No-signalling assisted zeroerror communication via quantum channels and the Lovász & number

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arXiv:1409.3426



If you've been partying...



### Hungover Summary

1. 
$$C_0(G) \leq \log \vartheta(G)$$

2. 
$$C_{OE}(G) \leq \log \vartheta(G)$$

3.-5. 
$$C_{ONS}(G) = log \vartheta(G)$$

#### Hungover Summary

Zero-error capacity

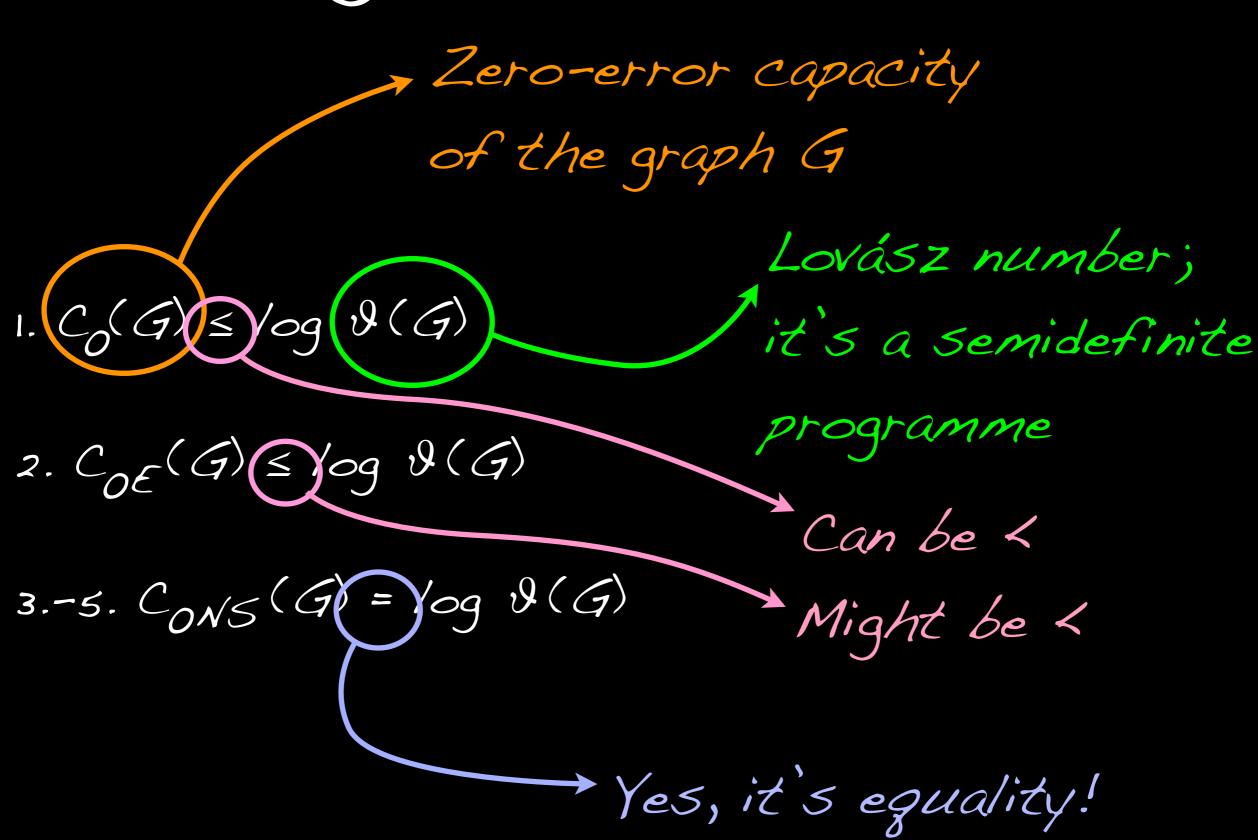
of the graph GLovász number;  $C_0(G) \leq log(\vartheta(G))$ it's a semidefinite

programme

2. 
$$C_{0\varepsilon}(G) \leq \log \vartheta(G)$$

3.-5.  $C_{ONS}(G) = log \vartheta(G)$ 

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Channel  $N: X \rightarrow Y$ , i.e. stochastic map

$$X \ni \chi \longrightarrow \bigvee \in Y$$

N(ylx): transition probabilities

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N(ylx): transition probabilities

Want to send information (in x), such that receiver (seeing y) can be certain about it.

i) Transition graph  $\Gamma$ : bipartite graph on XXY with adjacency matrix

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2) Confusability graph G on X: adj. matrix  $(1+A)_{xx} = \begin{cases} 1 & \text{if } \mathcal{N}(.1x)^T \mathcal{N}(.1x') > 0, \\ 0 & \text{if } \mathcal{N}(.1x)^T \mathcal{N}(.1x') = 0. \end{cases}$ 

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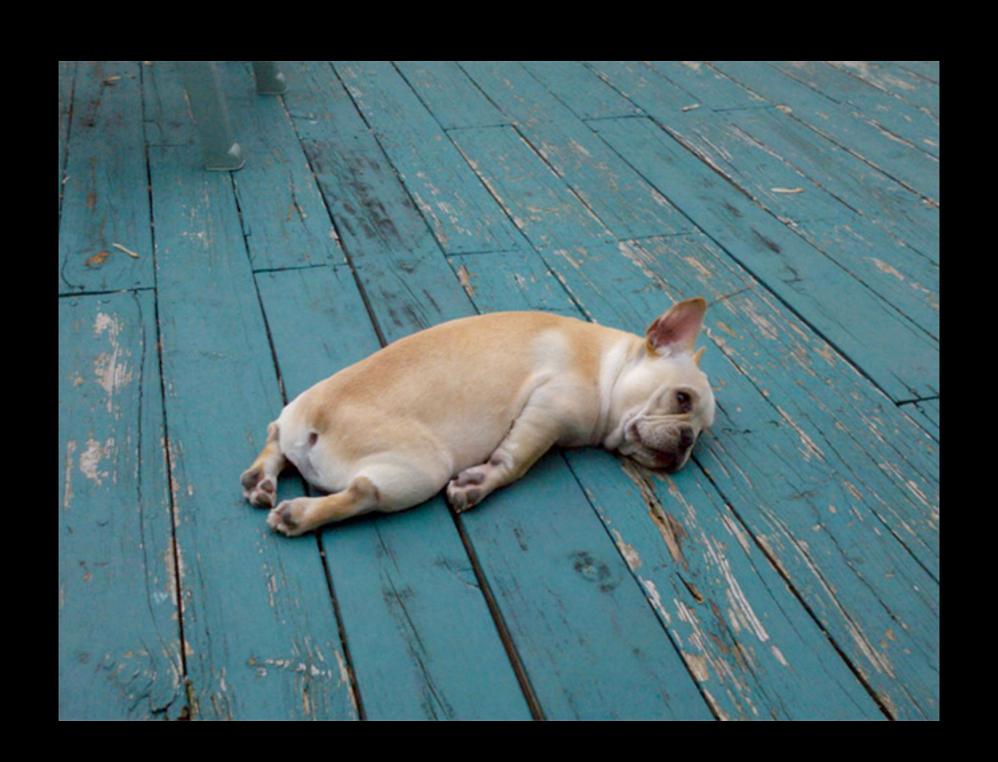
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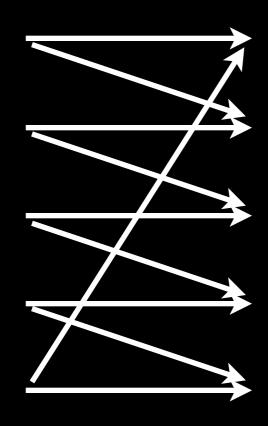
Lovász convention:

x-x'iff x=x' or xx' edge

### Example?



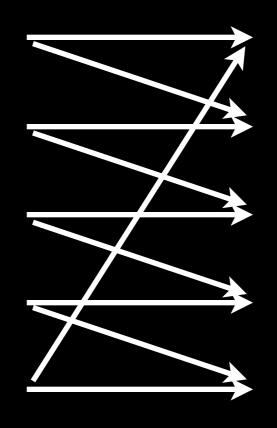
$$\Gamma$$
=75

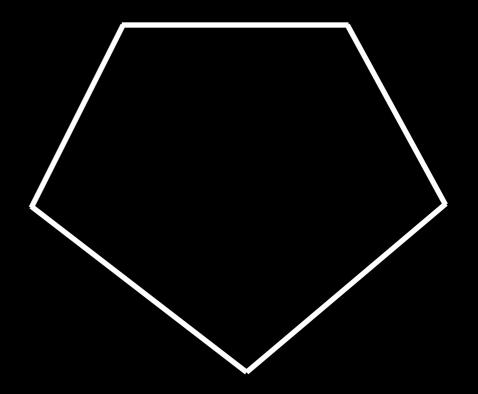


typewriter

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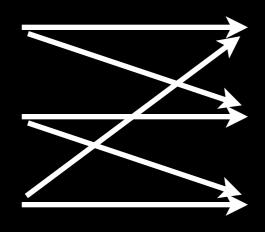


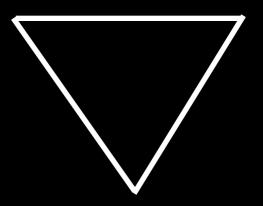


typewriter

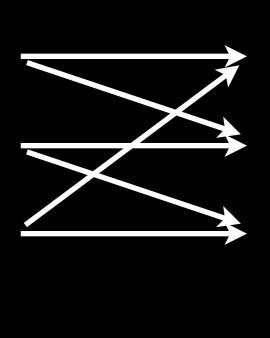
pentagon

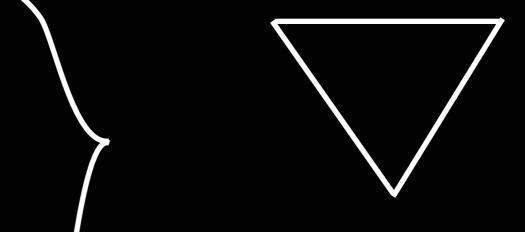
$$\Gamma = 7_3$$





$$\Gamma = \mathcal{T}_3$$





#### Product channels:

$$N \times N'(\gamma \gamma' | \chi \chi') = N(\gamma | \chi) N'(\gamma' | \chi')$$

$$X \ni \chi \longrightarrow \bigvee \longrightarrow \bigvee \in Y$$

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$$X \ni X \longrightarrow \mathcal{N} \longrightarrow \mathcal{Y} \in Y$$

$$X \ni X \longrightarrow \mathcal{N} \longrightarrow \mathcal{Y} \in Y'$$

Graphs via Kronecker/tensor product:

$$\Gamma(N \times N') = \Gamma \otimes \Gamma'$$

$$1 + A(N \times N') = (1 + A) \otimes (1 + A')$$

Product channels:

$$N \times N'(yy'|xx') = N(y|x)N'(y'|x')$$

$$X \ni X \longrightarrow \mathcal{N} \longrightarrow \mathcal{Y} \in Y$$

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Graphs via Kronecker/tensor product:

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$$Strong graph product G \times G'$$

$$; \longrightarrow x = f(i) \longrightarrow N \longrightarrow y \text{ possible:}$$

$$N(y|x) > 0$$

$$; \longrightarrow \chi = f(i) \longrightarrow \mathbb{N} \longrightarrow \text{y possible:}$$

$$\mathcal{N}(y|\chi) > 0$$

Hence: codebook  $\{f(i)\} \subset X$  must be an independent set in G.

Maximum Size:

 $\alpha(G) := independence number of G.$ 

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Upper bounds!?

[L. Lovász, IEEE-IT 25(1):1-7, 1979]

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$$\leq \alpha^*(\Gamma) = \max \sum_{x} \omega_x \text{ s.t. } \omega_x \geq 0 \text{ \&}$$

$$\forall y \sum_{x} \Gamma(y|x) \omega_x \leq 1.$$

[C.E. Shannon, IRE-IT 2(3):8-19, 1956]

[L. Lovász, IEEE-IT 25(1):1-7, 1979]

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Best:  $\alpha^*(G) = \min \alpha^*(\Gamma)$  s.t.  $G \supset graph of \Gamma$ 

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Best:  $\alpha^*(G) = \min_{\alpha} \alpha^*(\Gamma)$  s.t.  $G \supset graph \ of \ \Gamma$ 

(Attained at  $\Gamma$  that has an output for every maximal clique of  $G: \Gamma(C|x)=1$  iff  $x \in C$ .)

Asymptotically many channel uses - capacity:

$$i \longrightarrow \mathcal{A}(i) \left\{ \begin{array}{c} x_1 \longrightarrow \mathcal{N} \\ \\ x_2 \longrightarrow \mathcal{N} \\ \\ \vdots \\ \\ x_n \longrightarrow \mathcal{N} \end{array} \right\} \longrightarrow j$$

$$C_o(G) = \lim_{n \to \infty} \frac{1}{n} \log \alpha(G^{\times n})$$

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$$C_{o}(G) = \lim_{n} \frac{1}{n} \log \alpha(G^{\times n})$$

$$= \sup_{\alpha \in G \times H} because$$

$$\alpha(G \times H) \ge \alpha(G)\alpha(H)$$

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$$C_{o}(G) = \lim_{n} \log \alpha(G^{\times n}) \leq \log \vartheta(G)$$

$$= \sup_{x \in \mathcal{A}} \frac{\log \vartheta(G)}{\log \vartheta(G)} \leq \log \vartheta(G)$$

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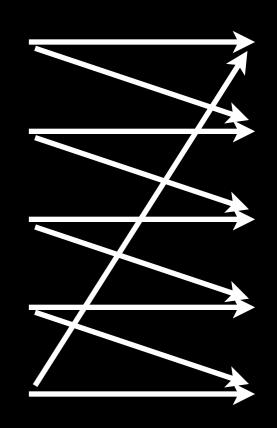
Also fractional packing I number multiplicative:

$$\alpha^*(\Gamma \otimes \Gamma') = \alpha^*(\Gamma) \alpha^*(\Gamma'),$$

$$\alpha^*(G_XY) = \alpha^*(G)\alpha^*(Y)!$$

All inequalities can be strict; first and last:

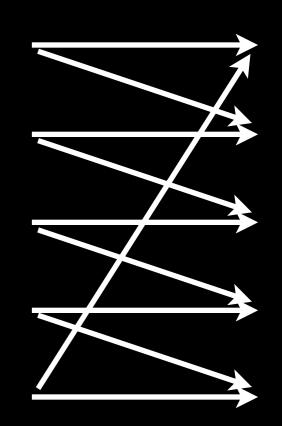
Ex. Typewriter channel/pentagon  $\alpha(C_5)=2$ ,  $\alpha(C_5\times C_5)=5>4$ , but  $\vartheta(C_5)=\sqrt{5}$ , and  $\alpha^*(T_5)=5/2$ .



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Note:  $\alpha^*(7_3) = 3/2$ , but  $\alpha^*(*) = 1!$ 



All inequalities can be strict; first and last:

Random graphs G ~ G(n, 1/2) have, whp,

 $\alpha(G) \approx \log n, \ \vartheta(G) \approx \sqrt{n}, \ \alpha^*(G) \approx n/(\log n)$ 

All inequalities can be strict; middle due to W. Haemers [IEEE-IT 25(2);231-232, 1979], via a different algebraic and multiplicative bound on  $\alpha$  which sometimes(!) is better than  $\vartheta$ .

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However: w/o sacrificing multiplicativity,

d cannot be improved [Acin/Duan/Sainz/AW, 2014].

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Determination of  $C_o(G)$  open, not even known to be computable...

[N. Alon/E. Lubetzky, IEEE-IT 52(5):2172-2176, 2006]

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Idea: Perhaps we can close the gap by allowing additional resources in the en-/decoding?

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+ feedback [C.E. Shannon, IRE-IT 2(3):8-19, 1956]  $C_{of}(\Gamma) = \log \alpha^*(\Gamma), \text{ with constant}$ 

activating noiseless bits.

### $C_o(G) \leq log \vartheta(G)$

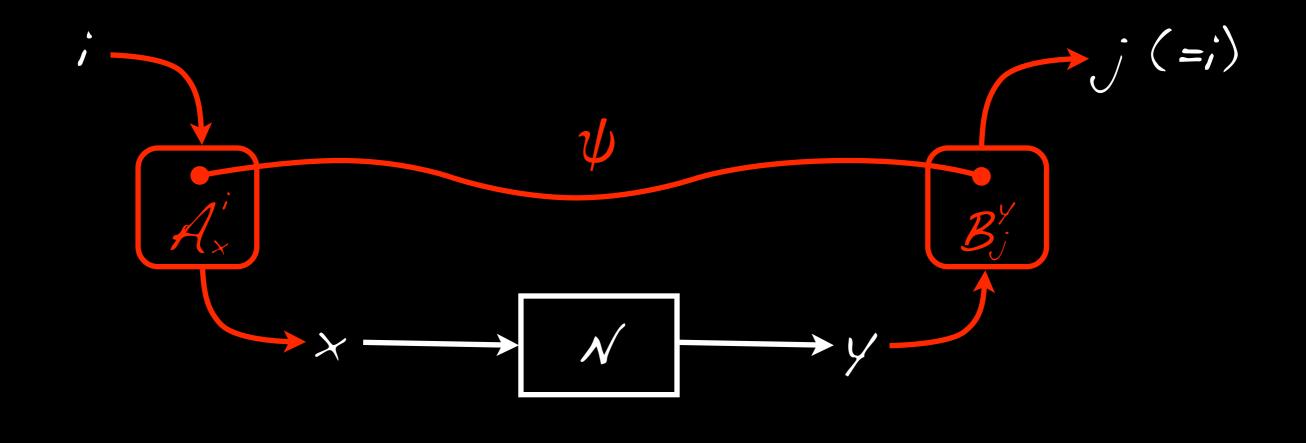
Idea: Perhaps we can close the gap by allowing additional resources in the en-/ decoding?

+ feedback [C.E. Shannon, IRE-IT 2(3):8-19, 1956] + entanglement (quantum correlations) + no-signalling correlations

### 2. Free non-local resources



### For instance, with free entanglement:



Maximum #messages =:  $\widetilde{\alpha}(G)$ 

Can show that this depends only on G; furthermore can be  $> \alpha(G)$ ...

[T.S. Cubitt et al., PRL 104:230503, 2010]

[S. Beigi, PRA 82:010303, 2010; R. Duan/S. Severini/AW, IEEE-IT 59(2):1164-1174, 2013.]

Since  $\vartheta$  is multiplicative under strong graph product,  $\vartheta(GxH)=\vartheta(G)\vartheta(H)$ , get:

 $C_o(G) \leq C_{o\varepsilon}(G) = \lim_{n \to \infty} \frac{1}{n} \log \widetilde{\alpha}(G^{\times n}) \leq \log \vartheta(G)$ 

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Known examples of separation

[D. Leung/L. Mancinska/W. Matthews/+2, CMP 311:97-111, 2012;

J. Briët/H. Buhrman/D. Gijswijt, PNAS 110:19227, 2012]

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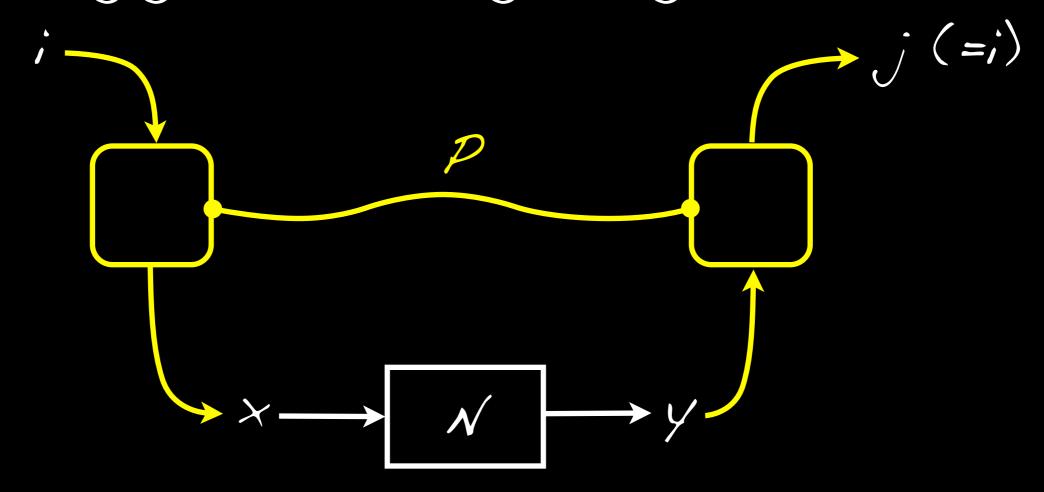
 $C_{o}(G) \leq C_{o\varepsilon}(G) = \lim_{n \to \infty} \frac{1}{n} \log \widetilde{\alpha}(G^{*n}) \leq \log \vartheta(G)$   $Unknown \ whether = or < !$ 

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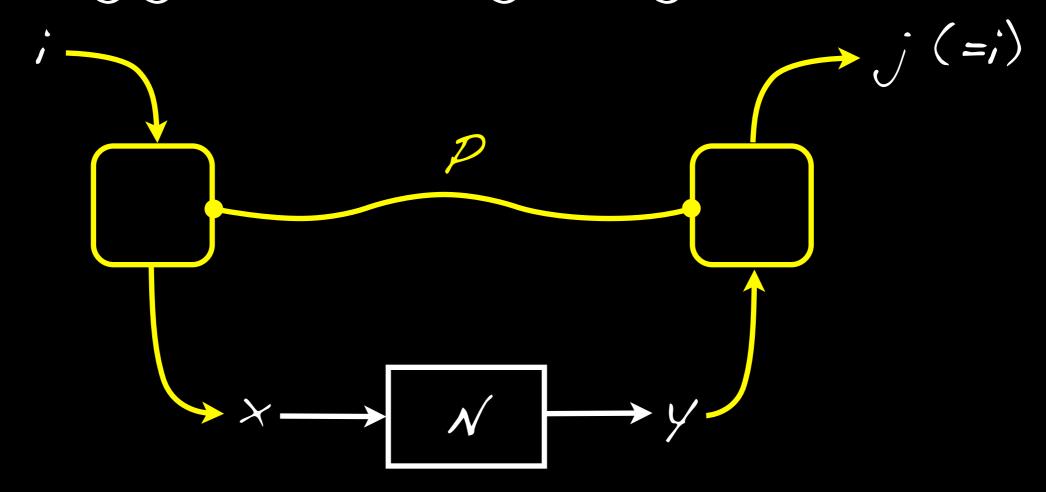
Allowing general no-signalling correlation:



This is no-signalling assisted zero-error code if j=i with probability 1.

I.e., for all j = i & edges xy in \(\Gamma\), \(\Reg(\chi)\) = 0

Allowing general no-signalling correlation:



This is no-signalling assisted zero-error code if j=i with probability 1.

I.e., for all  $j \neq i$  & edges xy in  $\Gamma$ , P(xjliy)=0Maximum #msg. with  $P \in NS =: \overline{\alpha}(\Gamma)$   $\overline{\alpha}(\Gamma) = \max m \text{ s.t. } \mathcal{A}(x) | iy \in \mathbb{N}, ij = 1...m,$   $\forall i \neq j \forall xy \in \Gamma \quad \mathcal{A}(x) | iy = 0.$ 

 $\overline{\alpha}(\Gamma) = \max m \text{ s.t. } \mathcal{A}(x) = NS, ij=1...m,$   $\forall i \neq j \forall xy \in \Gamma \quad \mathcal{A}(x) = 0.$ 

Clear: Can test given m efficiently by linear programming. Less obvious:

 $\overline{\alpha}(\Gamma) = \max m \text{ s.t. } \mathcal{A}(\gamma) \in \mathcal{NS}, \ ij=1...m,$   $\forall i \neq j \forall x y \in \Gamma \ \mathcal{A}(\gamma) = 0.$ 

Clear: Can test given m efficiently by linear programming. Less obvious:

Thm.  $\overline{\alpha}(\Gamma) = \lfloor \alpha^*(\Gamma) \rfloor$ , with  $\alpha^*$  the fractional packing number of  $\Gamma$ :  $\alpha^*(\Gamma) = \max \sum_{x} \omega_x \text{ s.t. } \omega_x \geq 0 \text{ & for all y,}$   $\sum_{x} \Gamma(y|x)\omega_x \leq 1.$ 

[T.S. Cubitt et al., IEEE-IT 57(8):5509-5523, 2011]

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Cor. Due to multiplicativity of  $\alpha^*$ ,

 $C_{oNS}(\Gamma) = \lim_{n \to \infty} \frac{1}{n} \log \overline{\alpha}(\Gamma^{\otimes n}) = \log \alpha^*(\Gamma).$ 

[C.E. Shannon, 1956: Same answer for feedback-assisted capacity!]

... so this is too big - what now?!

## 3. Quantum Version...



# 3. Quantum Version...

Should really consider quantum channels  $N:B(A) \longrightarrow B(B)$ , cptp map on states:

$$\rho \longrightarrow \qquad M \longrightarrow \sigma = M(\rho)$$

Kraus form:  $N(\rho) = \sum_{i} \mathcal{E}_{i} \rho \mathcal{E}_{i}^{t}, \sum_{i} \mathcal{E}_{i}^{t} \mathcal{E}_{i} = 1$ 

For quantum channel (cptp map)  $N: B(A) \longrightarrow B(B), \text{ with Kraus op's } E_i:$ 

Define  $K = Span \{E, \} \subset B(A \rightarrow B)$  and  $S = K^{\dagger}K = Span \{E, E, \} \subset B(A)$  as natural analogues of the transition and confusability graphs.

[R. Duan/S. Severini/AW, IEEE-IT 59(2):1164-1174, 2013; R. Duan/AW, arXiv:1409.3426] For quantum channel (cptp map)  $N: B(A) \longrightarrow B(B), \text{ with Kraus op's } E_i:$ 

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 $(S = S^{\dagger} \ni 1)$ , so S is an operator system)

[R. Duan/S.Severini/AW, IEEE-IT 59(2):1164-1174, 2013; R. Duan/AW, arXiv:1409.3426] Define  $K = Span \{ E, \} \subset B(A \rightarrow B)$  and  $S = K^{\dagger}K = Span \{ E, \} \subset B(A)$ .

For classical channel, Kraus operators are  $\propto \Gamma(y|x)$  ly><x1, so:

 $K = Span \{ \Gamma(y|x) | y > < x | \} \longleftrightarrow \Gamma,$ 

 $S = span [1x'] < x 1 s.t. x - x'] \leftrightarrow G.$ 

...hence S, K extend G,  $\Gamma$  to quantum...

[R. Duan/S.Severini/AW, IEEE-IT 59(2):1164-1174, 2013;
R. Duan/AW, arXiv:1409.3426]

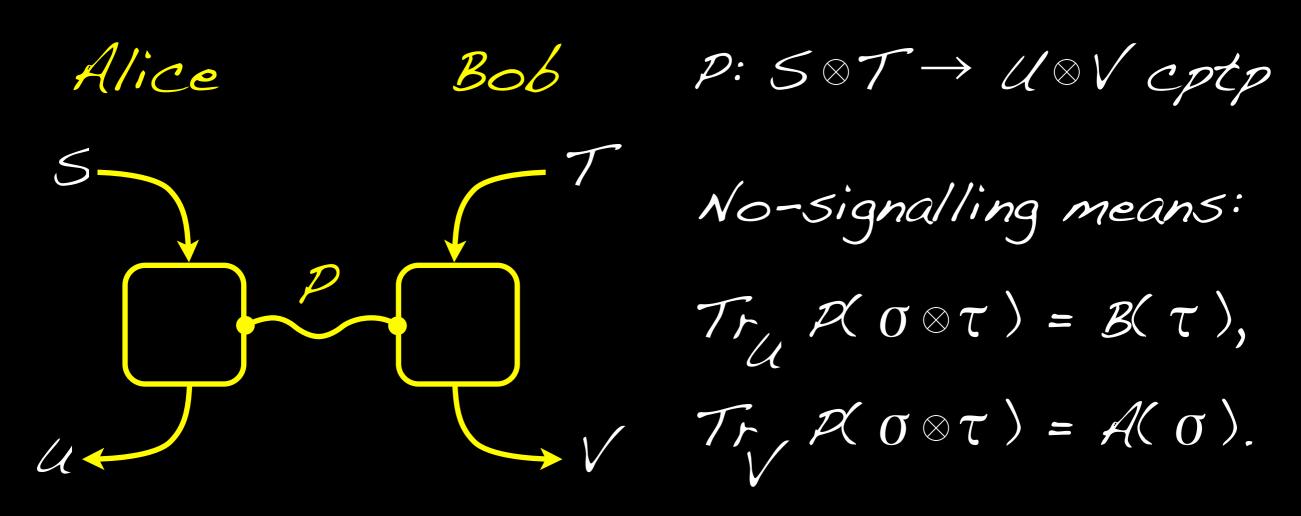
Define  $K = Span \{E, \} \subset B(A \rightarrow B)$  and  $S = K^{\dagger}K = Span \{E, E, \} \subset B(A)$ .

Can show: Zero-error transmission assisted by entanglement (or without) depends only on S.

Below treat assistance by quantum nosignalling correlations, which will turn out to depend only on K.

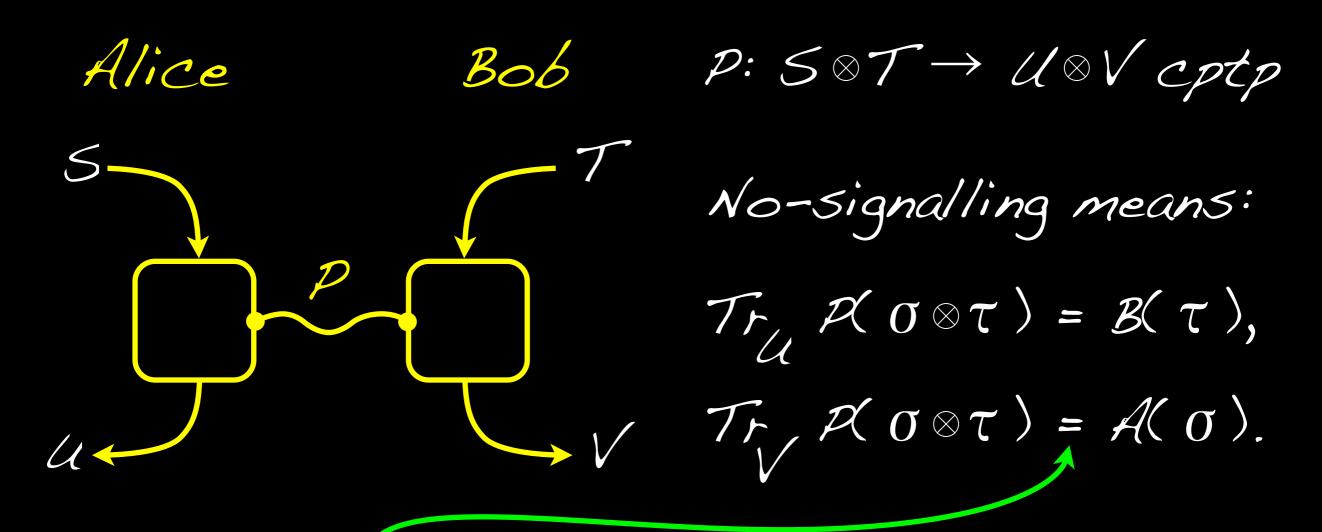
> [R. Duan/S. Severini/AW, IEEE-IT 59(2):1164-1174, 2013; R. Duan/AW, arXiv:1409.3426]

# Cf. W. Matthews' Quantum no-signalling: talk on Wed!



LD. Beckman et al., PRA 64:052309, 2001; T. Eggeling et al., Europhy. Lett. 57(6):782-788, 2002; M. Piani et al., PRA 74:012305, 2006]

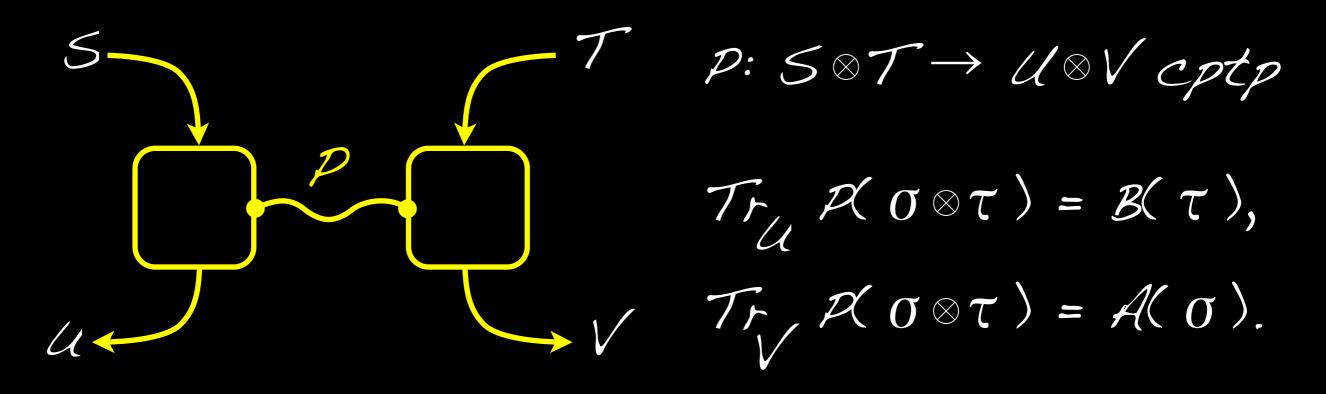
# Cf. W. Matthews' Quantum no-signalling: talk on Wed!



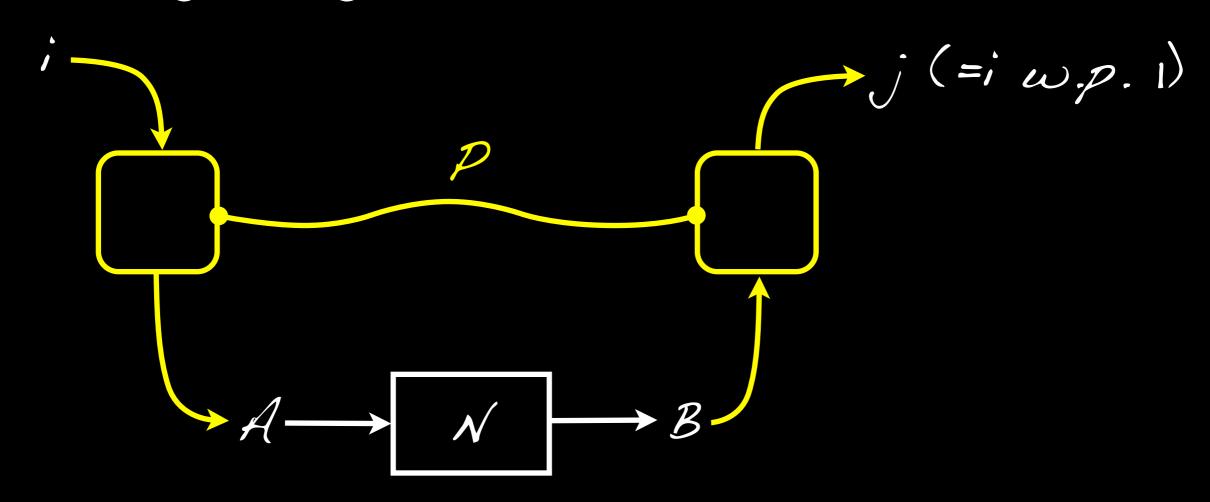
Equiv: Plinear combination of A. B. plus semidef. constraint for "cptp"

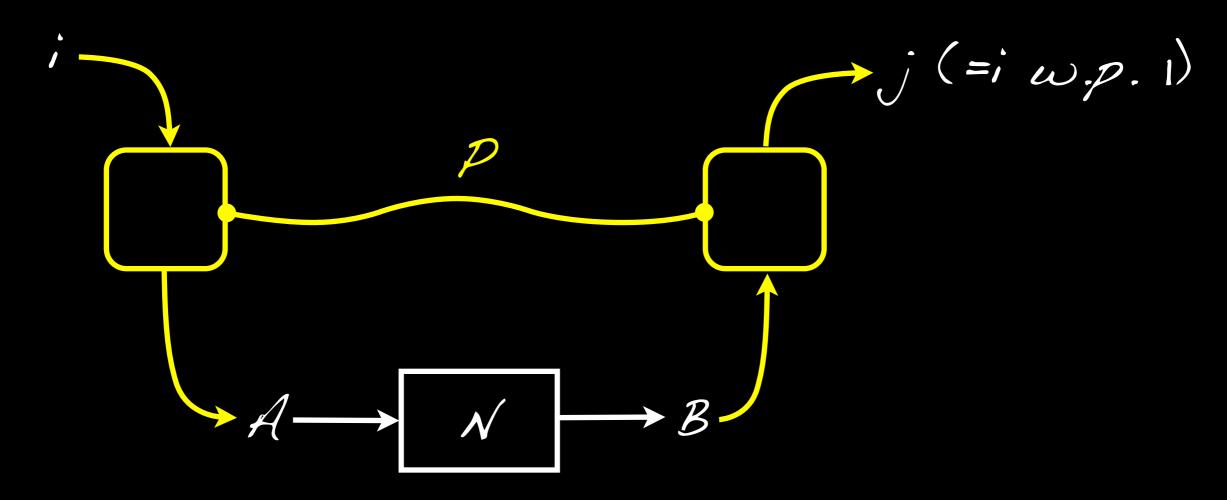
[D. Beckman et al., PRA 64:052309, 2001; T. Eggeling et al., Europhy. Lett. 57(6):782-788, 2002; M. Piani et al., PRA 74:012305, 2006]

# Cf. W. Matthews' Quantum no-signalling: talk on Wed!

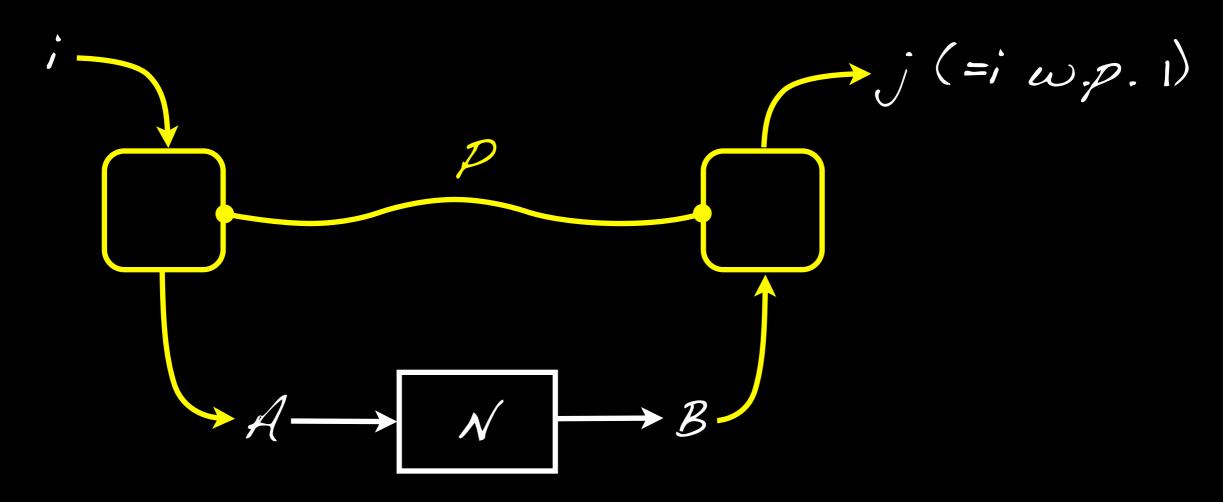


Although formally a channel with two simultaneous inputs, the no-signalling condition ensures that Alice can use her box without waiting. Bob is left with a conditional channel...



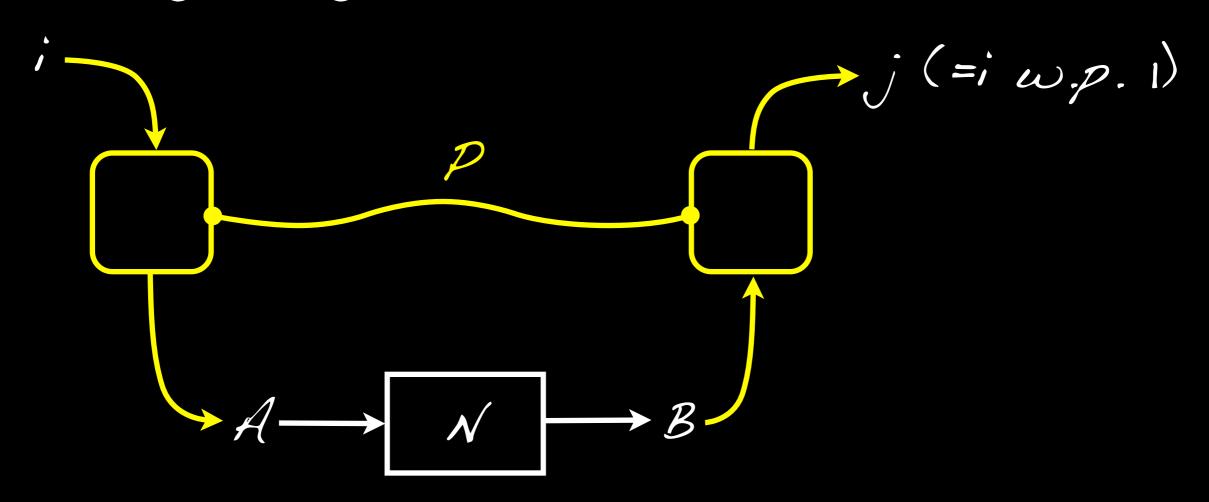


Maximum #msg. with  $P \in NS =: \overline{\alpha}(K)$ 



Maximum #msg. with  $P \in NS =: \overline{\alpha}(K)$ 

Similar definitions of  $\alpha(S)$  and  $\widetilde{\alpha}(S)$  via max. # of messages; all reducing to previous notions for classical channels.

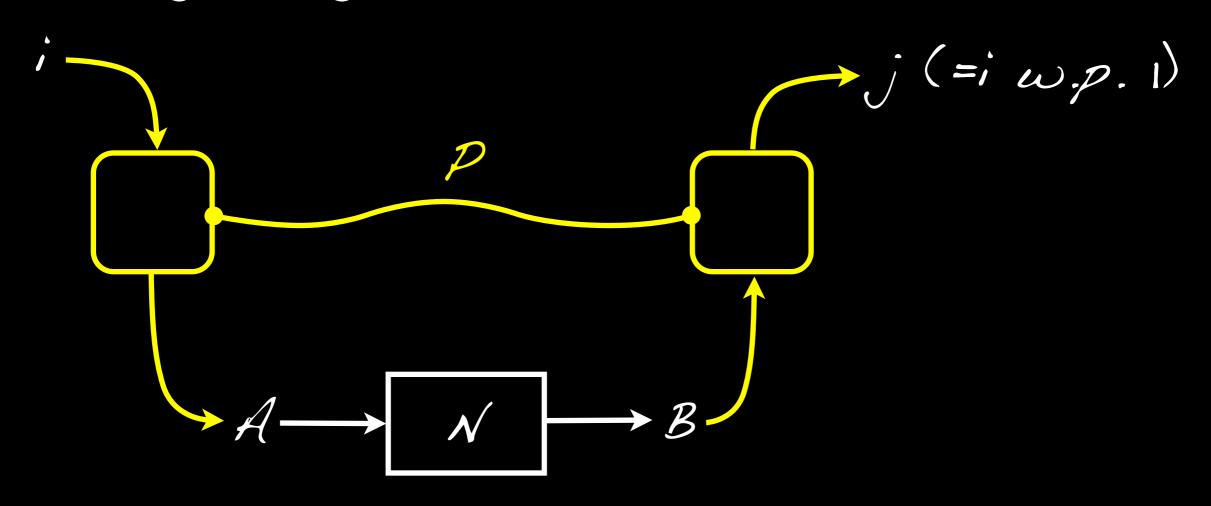


Thm. 
$$[RD/AW] \overline{\alpha}(K) = [\Upsilon(K)], \text{ where}$$

$$\Upsilon(K) = \max Tr \leq s.t. \ o \leq \mathcal{U}^{AB} \leq s \otimes 1,$$

$$Tr_{A} \mathcal{U} = 1^{B},$$

$$\Pi(s \otimes 1 - \mathcal{U}) = o.$$



Thm. 
$$[RD/AW] \overline{\alpha}(K) = [\Upsilon(K)], \text{ where}$$

$$\Upsilon(K) = \max \text{ Tr } S \text{ s.t. } 0 \leq \mathcal{U}^{AB} \leq S \otimes 1,$$

$$\Pi: \text{ support projection } Tr_{A} \mathcal{U} = 1^{B},$$
of Choi matrix of  $N \qquad \Pi(S \otimes 1 - \mathcal{U}) = 0.$ 

$$\Upsilon(K) = \max Tr \leq s.t. \ o \leq \mathcal{U}^{AB} \leq S \otimes 1,$$
$$Tr_{A} \mathcal{U} = 1^{B},$$
$$\Pi(S \otimes 1 - \mathcal{U}) = 0.$$

...reduces to classical fractional packing number for classical channel.

[T.S. Cubitt et al., IEEE-IT 57(8):5509-5523, 2011]

$$\Upsilon(K) = \max Tr \leq s.t. \ o \leq \mathcal{U}^{AB} \leq s \otimes 1,$$
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...reduces to classical fractional packing number for classical channel.

However, in general much more complex; for instance not multiplicative, i.e.  $\Upsilon(K \otimes K') \geq \Upsilon(K) \Upsilon(K'), \text{ sometimes strict.}$ 

$$\Upsilon(K) = \max Tr \leq s.t. \ o \leq \mathcal{U}^{AB} \leq S \otimes 1,$$

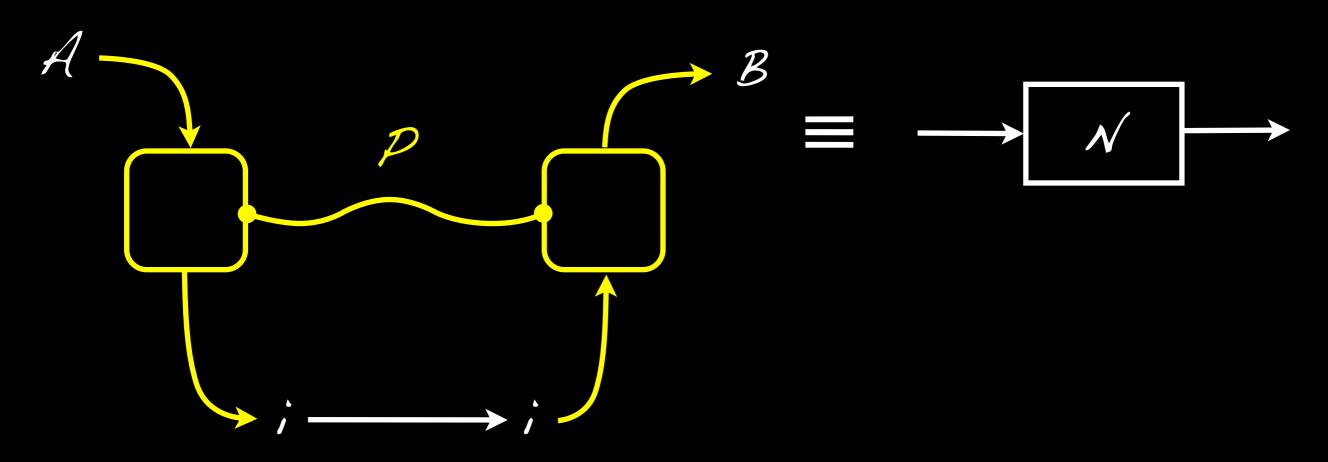
$$Tr_{A} \mathcal{U} = 1^{B},$$

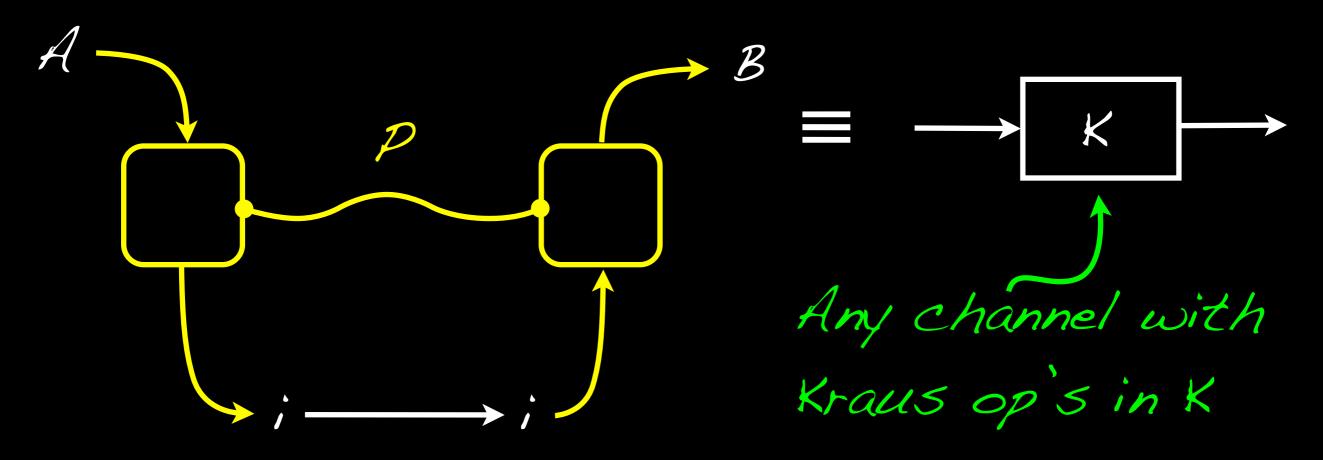
$$\Pi(S \otimes 1 - \mathcal{U}) = 0.$$

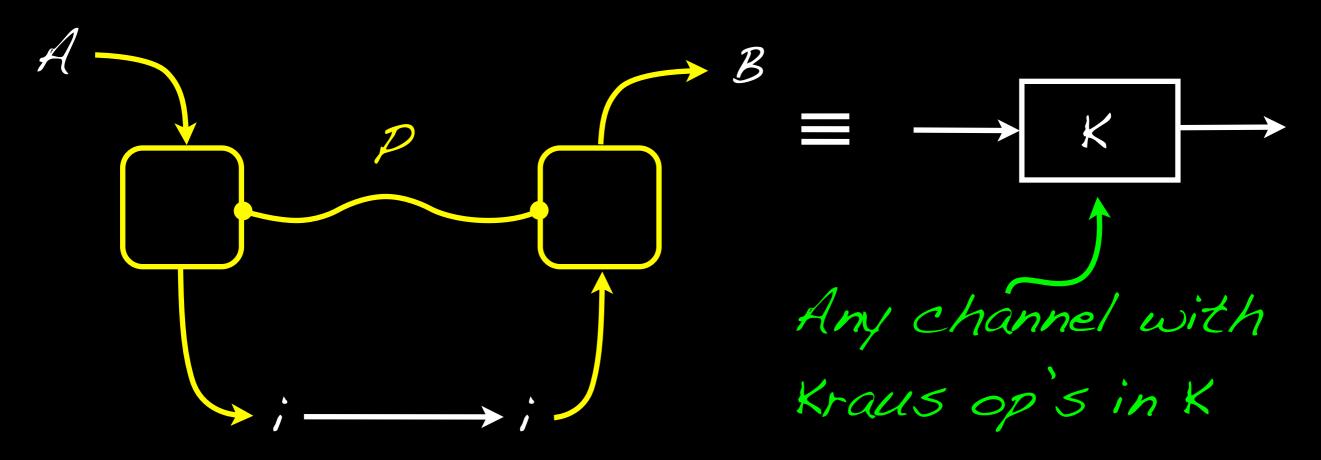
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However, in general much more complex; for instance not multiplicative, i.e.  $\Upsilon(K \otimes K') \geq \Upsilon(K) \Upsilon(K'), \text{ sometimes strict.}$ 

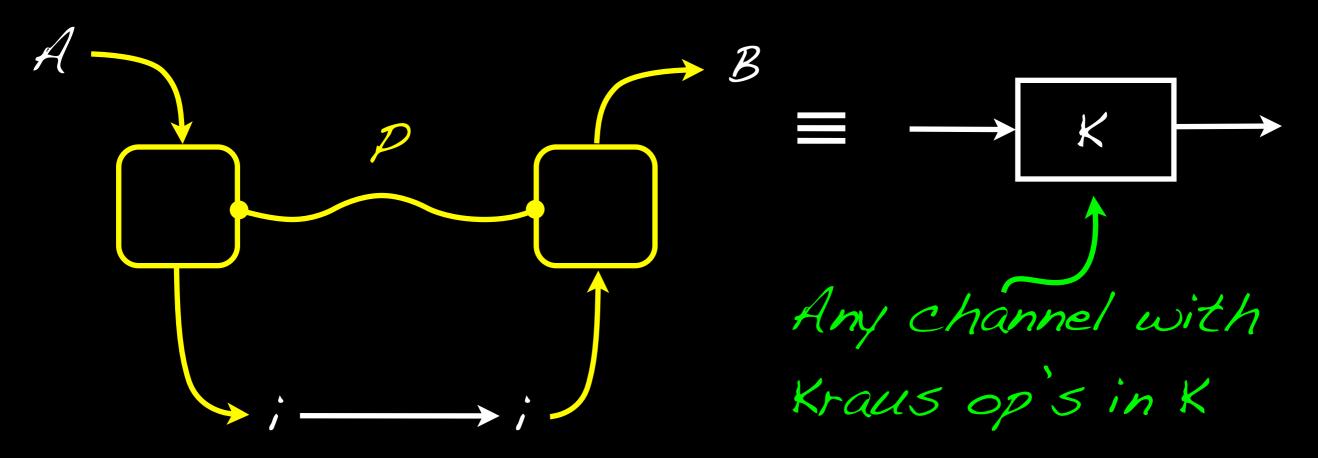
What is  $C_{ONS}(K) = \lim_{n \to \infty} \frac{1}{n} \log \Upsilon(K^{\otimes n})$ ?







Thm. 
$$[RD/AW]$$
 Min  $\# msg = [\Sigma(K)], \omega/$   
 $\Sigma(K) = min Tr T s.t. o \leq V^{AB} \leq 1 \otimes T,$   
 $Tr_{B}V = 1^{A}, \Pi^{\perp}V = o.$ 



Thm. 
$$[RD/AW]$$
 Min  $\# msg = [\Sigma(K)], \omega/\Sigma(K) = min Tr T s.t.  $0 \le V^{AB} \le 1 \otimes T,$ 

$$Tr_{B}V = 1^{A}, \Pi^{\perp}V = 0.$$$ 

...reduces to  $\alpha^*(\Gamma)$  for classical channels.

# 4. Cg-channels

These are still "very classical", e.g. have confusability graph G,  $x\sim x$  iff  $\Pi_x \Pi_{x'} \neq 0$ .

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These are still "very classical", e.g. have confusability graph G,  $x \sim x$  iff  $\Pi_x \Pi_{x'} \neq 0$ .  $\Pi = \sum_{x} |x > 4 \times x| \otimes \Pi_x \text{ Choi matrix support projection, simplifies SDP } \Upsilon(K)...$ 

$$\Upsilon(K) = \max Tr \leq s.t. \ o \leq \mathcal{U}^{AB} \leq s \otimes 1,$$
$$Tr_{A} \mathcal{U} = 11^{B},$$
$$\Pi(s \otimes 1 - \mathcal{U}) = 0.$$

$$\Upsilon(K) = \max \sum_{x} s_{x} \text{ s.t. } 0 \leq R_{x} \leq s_{x} \Pi_{x}^{\perp},$$
$$\sum_{x} (R_{x} + s_{x} \Pi_{x}) = 1.$$

$$\Upsilon(K) = \max \sum_{x} s_{x} \text{ s.t. } 0 \leq R_{x} \leq s_{x} \Pi_{x}^{\perp},$$
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$$\leq A(K) := max \sum_{x} s_{x} s.t. o \leq s_{x},$$

$$\sum_{x} s_{x} \Pi_{x} \leq 1.$$

$$\Upsilon(\mathbf{K}) = \max \sum_{x} s_{x} s.t. \ 0 \le R_{x} \le s_{x} \Pi_{x}^{\perp},$$

$$\sum_{x} (R_{x} + s_{x} \Pi_{x}) = 1.$$

$$\leq A(K) := max \sum_{x} s_{x} s.t. o \leq s_{x},$$

$$\sum_{x} s_{x} \Pi_{x} \leq 1.$$

Semidefinite packing number; also reduces to fractional packing no. in classical case, but is multiplicative:  $A(K \otimes K') = A(K)A(K')$ .

$$\Upsilon(K) = \max \sum_{x} s_{x} s.t. \ o \leq R_{x} \leq s_{x} \Pi_{x}^{\perp},$$

$$\sum_{x} (R_{x} + s_{x} \Pi_{x}) = 1.$$

$$\leq A(K) := max \sum_{x} s_{x} s.t. o \leq s_{x},$$

$$\sum_{x} s_{x} \Pi_{x} \leq 1.$$

Thm. 
$$C_{ONS}(K) = \lim_{n \to \infty} \frac{1}{n} \log \Upsilon(K^{\otimes n}) = \log A(K)$$
.

$$\Upsilon(K) = \max \sum_{x} s_{x} \text{ s.t. } 0 \leq R_{x} \leq s_{x} \Pi_{x}^{\perp},$$

$$\sum_{x} (R_{x} + s_{x} \Pi_{x}) = 1.$$

$$\leq \mathsf{A}(\mathsf{K}) := \max \sum_{x} s_{x} s.t. \ o \leq s_{x},$$
 
$$\sum_{x} s_{x} \Pi_{x} \leq 1.$$

Thm.  $C_{ONS}(K) = \lim_{n \to \infty} \frac{1}{n} \log \Upsilon(K^{\otimes n}) = \log A(K)$ .

Show actually  $\Upsilon(K^{\otimes n}) \geq n^{-c} A(K)^n$ , starting from an optimal solution for A(K); then by group (permutation) symmetry that we can satisfy the extra constraints loosing little..

$$\Upsilon(K) = \max \sum_{x} s_{x} s.t. \ o \leq R_{x} \leq s_{x} \Pi_{x}^{\perp},$$
$$\sum_{x} (R_{x} + s_{x} \Pi_{x}) = 1.$$

$$\leq A(K) := max \sum_{x} s_{x} s.t. o \leq s_{x},$$

$$\sum_{x} s_{x} \Pi_{x} \leq 1.$$

Thm.  $C_{ONS}(K) = \lim_{n \to \infty} \frac{1}{n} \log \Upsilon(K^{\otimes n}) = \log A(K)$ .

Thm. 
$$G_{oNS}(K) = log \Sigma(K)$$
 asympt. simul. cost 
$$\Sigma(K) = min Tr T s.t. \ 0 \le V_{\chi} \le T,$$
 
$$Tr V_{\chi} = I, V_{\chi} \le \Pi_{\chi}.$$

[R. Duan/AW, arXiv:1409.3426]

#### Example: Two-pure-state cg-channel

#### Example: Two-pure-state cg-channel

$$\frac{0}{1 \psi_{0}} >= \alpha 10 > + \beta 11 > K = span \ge 1 \psi_{0} > < 01,$$

$$\frac{1}{1 \psi_{1}} >= \alpha 10 > - \beta 11 > 1 \psi_{1} > < 11 \stackrel{?}{3}$$

$$\frac{1}{1 \psi_{1}} >= \alpha 10 > - \beta 11 > (1 > \alpha > \beta > 0; \alpha^{2} + \beta^{2} = 1)$$

$$\Upsilon(K) = 1$$
, but  $\Upsilon(K \otimes K) \ge 1/(\alpha^4 + \beta^4)$ , and for  $1 \text{ large enough}, \Upsilon(K^{\otimes n}) \ge 1/(\alpha^{2n} + \beta^{2n})$ .

#### Example: Two-pure-state cg-channel

$$\frac{0}{1 \psi_{0}} = \alpha 10 + \beta 11 \times \text{K=span} \{1 \psi_{0} > < 01, \\
1 \psi_{1} > = \alpha 10 > -\beta 11 \times 1 \psi_{1} > < 11\}$$

$$(1>\alpha>\beta>0; \alpha^{2}+\beta^{2}=1)$$

 $\Upsilon(K) = 1$ , but  $\Upsilon(K \otimes K) \ge 1/(\alpha^4 + \beta^4)$ , and for n large enough,  $\Upsilon(K^{\otimes n}) \ge 1/(\alpha^{2n} + \beta^{2n})$ . Easy:  $A(K) = 1/\alpha^2$ ,  $\Sigma(K) = 1 + 2\alpha\beta$ .

## 5. Lovász number encore

Now the best: Minimize A(K) over all cg-channels with the same confusability graph  $G(x\sim x')$  iff  $\Pi_{\chi} \not\perp \Pi_{\chi'}$ .



## 5. Lovász number encore

Now the best: Minimize A(K) over all cg-channels with the same confusability graph  $G(x\sim x')$  iff  $\Pi_{\chi} \not\perp \Pi_{\chi'}$ .

Thm. min  $A(K) = \vartheta(G)$ ; min  $C_{ONS}(K) = log \vartheta(G)$ .

In words: Lovász' number gives the nosignalling assisted capacity of the worst cg-channel with confusability graph G.

First capacity interpretation of  $\vartheta(G)$ :-)

[R. Duan/AW, arXiv:1409.3426]

### 6. Last words:

- SDP formulas for assisted capacity and simulation cost (one-shot)
- SDP can regularize to a relaxed SDP :-)
- Capacity interpretation of Lovász number
- Gap between  $C_{o\varepsilon}(G)$  and  $\log \vartheta(G)$ ?
- Regularization necessary? There could be K such that  $Y(K)=\vartheta(G)$  cf. Ching-Yi Lai's poster on Monday!
- $\Sigma(G) := \min \{ \Sigma(K) : G \supset K^{\dagger}K \} = ??$ Know only: between  $\vartheta(G)$  and  $\alpha^*(G)$