V(\cdot|a))$  is the coefficient  $E_0^{cc}$  of the sphere packing bound for channel  $\mathcal{C}_a$  with composition  $V(\cdot|a)$ .*

## An application

- Take a code with  $M = e^{nR}$  codewords of composition  $P$
- Given a matrix  $V$ , we can identify a subset  $\mathcal{T}$  of codewords with conditional composition  $V$  from a particular sequence  $\bar{x}$  where  $|\mathcal{T}| = e^{n(R-I(P,V)-o(1))} = e^{nR'}$ .
- We can use the sphere packing bound at rate  $R'$  for these codewords when used on auxiliary channels; channel  $\tilde{\mathcal{C}}_{\bar{x}_i}$  at time instant  $i$ .
- Example for pure states: if  $S_x = |\psi_x\rangle\langle\psi_x|$ , for some  $\rho \geq 1$  choose  $\tilde{S}_{\bar{x}_i,x} = |\tilde{\psi}_{\bar{x}_i,x}\rangle\langle\tilde{\psi}_{\bar{x}_i,x}|$  such that  $|\langle\tilde{\psi}_{\bar{x}_i,x}|\psi_{\bar{x}_i,x'}\rangle| \leq |\langle\psi_x|\psi_{x'}\rangle|^{1/\rho}$ .

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$$E_{spu}^{cc}(R, P) = \inf \rho [E_{sp}^{cc}(\{\tilde{\mathcal{C}}_a\}, R - I(P, V) - \varepsilon, P, V) + R - I(P, V)],$$

over  $\varepsilon > 0$ ,  $\rho \geq 1$ ,  $\{\tilde{\mathcal{C}}_a\}$ ,  $V$ .

- It includes the previous “umbrella” bounds (Dalai ISIT 2013, TIT 2013)
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