

# Zero-error communication via quantum channels

L0

Andreas Winter (Bristol/Singapore)

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[with T.S. Cubitt, D. Loung, W. Matthews      0911.5300  
    1003. 3195]

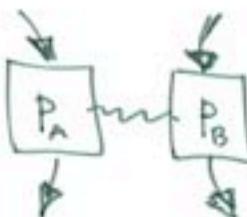
and with S. Severini, R. Duan      1002. 2514

## Outline

[0']

I. Classical channels : Shannon & O-error

II. Assistance by non-local correlations :  $| \phi \rangle^{AB}$  &

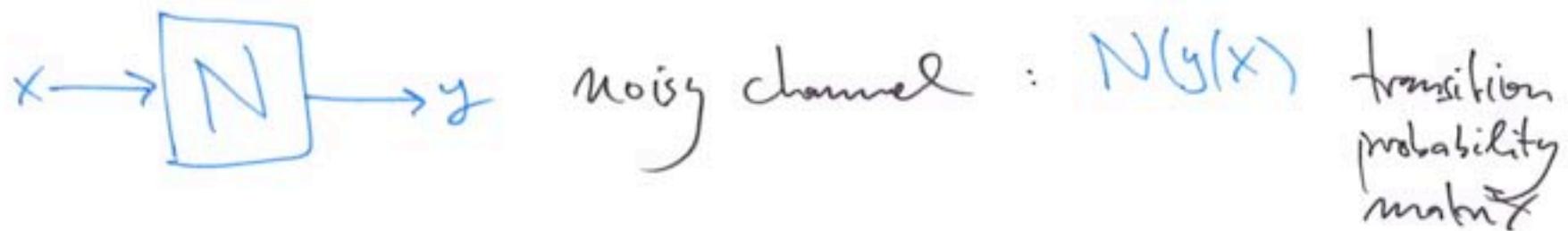


III. Quantum channels : superactivation & Lovász- $\vartheta$

IV. Open questions

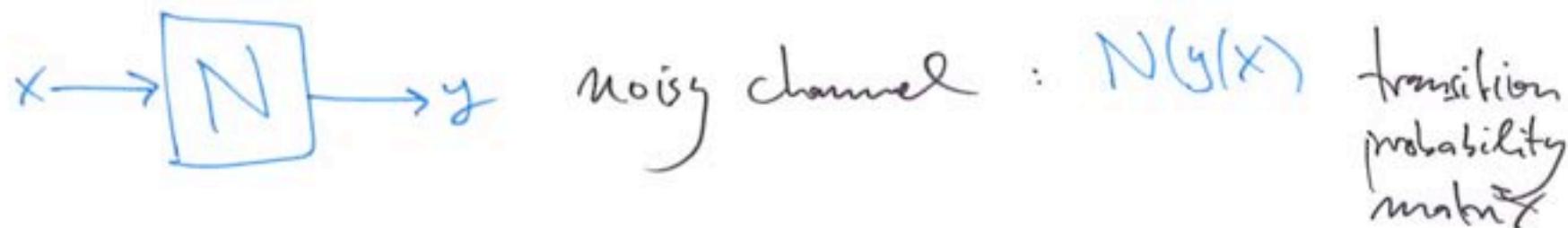
I-1

## I. Communication via classical noisy channels



(I-1)

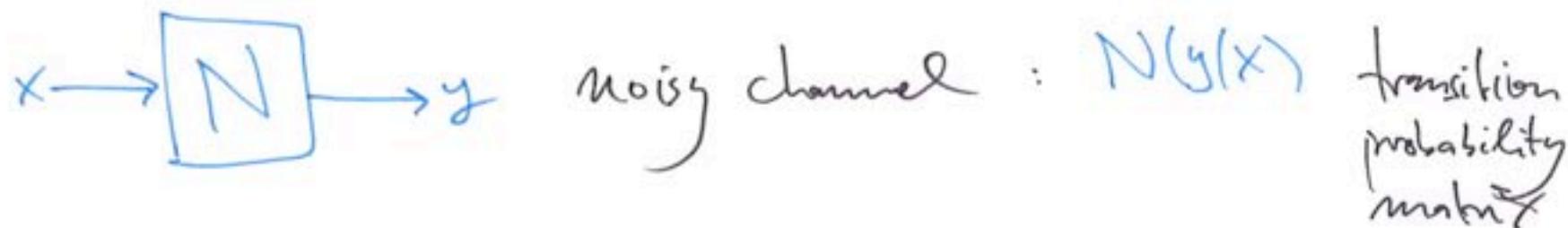
# I. Communication via classical noisy channels



Examples : \*Binary symmetric channel,  $BSC_p : \{0,1\} \rightarrow \{0,1\}$

$x \backslash y$	0	1
0	$1-p$	$p$
1	$p$	$1-p$

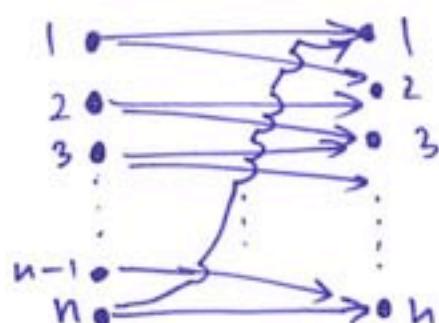
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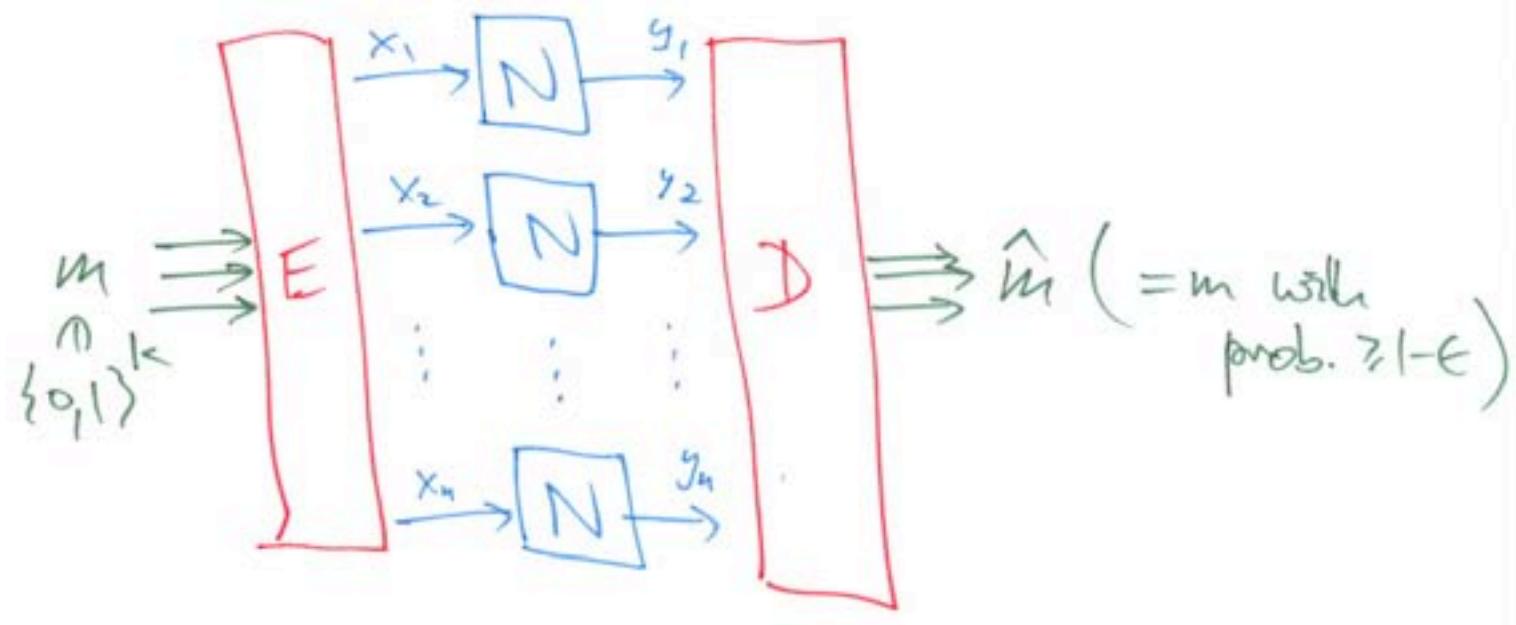
$x \setminus y$	0	1
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\* Typewriter channel  $T_n : \{1, \dots, n\} \rightarrow \{1, \dots, n\}$

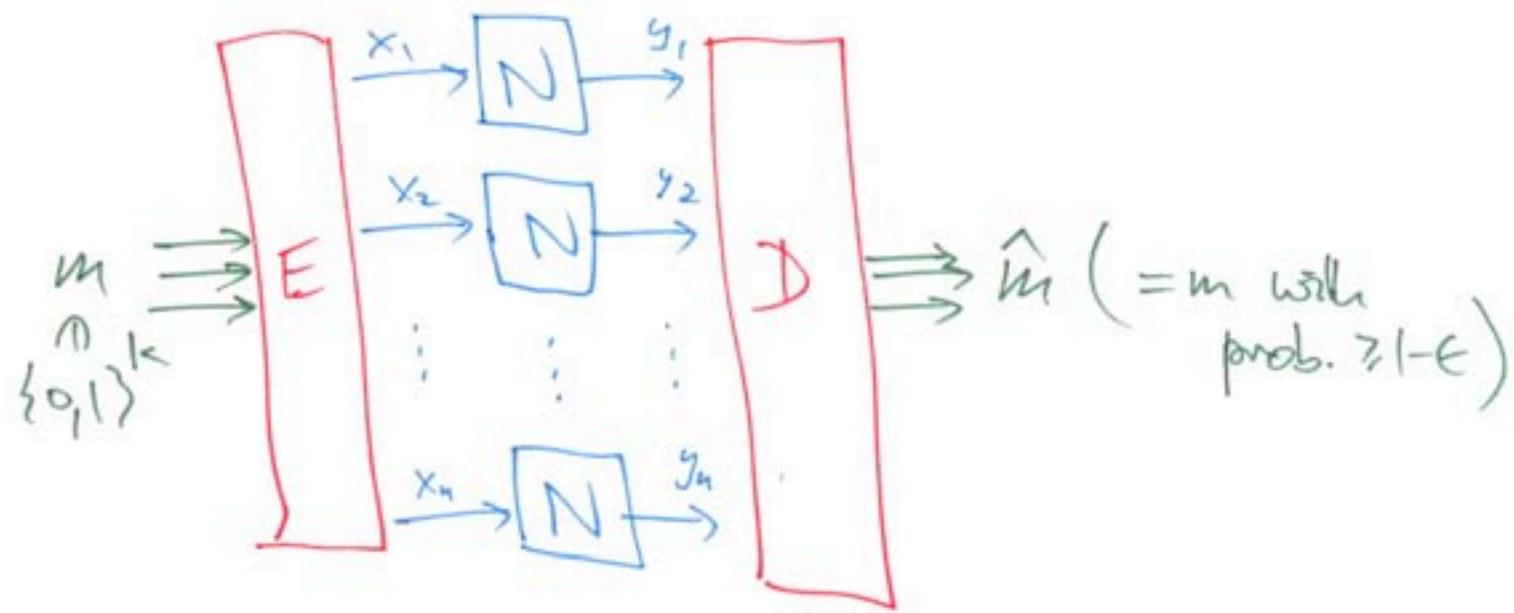


(each transition arrow with probability  $\frac{1}{2}$ )

Shannon 1948: capacity  $C(N) = \lim_{\epsilon \rightarrow 0} \lim_{n \rightarrow \infty} \frac{k(\epsilon)}{n}$

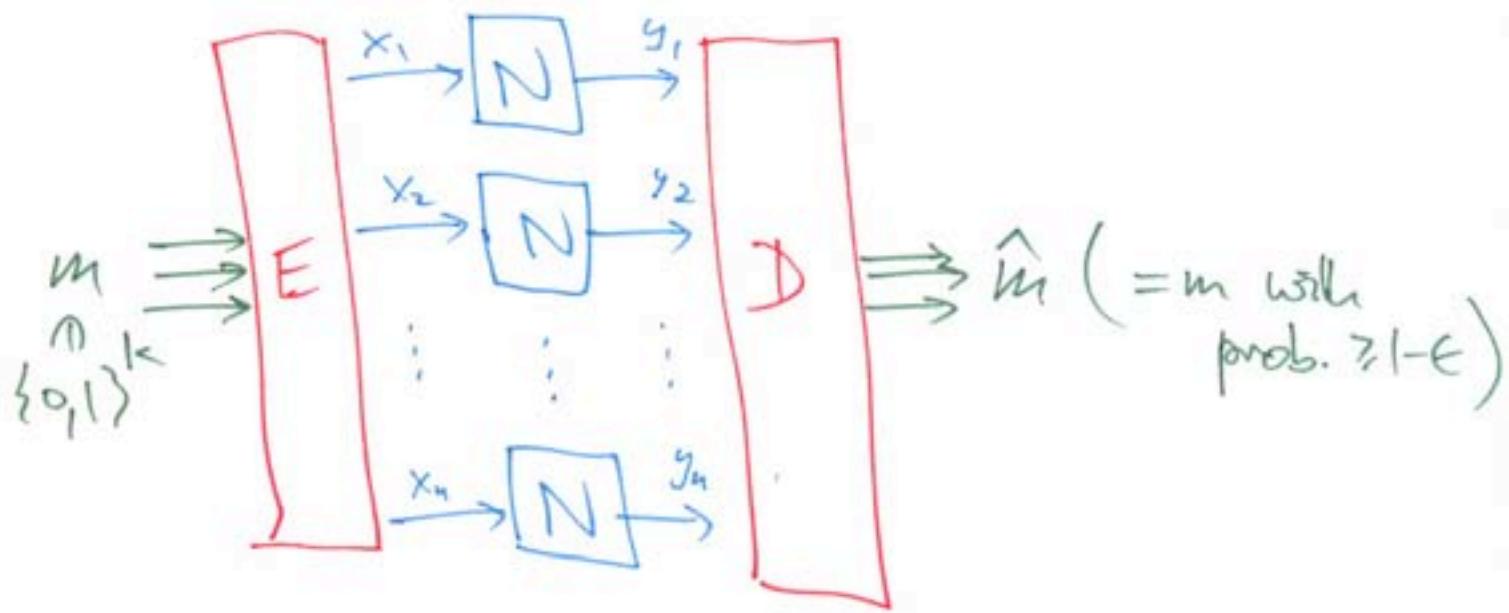


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Thm.  $C(N) = \max_{\substack{X \\ Y=N(X)}} H(X) + H(Y) - \underbrace{H(XY)}_{\text{"mutual information"} \\ I(X:Y)}$

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⚠  $C(N) > 0$  for all non-trivial  
(= non-constant) channels

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E-3

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$\underbrace{\phantom{\lim_{n \rightarrow \infty} \lim_{\epsilon \rightarrow 0}}}_{= k(0)/n}$

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$$\lim_{n \rightarrow \infty} \lim_{\epsilon \rightarrow 0} \frac{k(\epsilon)}{n}$$

$$= \overbrace{\frac{k(0)}{n}}$$

$$=: C_0(N)$$

zero-error  
capacity

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w.l.o.g.  $n=1$

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For  $k(0)$ : different messages  $m, m'$

w.l.o.g.  $n=1$   $\xrightarrow{\text{different encodings}} x = E(m), x' = E(m')$   
 ... must lead to perfectly distinguishable output distributions  $N(\cdot|x) \perp N(\cdot|x')$

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$\Rightarrow$  Bipartite graph of "possible transitions"  
 determines  $k(0)$ .

Adjacency matrix  $\Gamma(y|x) = \Gamma N(y|x)$

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Adjacency matrix  $\Gamma(y|x) = [N(y|x)]$

In fact: confusability graph  $G = G(N)$  with  
 vertices  $x, x'$  connected iff confusable

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of vertices  $x$  in  $G$  I-4

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Strong graph  
product  
 $G_1 \times G_2$

- Cartesian product of vertex sets

$$\bullet X_1 X_2 \sim X'_1 X'_2$$

$$\Leftrightarrow X_1 \sim X'_1 \text{ & } X_2 \sim X'_2$$

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Examples:

- If all  $N(y|x) \neq 0$  [e.g. BSC<sub>p</sub>], then  $G$  = complete graph
- For typewriter channel  $T_n$ ,  $G$  =  $n$ -gon.

E.g.  $n=5$   pentagon

Remarks : (i)  $C(N) = C(F) = \lim_{n \rightarrow \infty} \frac{1}{n} \log \alpha(F^n)$

(ii)

(iii)

(iv)

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(ii) Every  $G = (V, E)$  occurs as confusability graph :

"standard construction"  $N : V \rightarrow E$

$\stackrel{\psi}{\mapsto}$  random edge  $y \in E$   
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$\alpha(\square) = 2$  but  $\alpha(\square \times \square) = 5$ ,  
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$$\alpha(\square) = 2 \text{ but } \alpha(\square \times \square) = 5, \\ \text{implying } C_0(\square) \geq \frac{1}{2} \log 5$$

(iv) One of the biggest open problems in graph theory  
is to compute  $C_0(G)$ .

E.g. only 1979, Lovasz showed  $C_0(\square) = \frac{1}{2} \log 5$ ,  
but  $C_0$  (heptagon 

## Upper bounds on $\alpha(G)$ :

I-6

\* Fractional packing number  $\alpha^*$  : LP (Shannon '56)

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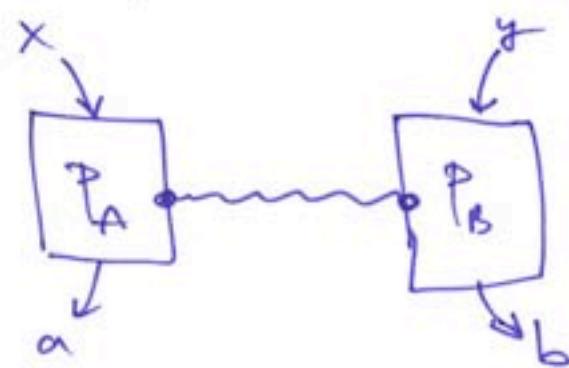
$$\begin{aligned} \alpha^*(\Gamma) := \max & \sum_x v(x) \\ \text{s.t. } & v(x) \geq 0, \\ & \sum_x v(x) \Gamma(y|x) \leq 1 \end{aligned}$$

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- Easy to see  $\alpha(G) \leq \vartheta(G) \leq \alpha^*(\Gamma)$
  - Both  $\vartheta$  and  $\alpha^*$  are multiplicative  
 $\Rightarrow \underline{\alpha_0(G) \leq \log \vartheta(G)}$
  - Best known upper bound, except for  
 some special graphs [Haemers '79; Alon '80]

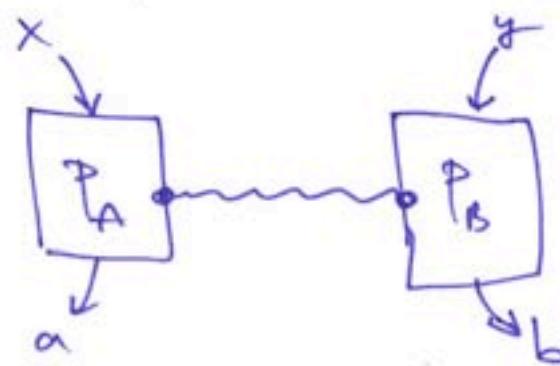
## II. Assistance by nonlocal correlations

In addition to N, we'll now grant Alice and Bob access to pre-shared correlations:



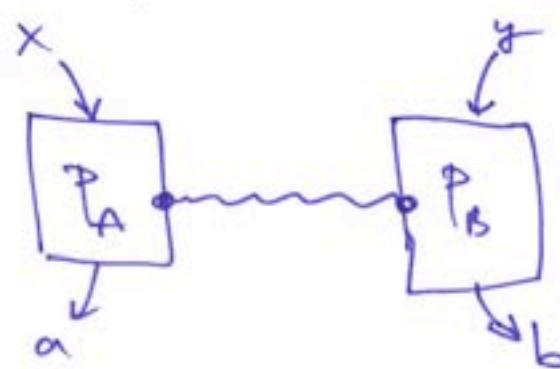
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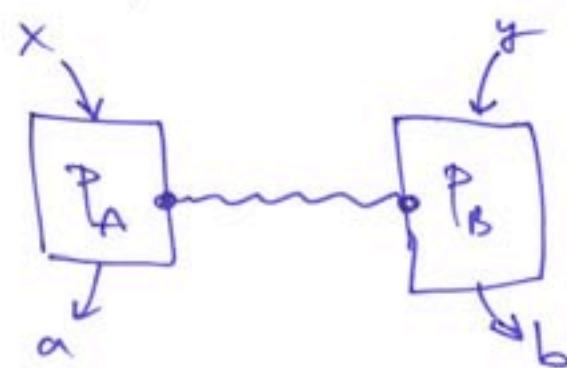
I.e.,

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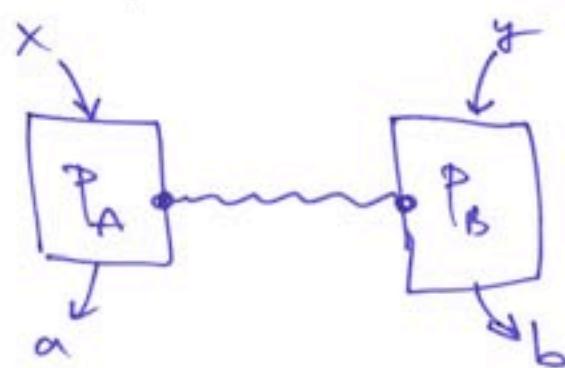
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Entanglement (E)

General non-signalling (NS)

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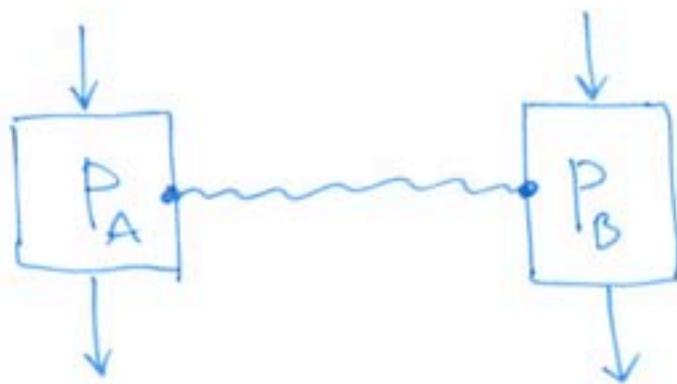
### CHSH game [a,b,x,y binary]

$$\Pr\{a \oplus b = x \cdot y\} \leq \frac{3}{4} \quad (\text{Bell})$$

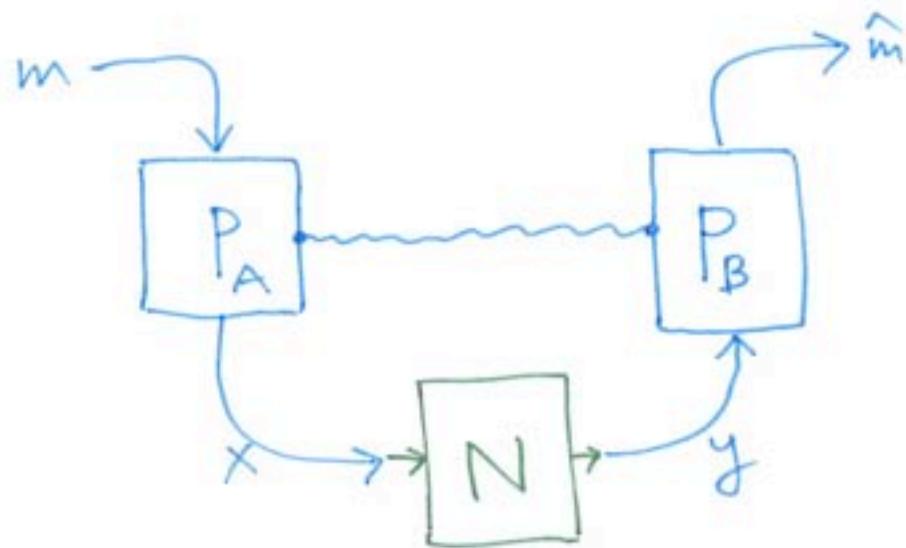
$$\begin{aligned} \text{--"--} &\leq \frac{1}{2}(1 + \sqrt{\frac{1}{2}}) \\ &\quad (\text{Tseitin}) \end{aligned}$$

$$\begin{aligned} \text{--"--} &\leq 1 \quad (\text{Popescu/Rohrlich}) \end{aligned}$$

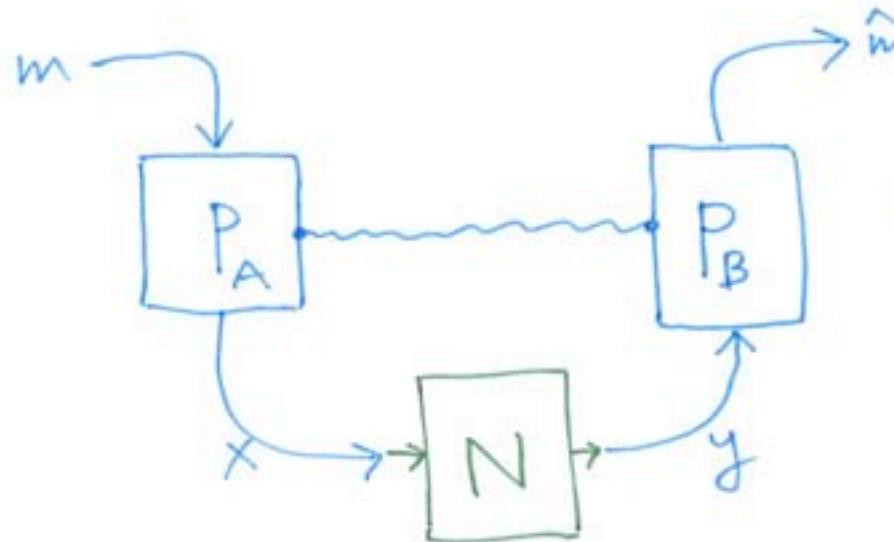
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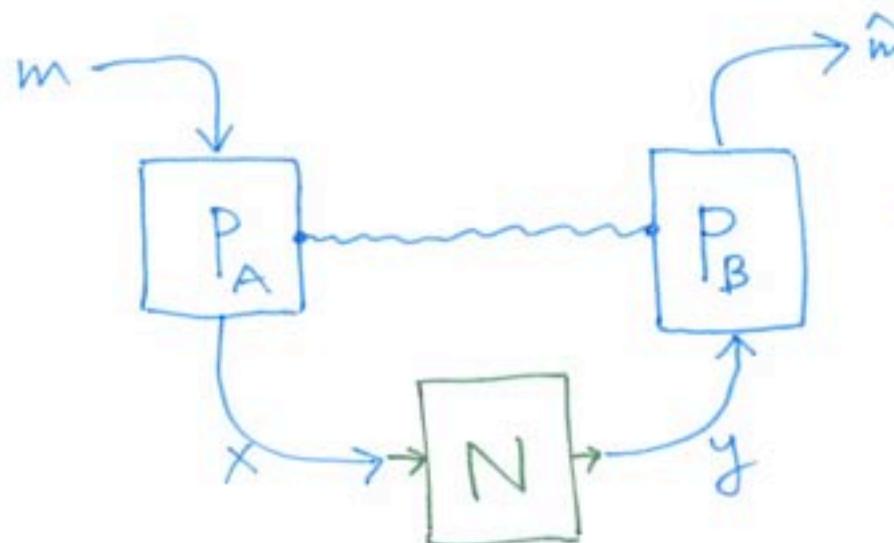


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### Constraints (on P):

- Probability distribution —  $P(x\hat{m}|my) \geq 0, \sum_{x\hat{m}} P(x\hat{m}|my) = 1$
- $P \in \{\text{restricted class}\}$ : SR, NS linear  
E more complicated
- $\Gamma(y|x) = 1 \text{ & } m \neq \hat{m} \Rightarrow P(x\hat{m}|my) = 0$

Looking back on Shannon-theoretic setting:

$$C_{NS}(N) = C_E(N) = C(N) = \max_X I(X:Y)$$

remains unchanged!

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- By the "Reverse Shannon Theorem" can simulate N using rate  $C(N)$  of noiseless communication + SR [Bennett / Shor / Smolin / Thapliyal 2001]

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Zero-error: looks differently (though SR clearly doesn't help)...

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$\tilde{\chi}(G) := \max \# \text{ of messages sent w/ 0-error}$   
via N and arbitrary entanglement.

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Proof.: Choose q Kochen-Specker bases  $B_m = \{| \varphi_1^{(m)} \rangle, \dots, | \varphi_d^{(m)} \rangle\}$  in  $\mathbb{C}^d$ .

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I.e., however you select one vector  $|\psi_{j,m}^{(m)}\rangle \in B_m$  from each basis, there will be two that are orthogonal:

$$|\psi_{j,m}^{(m)}\rangle \perp |\psi_{j,m'}^{(m')}\rangle$$

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Construct  $G$  with vertices  $(m, j)$  and edges  
 $(m, j) \sim (m', j') \text{ iff } |\psi_j^{(m)}\rangle \perp |\psi_{j'}^{(m')}\rangle$ .

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 $(m, j) \sim (m', j')$  iff  $|\psi_j^{(m)}\rangle \perp |\psi_{j'}^{(m')}\rangle$ .

$\alpha(G) < q$ , as " $=q$ " would contradict  
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I.e., however you select one vector  
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Shared E:

$\tilde{\alpha}(G) := \max \# \text{ of messages sent w/ 0-error}$   
via N and arbitrary entanglement.

⚠ Appears to depend on  $\Gamma$ , but really only on  $G$

Thm. There exist graphs with  $\tilde{\alpha}(G) > \alpha(G)$ .

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$\tilde{\alpha}(G) \geq q$ : Use maximally ent. state  $|E\rangle_d$ ; for message m, Alice measures  
basis  $B_m$  and puts  $(m, j)$  into the channel. Bob's state  $|\psi_j^{(m)}\rangle$  is  
one of an orthogonal set indicated by the output y: measures ...



II-II

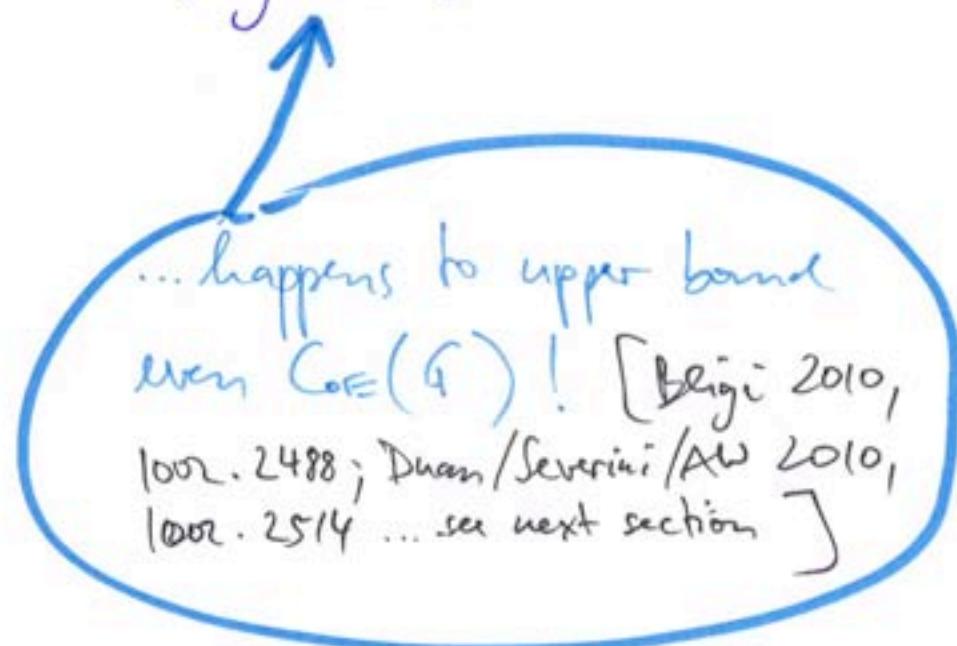
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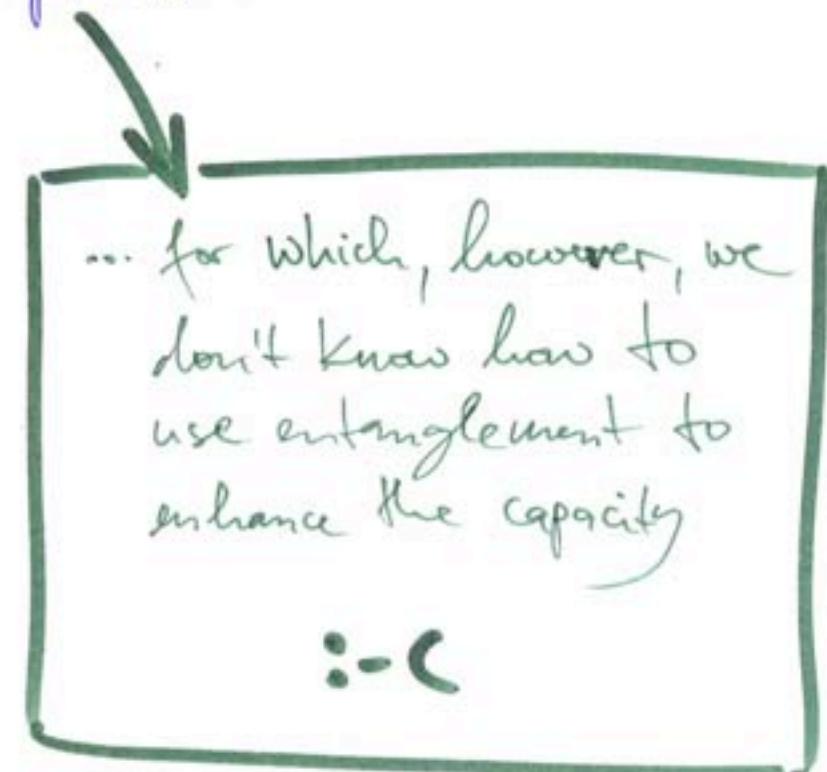
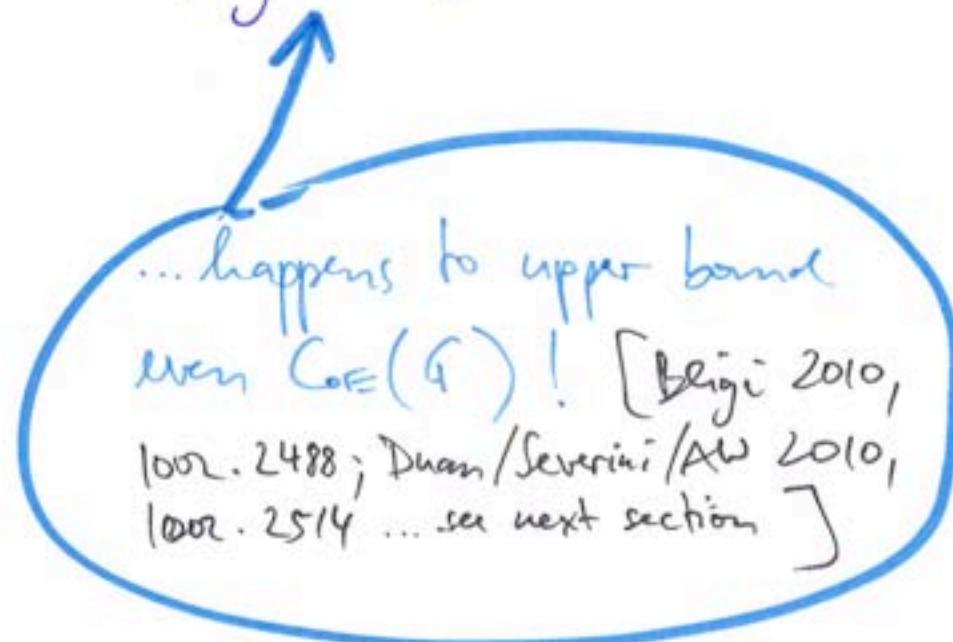
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Remark: Shannon (1956) proved that  $N$ , assisted by feedback, has zero-error capacity  $C_0(N) = \begin{cases} 0 & \text{if } C_0(N) = 0, \\ \log \alpha^*(\Gamma) & \text{if } C_0(N) > 0. \end{cases}$   
 Deeper reason for equality?

Example:  $N : \begin{matrix} [4] \\ \downarrow \\ \{1, 2, 3, 4\} \end{matrix} \longrightarrow \begin{pmatrix} [4] \\ 2 \\ \downarrow \\ \{12, 13, 14, 23, 24, 34\} \end{pmatrix}$

$x \longmapsto$  random  $y = xx^T$  containing  $x$

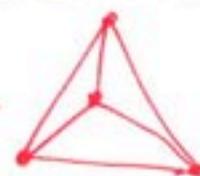
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$$\Gamma = \begin{array}{|c|c|c|c|c|c|c|} \hline & x & y & 12 & 13 & 14 & 23 & 24 & 34 \\ \hline x & | & | & 1 & 1 & 1 & & & \\ \hline y & | & | & & & & 1 & 1 & \\ \hline 1 & | & | & 1 & 1 & 1 & & & \\ \hline 2 & | & | & & & 1 & 1 & & \\ \hline 3 & | & | & & 1 & 1 & 1 & & \\ \hline 4 & | & | & & & 1 & 1 & 1 & \\ \hline \end{array}$$

$$G = K_4 =$$



complete graph



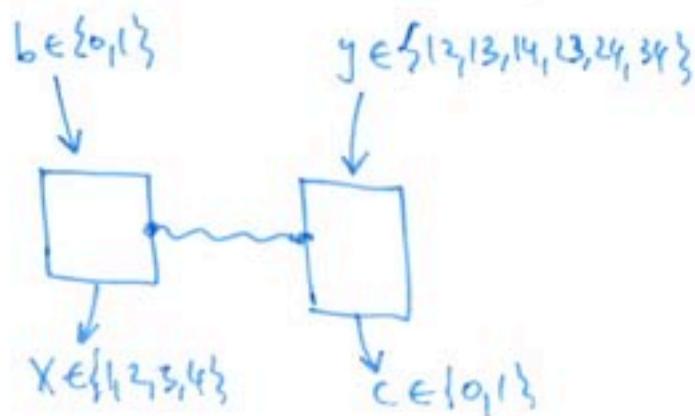
$$\alpha(G) = 1, C_0(G) = 0$$

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However,  $\alpha^*(\Gamma) = 2$ , and can indeed transmit one bit.



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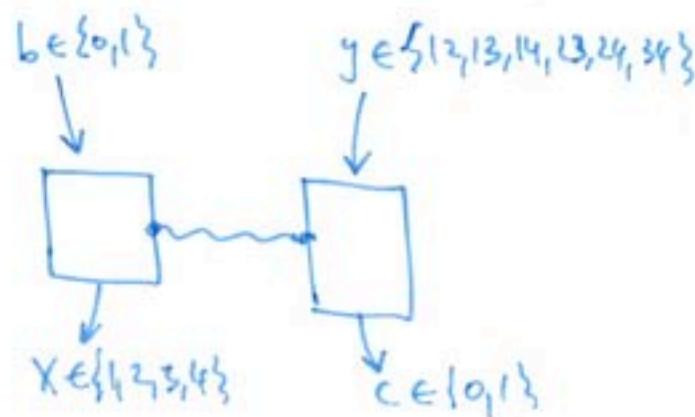
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Remark:  $\tilde{\alpha}$  depends only on  $\Gamma$ , but for  $\alpha_{NS}$  need  $\Gamma$ . That's because quantum states are distinguishable iff pairwise distinguishable; this is not true for general probabilistic theories...

Reverse problem: simulate  $N$  perfectly with limited noiseless communication & pre-shared correlations (at least SR).

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This becomes an equality when the l.h.s. is minimised over all  $N$  with given  $\Gamma$ .

I.e. zero-error reversibility for all  $\Gamma$  when assisted by NS correlations!

### III. Quantum channels

C.p.t.p. map  $\mathcal{N} : \mathcal{I}(A) \rightarrow \mathcal{I}(B)$ , written either  
 in Kraus form  $\mathcal{N}(\rho) = \sum_j E_j \rho E_j^+$  ( $\sum_j E_j^+ E_j = \mathbb{1}$ )

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Want to communicate error-free classical  
 or quantum information ; assisted by entanglement  
 or not ...

To encode classical information : need input states

$$\varphi_m = |\varphi_{m_1} \times \varphi_{m_2}|, \varphi_{m'} = |\varphi_{m'_1} \times \varphi_{m'_2}| \quad (\text{w.l.o.g. pwe}) \text{ s.t.}$$

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Hence regard  $S \subset \mathcal{L}(A)$   
as "non-commutative" (composability)  
graph ...

Examples :

- $S = C\mathbb{1}$  iff  $M$  is an isometry
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... Also composition of channels, direct sums, etc,

translate nicely into graph language. E.g. for channels  
 $M_1, M_2$  with graphs  $S_1, S_2$ , tensor product  $M_1 \otimes M_2$  has graph  $S_1 \otimes S_2$ .

Using non-commutative graph  $S \subset \mathcal{L}(A)$ :

- Independence number  $\alpha(S) := \max M$  s.t.  $\exists |\varphi_1\rangle, \dots, |\varphi_M\rangle \in A$   
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- Quantum independence number  $\alpha_q(S) := \max \text{rank } P$  s.t.  $P = P^\dagger P \in \mathcal{L}(A)$   
 projector with  
 $PSP = \mathbb{C}P$

(I.e. the Knill-Lafitte condition for the support of  $P$  being)  
 a quantum error correcting code for  $n$ .

- Entanglement-assisted independence number

$$\tilde{\chi}(S) := \max M \text{ s.t. } \exists F_1, \dots, F_N \in \mathcal{L}(A) \otimes \mathcal{L}(R)$$

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Then, obvious definitions of capacities:

$$Q_0(S) = \lim_{n \rightarrow \infty} \frac{1}{n} \log \alpha_0(S)$$

$$C_0(S) = \lim_{n \rightarrow \infty} \frac{1}{n} \log \alpha(S)$$

$$C_{0E}(S) = \lim_{n \rightarrow \infty} \frac{1}{n} \log \tilde{\chi}(S)$$

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- $\alpha_q(S) \leq \alpha(S) \leq \tilde{\alpha}(S) \leq 1 + \dim S^\perp$

III-2

$$\frac{d}{dt} \left( \int_{\Omega} u^2 \partial_x u \partial_t u \right) = 2 \int_{\Omega} \partial_x u \partial_t u \partial_x u + \int_{\Omega} \partial_x u \partial_x u \partial_t u = - \int_{\Omega} \partial_x u \partial_x u \partial_t u.$$

A quantum extension of Lovász'  $\vartheta$ :

$$\text{Proto-definition: } \vartheta(S) := \max_T \| \mathbb{1} + T \| \quad \text{s.t. } T \perp S \text{ and } \mathbb{1} + T \geq 0$$

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- Can be recast as SDP
- It's multiplicative
- $\tilde{\vartheta}(S) \leq \tilde{\vartheta}(S)$ , hence  $\text{GOE}(S) \leq \log \tilde{\vartheta}(S)$

Proof that  $\tilde{\alpha}(S) \leq \tilde{\alpha}(T)$ :

E-assisted independent set : given by state  $\rho$  on  $A \otimes R$   
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 $=: R'$

and check  $1+T \geq \sum_{m, m'} U_m X U_{m'}^\dagger \otimes |m\rangle \langle m'|$

$$= \left( \sum_m U_m \sqrt{X} \otimes |m\rangle \right) \left( \sum_{m'} U_{m'} \sqrt{X} \otimes |m'\rangle \right)^+ \geq 0$$

Proof that  $\tilde{\alpha}(S) \leq \tilde{\alpha}(F)$ :

E-assisted independent set: given by state  $\rho$  on  $A \otimes R$  and unitaries  $U_m$  ( $m=1, \dots, M$ )  
 s.t.  $U_m \rho U_{m'}^\dagger \in S^\perp \otimes \mathcal{L}(R)$   $\forall m \neq m'$

$\rho \propto X \geq 0$ ,  $\|X\|=1$  with 1-eigenvector  $|p\rangle$

Define  $T := \sum_{m \neq m'} U_m X U_{m'}^\dagger \otimes |m\rangle \langle m'| \in S^\perp \otimes \mathcal{L}(R \otimes \mathbb{C}^M)$   
 $=: R'$

and check  $1+T \geq \sum_{m, m'} U_m X U_{m'}^\dagger \otimes |m\rangle \langle m'|$

$$= \left( \sum_m U_m \sqrt{X} \otimes |m\rangle \right) \left( \sum_{m'} U_{m'} \sqrt{X} \otimes |m'\rangle \right)^+ \geq 0$$

Finally,  $\|1+T\| \geq \left\| \sum_m U_m \sqrt{X} \otimes |m\rangle \right\|^2$   
 $\geq \left\| \sum_m U_m |p\rangle \otimes |m\rangle \right\|^2 = M$  .  $\square$

#### IV. Conclusion and open questions

- (1)  $\tilde{\alpha}(S) \leq \tilde{I}(S)$  for classical channels  
gives  $\tilde{\alpha}(G) \leq \tilde{I}(G)$  — improving Lovasz' bound  
because we know that  $\alpha(G) < \tilde{\alpha}(G)$ , sometimes.

## IV. Conclusion and open questions

(1)  $\tilde{\alpha}(S) \leq \tilde{\beta}(S)$  for classical channels

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(2) Is  $C_{\text{OE}}(S) = \log \tilde{\beta}(S)$ ?

## IV. Conclusion and open questions

(1)  $\tilde{\chi}(S) \leq \tilde{\vartheta}(S)$  for classical channels

gives  $\tilde{\chi}(G) \leq \tilde{\vartheta}(G)$  — improving Lovasz' bound  
because we know that  $\alpha(G) < \tilde{\chi}(G)$ , sometimes.

(2) Is  $C_{\text{OE}}(S) = \log \tilde{\vartheta}(S)$ ?

Interesting test cases:

- Classical channels, esp.  $G$  for which  $C_{\text{OE}}(G) < \log \tilde{\vartheta}(G)$  is known
- $S = \begin{bmatrix} \lambda_0 & & \\ & \ddots & \\ & & -1 \end{bmatrix}^\perp$  with  $1 \leq \lambda_0 \leq d-1$

Note that  $\tilde{\chi}(S) = 2$ ,  $\tilde{\vartheta}(S) = \lambda_0 + 1$ ; we  
don't even know whether  $\tilde{\chi}(S \otimes S) > 4$ !

(3) Better bounds on  $\alpha(s)$ ?

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[Would be useful to separate  $\alpha$  and  $\tilde{\alpha}$ , but also  
— if multiplicative — to separate  $C_0$  and  $C_0 E$ ,  
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(4) Feedback-assisted capacity  $C_{0F}(N) = ?$

(3) Better bounds on  $\alpha(\mathcal{S})$ ?

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 — if multiplicative — to separate  $C_0$  and  $C_{0E}$ ,  
 which we'd really like to do for classical  $\mathcal{S}$ ]

(4) Feedback-assisted capacity  $C_{0F}(\mathcal{N}) = ?$

[We know it's only a function of  $\Gamma = \text{span } \{E_j : j\}$   
 $\subset L(A \rightarrow B)$   
 (generalises the bipartite graph of classical case ...)]

- Extension of Shannon's 1956 theory?
- Multiplicative extension of fractional packing number?