

Random quantum channels and additivity violations

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Outline of the talk

- 1 Random quantum channels and their minimum output entropy
- 2 Lower bounding $H^{\min}(\Phi \otimes \bar{\Phi})$
- 3 Computing $H^{\min}(\Phi)$
- 4 Additivity violations

Random quantum channels
and their minimum output entropy

Additivity for MOE of quantum channels

- **Quantum channels**: CPTP maps $\Phi : \mathcal{M}_{\text{in}}(\mathbb{C}) \rightarrow \mathcal{M}_{\text{out}}(\mathbb{C})$.
- Rényi entropies

$$p > 0 \quad H^p(\rho) = \frac{\log \text{Tr} \rho^p}{1-p}, \quad H^1(\rho) = H(\rho) = -\text{Tr}(\rho \log \rho).$$

- p -Minimal Output Entropy of a quantum channel

$$\begin{aligned} H_{\min}^p(\Phi) &= \min_{\rho \in \mathcal{M}_{\text{in}}^{1,+}(\mathbb{C})} H^p(\Phi(\rho)) \\ &= \min_{x \in \mathbb{C}^{\text{in}}} H^p(\Phi(P_x)). \end{aligned}$$

- Is the p -MOE **additive** ?

$$H_{\min}^p(\Phi \otimes \Psi) = H_{\min}^p(\Phi) + H_{\min}^p(\Psi) \quad \forall \Phi, \Psi.$$

- **NO !!!**

- $p > 1$: Hayden + Winter '08;
- $p = 1$: Hastings '08

Importance of additivity

- Simple formula for the (classical) capacity of quantum channels: if additivity holds, then there is no need to use inputs entangled over multiple uses of Φ .
- Shor '04 equivalence of additivity questions
 - 1 additivity of MOE
 - 2 additivity of the Holevo capacity χ ($= C_{\otimes}$ in Andreas' talk)
 - 3 (strong super-) additivity of the entanglement of formation E_F .
- Additivity proved for some particular channels: unital qubit, depolarizing, entanglement breaking, etc.
- Holevo-Werner channel violates additivity of the p -Rényi entropy for $p > 4.79$. No known deterministic examples for $p = 1$ or p close to 1.
- Difficult, mathematically challenging problem.

Random quantum channels

- Counterexamples to additivity conjectures are **random**.
- Random quantum channels from **random isometries**

$$\Phi(\rho) = \text{Tr}_{\text{anc}}(V\rho V^*),$$

where V is a Haar partial isometry

$$V : \mathbb{C}^{\text{in}} \rightarrow \mathbb{C}^{\text{out}} \otimes \mathbb{C}^{\text{anc}}.$$

Equivalently, via the Stinespring dilation theorem

$$\Phi(\rho) = \text{Tr}_{\text{anc}}(U(\rho \otimes P_y)U^*),$$

where $y \in \mathbb{C}^{\frac{\text{out} \times \text{anc}}{\text{in}}}$ and $U \in \mathcal{M}_{\text{out} \times \text{anc}}(\mathbb{C})$ is a Haar unitary matrix.

- Random quantum channels from **i.i.d. random unitary matrices** (random mixed unitary channels)

$$\Phi(\rho) = \sum_{i=1}^k p_i U_i \rho U_i^*,$$

for (random) probabilities p_i and i.i.d. Haar distributed unitary operators U_i .

Model of interest

Here, we focus on random quantum channels coming from random isometries, with the following parameters.

- $\text{in} = tnk$,
- $\text{out} = k$,
- $\text{anc} = n$,

where $n, k \in \mathbb{N}$ and $t \in (0, 1)$. In general, we shall assume that

- $n \rightarrow \infty$
- k is fixed
- t is fixed.

In other words, we are interested in $\Phi : \mathcal{M}_{tnk}(\mathbb{C}) \rightarrow \mathcal{M}_k(\mathbb{C})$,

$$\Phi(\rho) = [\text{id}_k \otimes \text{Tr}_n](V\rho V^*),$$

where V is a random isometry obtained by keeping the first tnk columns of a $nk \times nk$ Haar random unitary.

How to get counterexamples ?

- Choose Φ to be random and $\Psi = \bar{\Phi}$; this way, $H_{\min}^p(\Psi) = H_{\min}^p(\Phi)$.
- Bound

$$H_{\min}^p(\Phi \otimes \bar{\Phi}) \leq B_2 < 2B_1 \leq 2H_{\min}^p(\Phi).$$

Lower bounding $H^{\min}(\Phi \otimes \bar{\Phi})$

- Remember: we want

$$H_{\min}^p(\Phi \otimes \bar{\Phi}) \leq B_2 < 2B_1 \leq 2H_{\min}^p(\Phi).$$

- Use trivial bound $H_{\min}^p(\Phi \otimes \bar{\Phi}) \leq H^p([\Phi \otimes \bar{\Phi}](X_{12}))$, for a particular choice of $X_{12} \in \mathcal{M}_{tnk}(\mathbb{C}) \otimes \mathcal{M}_{tnk}(\mathbb{C})$.
- $X_{12} = X_1 \otimes X_2$ do not yield counterexamples \Rightarrow choose a **maximally entangled state**

$$X_{12} = E_{tnk} = \left(\frac{1}{\sqrt{tnk}} \sum_{i=1}^{tnk} e_i \otimes e_i \right) \left(\frac{1}{\sqrt{tnk}} \sum_{j=1}^{tnk} e_j \otimes e_j \right)^*.$$

- Bound entropies of the (random) density matrix

$$Z_n = [\Phi \otimes \bar{\Phi}](E_{tnk}) \in \mathcal{M}_k(\mathbb{C}) \otimes \mathcal{M}_k(\mathbb{C}).$$

Main result - finite rank output

Theorem (Collins + N. '09)

For all k, t , almost surely as $n \rightarrow \infty$, the eigenvalues of $Z_n = [\Phi \otimes \bar{\Phi}](E_{tnk})$ converge to

$$\left(t + \frac{1-t}{k^2}, \underbrace{\frac{1-t}{k^2}, \dots, \frac{1-t}{k^2}}_{k^2-1 \text{ times}} \right) \in \Delta_{k^2}.$$

- Previously known bound (deterministic, comes from linear algebra): for all t, n, k , the largest eigenvalue of Z_n is at least t .
- Two improvements:
 - 1 “better” largest eigenvalue,
 - 2 knowledge of the whole spectrum.
- Precise knowledge of eigenvalues \rightsquigarrow **optimal** estimates for entropies.
- However, smaller eigenvalues are the “worst possible”.

Proof strategy for a.s. spectrum Z_n

- Use the **method of moments**

- ④ Convergence in moments:

$$\mathbb{E}\mathrm{Tr}(Z_n^p) \rightarrow \left(t + \frac{1-t}{k^2}\right)^p + (k^2 - 1) \left(\frac{1-t}{k^2}\right)^p;$$

- ② Borel-Cantelli for a.s. convergence:

$$\sum_{n=1}^{\infty} \mathbb{E} \left[(\mathrm{Tr}(Z_n^p) - \mathbb{E}\mathrm{Tr}(Z_n^p))^2 \right] < \infty.$$

- We need to compute moments $\mathbb{E} [\mathrm{Tr}(Z_n^{p_1})^{q_1} \dots \mathrm{Tr}(Z_n^{p_s})^{q_s}]$.
- Use the **Weingarten formula** to compute the unitary averages.

Unitary integration - Weingarten formula

- Using matrix coordinates, we can reduce our problem to computing integrals over the unitary group.

Theorem (Weingarten formula)

Let d be a positive integer and (i_1, \dots, i_p) , $(i'_1, \dots, i'_{p'})$, (j_1, \dots, j_p) , $(j'_1, \dots, j'_{p'})$ be p -tuples of positive integers from $\{1, 2, \dots, d\}$. Then

$$\int_{\mathcal{U}(d)} U_{i_1 j_1} \cdots U_{i_p j_p} \overline{U_{i'_1 j'_1}} \cdots \overline{U_{i'_{p'} j'_{p'}}} dU = \sum_{\alpha, \beta \in \mathcal{S}_p} \delta_{i_1 i'_{\alpha(1)}} \cdots \delta_{i_p i'_{\alpha(p)}} \delta_{j_1 j'_{\beta(1)}} \cdots \delta_{j_p j'_{\beta(p)}} \text{Wg}(d, \alpha \beta^{-1}).$$

If $p \neq p'$ then

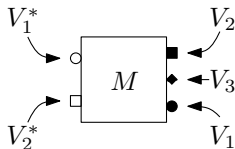
$$\int_{\mathcal{U}(d)} U_{i_1 j_1} \cdots U_{i_p j_p} \overline{U_{i'_1 j'_1}} \cdots \overline{U_{i'_{p'} j'_{p'}}} dU = 0.$$

- There is a **graphical** way of reading this formula on the diagrams !

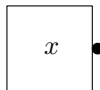
Boxes & wires

- Graphical formalism inspired by works of Penrose, Coecke, Jones, etc.

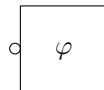
- Tensors \rightsquigarrow **decorated** boxes.



$$M \in V_1 \otimes V_2 \otimes V_3 \otimes V_1^* \otimes V_2^*$$

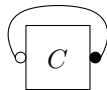
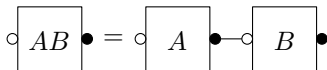


$$x \in V_1$$

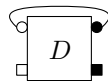


$$\varphi \in V_1^*$$

- Tensor **contractions** (or traces) $V \otimes V^* \rightarrow \mathbb{C} \rightsquigarrow$ wires.

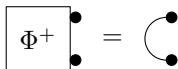


$$\text{Tr}(C)$$



$$\text{Tr}_{V_1}(D)$$

- Maximally entangled vector** $\text{Bell} = \sum_{i=1}^{\dim V} e_i \otimes e_i \in V \otimes V$

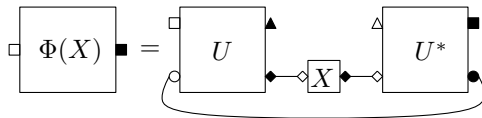


Graphical representation of quantum channels

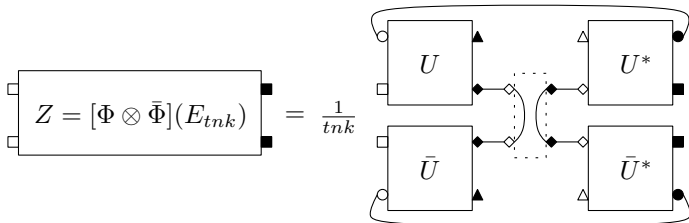
- Decorations/labels

$$\begin{array}{cccc} \bullet & \blacksquare & \blacklozenge & \blacktriangle \\ \circ = \mathbb{C}^n & \square = \mathbb{C}^k & \diamond = \mathbb{C}^{tnk} & \triangle = \mathbb{C}^{t^{-1}} \end{array}$$

- Single channel (finite rank output)



- Product of conjugate channels



“Graphical” Weingarten formula: graph expansion

Consider a diagram \mathcal{D} containing random unitary matrices/boxes U and U^* . Apply the following **removal** procedure:

- 1 Start by replacing U^* boxes by \bar{U} boxes (by reversing decoration shading).
- 2 By the (algebraic) Weingarten formula, if the number p of U boxes is different from the number of \bar{U} boxes, then $\mathbb{E}\mathcal{D} = 0$.
- 3 Otherwise, choose a pair of permutations $(\alpha, \beta) \in \mathcal{S}_p^2$. These permutations will be used to pair decorations of U/\bar{U} boxes.
- 4 For all $i = 1, \dots, p$, add a wire between each white decoration of the i -th U box and the corresponding white decoration of the $\alpha(i)$ -th \bar{U} box. In a similar manner, use β to pair black decorations.
- 5 Erase all U and \bar{U} boxes. The resulting diagram is denoted by $\mathcal{D}_{(\alpha, \beta)}$.

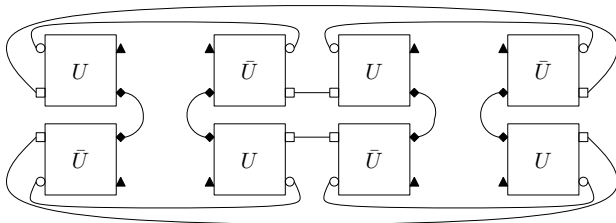
Theorem

$$\mathbb{E}\mathcal{D} = \sum_{\alpha, \beta \in \mathcal{S}_p} \mathcal{D}_{(\alpha, \beta)} \text{Wg}(d, \alpha\beta^{-1}).$$

Example: $\mathbb{E}\text{Tr}(Z^2)$

- We have to compute a sum over all pairings of 4 “ U ” boxes with 4 “ \bar{U} ” boxes.
- Diagrams associated to pairings are indexed by 2 permutations $(\alpha, \beta) \in \mathcal{S}_4^2$. Consider the permutation $\delta = (1\ 4)(2\ 3) \in \mathcal{S}_4$.

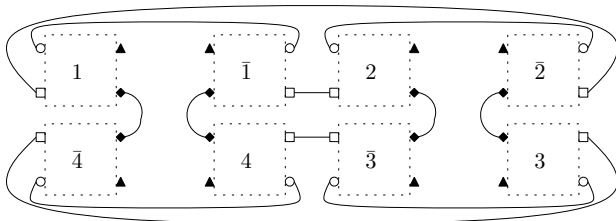
The original diagram



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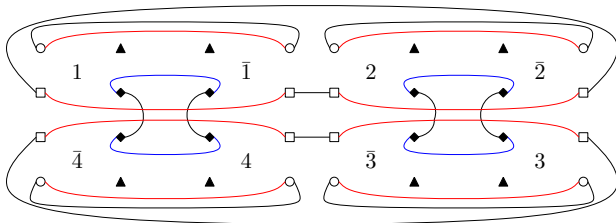
The diagram with the boxes removed



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The wiring for $\alpha = \beta = \text{id}$.

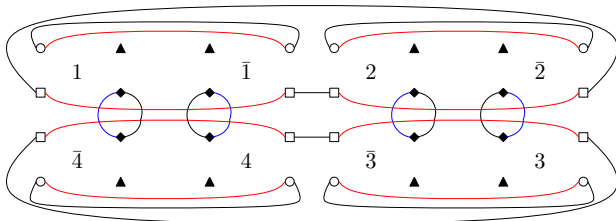


Contribution: $n^4 \cdot k^2 \cdot (tnk)^2 \cdot \text{Wg}(\text{id})$.

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The wiring for $\alpha = \text{id}$, $\beta = \delta$.

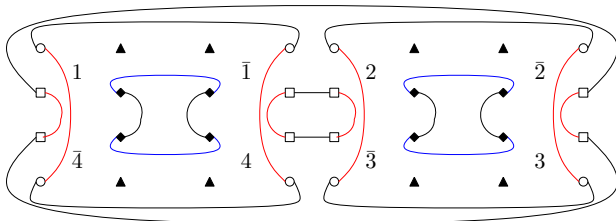


Contribution: $n^4 \cdot k^2 \cdot (tnk)^4 \cdot \text{Wg}(\delta)$.

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The wiring for $\alpha = \delta$, $\beta = \text{id}$.

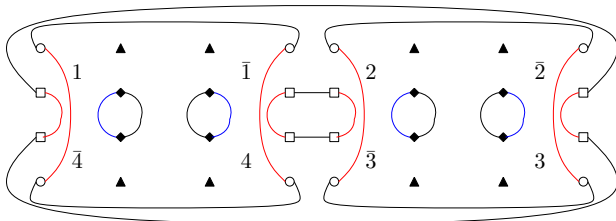


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- Diagrams associated to pairings are indexed by 2 permutations $(\alpha, \beta) \in \mathcal{S}_4^2$. Consider the permutation $\delta = (1\ 4)(2\ 3) \in \mathcal{S}_4$.

The wiring for $\alpha = \beta = \delta$.



Contribution: $n^2 \cdot k^2 \cdot (tnk)^4 \cdot \text{Wg}(\text{id})$.

Sketch of the proof

- We want to compute, for all $p \geq 1$, $\mathbb{E}\text{Tr}(Z^p)$.
- One needs to compute the contribution of each diagram $\mathcal{D}_{(\alpha,\beta)}$, where $\alpha, \beta \in \mathcal{S}_{2p}$.
- $\mathcal{D}_{(\alpha,\beta)}$ is a collection of loops associated to vector spaces of dimensions n , k , and tnk .
- Asymptotic for Weingarten weights ($\sigma \in \mathcal{S}_p$, $d \rightarrow \infty$, p fixed):

$$\text{Wg}(d, \sigma) = d^{-(p+|\sigma|)}(\text{Mob}(\sigma) + O(d^{-2})).$$

- One has to identify asymptotically dominating terms. Computations for fixed n are intractable due to the complexity of the Weingarten function. In the limit $n \rightarrow \infty$, the structure of the dominating terms is very simple.

Theorem (Collins + N. '09)

For all k, t , almost surely as $n \rightarrow \infty$,

$$\text{spec}(Z_n) \rightarrow \left(t + \frac{1-t}{k^2}, \underbrace{\frac{1-t}{k^2}, \dots, \frac{1-t}{k^2}}_{k^2-1 \text{ times}} \right) \in \Delta_{k^2}.$$

Computing $H^{\min}(\Phi)$

Strategy for B_1

- Remember: we want

$$H_{\min}^p(\Phi \otimes \bar{\Phi}) \leq B_2 < 2B_1 \leq 2H_{\min}^p(\Phi).$$

- We shall do more: we compute **the exact limit** (as $n \rightarrow \infty$) of $H_{\min}^p(\Phi)$.

Theorem (Belinschi, Collins, N. '13)

For all $p \geq 1$,

$$\lim_{n \rightarrow \infty} H_p^{\min}(\Phi) = H_p(a, b, b, \dots, b),$$

where a, b do not depend on p , $b = (1 - a)/(k - 1)$ and $a = \varphi(1/k, t)$ with

$$\varphi(s, t) = \begin{cases} s + t - 2st + 2\sqrt{st(1-s)(1-t)} & \text{if } s + t < 1; \\ 1 & \text{if } s + t \geq 1. \end{cases}$$

Entanglement of a vector

For a vector

$$x = \sum_{i=1}^k \sqrt{\lambda_i(x)} e_i \otimes f_i,$$

define $H(x) = H(\lambda(x)) = H(\rho) = -\sum_i \lambda_i(x) \log \lambda_i(x)$, the **entropy of entanglement** of the bipartite pure state x .

Note that

- 1 The state x is **separable**, $x = e \otimes f$, iff. $H(x) = 0$.
- 2 The state x is **maximally entangled**, $x = k^{-1/2} \sum_i e_i \otimes f_i$, iff. $H(x) = \log k$.

Recall that we are interested in computing

$$\begin{aligned} H^{\min}(\Phi) &= \min_{x \in \mathbb{C}^d, \|x\|=1} H(\Phi(P_x)) = \min_{y \in \text{Im}V, \|y\|=1} H([\text{id}_k \otimes \text{Tr}_n]P_y) \\ &= \min_{y \in \text{Im}V, \|y\|=1} H(y). \end{aligned}$$

Entanglement of a subspace

For a subspace $V \subset \mathbb{C}^k \otimes \mathbb{C}^n$, define

$$H_p^{\min}(V) = \min_{y \in V, \|y\|=1} H_p(y),$$

the minimal entanglement of vectors in V .

Here, we abuse notation: recall that we are interested in random isometries $V : \mathbb{C}^{tnk} \rightarrow \mathbb{C}^k \otimes \mathbb{C}^n$. Since the quantities H_p^{\min} only depend on the range of V , also write $V = \text{ran} V$.

A subspace V is called **entangled** if $H^{\min}(V) > 0$, i.e. if it does not contain separable vectors $x \otimes y$.

Singular values of vectors from a subspace

↪ Entropy is just a statistic, look at **the set of all singular values** directly!

For a subspace $V \subset \mathbb{C}^k \otimes \mathbb{C}^n$ of dimension $\dim V = d$, define the set eigen-/singular values or Schmidt coefficients

$$K_V = \{\lambda(x) : x \in V, \|x\| = 1\}.$$

↪ Our goal is to **understand** K_V .

- The set K_V is a compact subset of the ordered probability simplex Δ_k^\downarrow .
- **Local invariance:** $K_{(U_1 \otimes U_2)V} = K_V$, for unitary matrices $U_1 \in \mathcal{U}(k)$ and $U_2 \in \mathcal{U}(n)$.
- **Monotonicity:** if $V_1 \subset V_2$, then $K_{V_1} \subset K_{V_2}$.
- Recovering minimum entropies:

$$H_p^{\min}(\Phi) = H_p^{\min}(V) = \min_{\lambda \in K_V} H_p(\lambda).$$

Examples

The **anti-symmetric subspace** provides the (explicit) counter-example for the additivity of the p -Rényi entropy [Grudka, Horodecki, Pankowski '09].

- Let $k = n$ and put $V = \Lambda^2(\mathbb{C}^n)$
- The subspace V is almost half of the total space:
 $\dim V = n(n-1)/2$.
- Example of a vector in V :

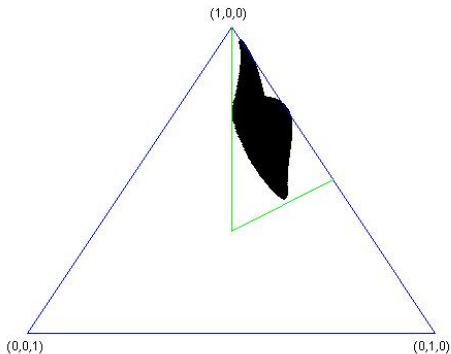
$$V \ni x = \frac{1}{\sqrt{2}}(e \otimes f - f \otimes e).$$

- **Fact:** singular values of vectors in V come in pairs.
- Hence, the least entropy vector in V is as above, with $e \perp f$ and $H(x) = \log 2$.
- Thus, $H^{\min}(V) = \log 2$ and one can show that

$$K_V = \{(\lambda_1, \lambda_1, \lambda_2, \lambda_2, \dots) \in \Delta_n : \lambda_i \geq 0, \sum_i \lambda_i = 1/2\}.$$

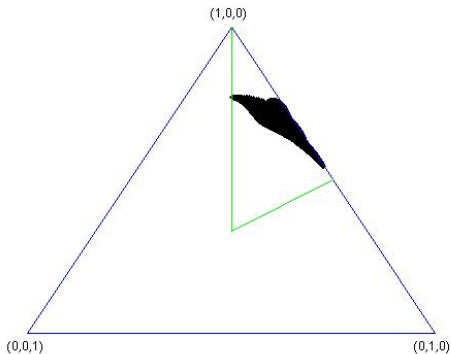
Examples - K_V

$V = \text{span}\{G_1, G_2\}$, where $G_{1,2}$ are 3×3 independent Ginibre random matrices.



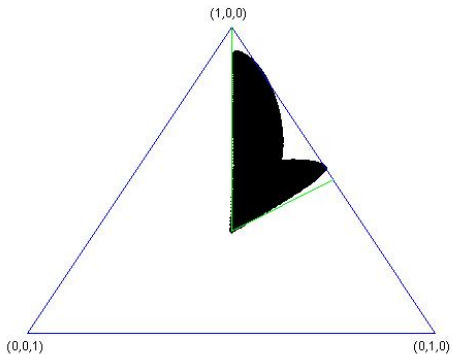
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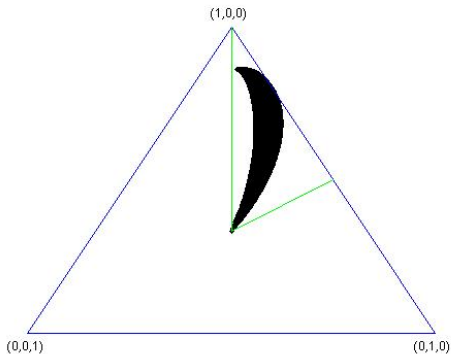
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An open problem

Find **explicit** (i.e. non-random) examples of subspaces $V \subset \mathbb{C}^k \otimes \mathbb{C}^n$ with

- 1 **large** $\dim V$;
- 2 **large** $H^{\min}(V)$.

Recall that we are interested in random isometries/subspaces in the following asymptotic regime: k fixed, $n \rightarrow \infty$, and $d \sim tkn$, for a fixed parameter $t \in (0, 1)$.

Theorem (Belinschi, Collins, N. '10)

For a sequence of uniformly distributed random subspaces V_n , the set K_{V_n} of singular values of unit vectors from V_n converges (almost surely, in the Hausdorff distance) to a **deterministic, convex** subset $K_{k,t}$ of the probability simplex Δ_k

$$K_{k,t} := \{\lambda \in \Delta_k \mid \forall x \in \Delta_k, \langle \lambda, x \rangle \leq \|x\|_{(t)}\}.$$

Corollary: exact limit of the minimum output entropy

By the previous theorem, in the specific asymptotic regime t, k fixed, $n \rightarrow \infty$, $d \sim tkn$, we have the following a.s. convergence result for random quantum channels Φ (defined via random isometries $V : \mathbb{C}^d \rightarrow \mathbb{C}^k \otimes \mathbb{C}^n$):

$$\lim_{n \rightarrow \infty} H_p^{\min}(\Phi) = \min_{\lambda \in K_{k,t}} H_p(\lambda).$$

It is not just a bound, the **exact limit value** is obtained.

Theorem (Belinschi, Collins, N. '13)

For all $p \geq 1$,

$$\lim_{n \rightarrow \infty} H_p^{\min}(\Phi) = \min_{\lambda \in K_{k,t}} H_p(\lambda) = H_p(a, b, b, \dots, b),$$

where a, b do not depend on p , $b = (1 - a)/(k - 1)$ and $a = \varphi(1/k, t)$ with

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Theorem (Voiculescu '98)

Let (A_n) and (B_n) be sequences of $n \times n$ matrices such that A_n and B_n converge in distribution (with respect to $n^{-1}\text{Tr}$) for $n \rightarrow \infty$.

Furthermore, let (U_n) be a sequence of Haar unitary $n \times n$ random matrices. Then, A_n and $U_n B_n U_n^*$ are **asymptotically free** for $n \rightarrow \infty$.

If A_n, B_n are matrices of size n , whose spectra converge towards μ_a, μ_b , the spectrum of $A_n + U_n B_n U_n^*$ converges to $\mu_a \boxplus \mu_b$; here, $\mu_a \boxplus \mu_b$ is the distribution of $a + b$, where $a, b \in (\mathcal{A}, \tau)$ are **free** random variables having distributions resp. μ_a, μ_b .

If A_n, B_n are matrices of size n such that $A_n \geq 0$, whose spectra converge towards μ_a, μ_b , the spectrum of $A_n^{1/2} U_n B_n U_n^* A_n^{1/2}$ converges to $\mu_a \boxtimes \mu_b$.

Example: truncation of random matrices

Let $P_n \in \mathcal{M}_n$ a projection of rank $n/2$; its eigenvalues are 0 and 1, with multiplicity $n/2$. Hence, the distribution of P_n converges, when $n \rightarrow \infty$, to the Bernoulli probability measure $\frac{1}{2}\delta_0 + \frac{1}{2}\delta_1$.

Let $C_n \in \mathcal{M}_{n/2}$ be the top $n/2 \times n/2$ **corner** of $U_n P_n U_n^*$, with U_n a Haar random unitary matrix. What is the distribution of C_n ?

Up to zero blocks, $C_n = Q_n(U_n P_n U_n^*)Q_n$, where Q_n is the diagonal orthogonal projection on the first $n/2$ coordinates of \mathbb{C}^n . The distribution of Q_n converges to $\frac{1}{2}\delta_0 + \frac{1}{2}\delta_1$.

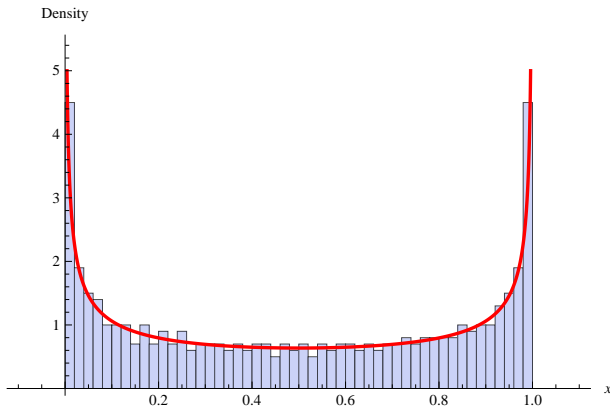
Free probability theory tells us that the distribution of C_n will converge to

$$\left(\frac{1}{2}\delta_0 + \frac{1}{2}\delta_1\right) \boxtimes \left(\frac{1}{2}\delta_0 + \frac{1}{2}\delta_1\right) = \frac{1}{\pi\sqrt{x(1-x)}} \mathbf{1}_{[0,1]}(x) dx,$$

which is the **arcsine distribution**.

Example: truncation of random matrices

Histogram of eigenvalues of a truncated randomly rotated projector of relative rank $1/2$ and size $n = 4000$; in red, the density of the arcsine distribution.



Definition

For a positive integer k , embed \mathbb{R}^k as a self-adjoint real subalgebra \mathcal{R} of a C^* -ncps (\mathcal{A}, τ) , so that $\tau(x) = (x_1 + \cdots + x_k)/k$. Let p_t be a projection of rank $t \in (0, 1]$ in \mathcal{A} , **free** from \mathcal{R} . On the real vector space \mathbb{R}^k , we introduce the following norm, called the **(t)-norm**:

$$\|x\|_{(t)} := \|p_t x p_t\|_\infty,$$

where the vector $x \in \mathbb{R}^k$ is identified with its image in \mathcal{R} .

- One can show that $\|\cdot\|_{(t)}$ is indeed a norm, which is permutation invariant.
- When $t > 1 - 1/k$, $\|\cdot\|_{(t)} = \|\cdot\|_\infty$ on \mathbb{R}^k .
- $\lim_{t \rightarrow 0^+} \|x\|_{(t)} = k^{-1} |\sum_i x_i|$.

Corners of randomly rotated projections

Theorem (Collins '05)

In \mathbb{C}^n , choose at random according to the Haar measure two independent subspaces V_n and V'_n of respective dimensions $q_n \sim sn$ and $q'_n \sim tn$ where $s, t \in (0, 1]$. Let P_n (resp. P'_n) be the orthogonal projection onto V_n (resp. V'_n). Then,

$$\lim_n \|P_n P'_n P_n\|_\infty = \varphi(s, t) = \sup \text{supp}((1-s)\delta_0 + s\delta_1) \boxtimes ((1-t)\delta_0 + t\delta_1),$$

with

$$\varphi(s, t) = \begin{cases} s + t - 2st + 2\sqrt{st(1-s)(1-t)} & \text{if } s + t < 1; \\ 1 & \text{if } s + t \geq 1. \end{cases}$$

Hence, we can compute

$$\| \underbrace{1, \dots, 1}_{j \text{ times}}, \underbrace{0, \dots, 0}_{k-j \text{ times}} \|_{(t)} = \varphi\left(\frac{j}{k}, t\right).$$

$K_{V_n} \rightarrow K_{k,t}$: idea of the proof

A simpler question: what is the largest maximal singular value $\max_{x \in V, \|x\|=1} \lambda_1(x)$ of vectors from the subspace V ?

$$\begin{aligned} \max_{x \in V, \|x\|=1} \lambda_1(x) &= \max_{x \in V, \|x\|=1} \lambda_1([\text{id}_k \otimes \text{Tr}_n]P_x) \\ &= \max_{x \in V, \|x\|=1} \|[\text{id}_k \otimes \text{Tr}_n]P_x\| \\ &= \max_{x \in V, \|x\|=1} \max_{y \in \mathbb{C}^k, \|y\|=1} \text{Tr} [([\text{id}_k \otimes \text{Tr}_n]P_x) \cdot P_y] \\ &= \max_{x \in V, \|x\|=1} \max_{y \in \mathbb{C}^k, \|y\|=1} \text{Tr} [P_x \cdot P_y \otimes I_n] \\ &= \max_{y \in \mathbb{C}^k, \|y\|=1} \max_{x \in V, \|x\|=1} \text{Tr} [P_x \cdot P_y \otimes I_n] \\ &= \max_{y \in \mathbb{C}^k, \|y\|=1} \|P_V \cdot P_y \otimes I_n \cdot P_V\|_\infty. \end{aligned}$$

The set $K_{k,t}$ and t -norms

- $K_{k,t} := \{\lambda \in \Delta_k \mid \forall x \in \Delta_k, \langle \lambda, x \rangle \leq \|x\|_{(t)}\}$.
- Recall that

$$\max_{x \in V, \|x\|=1} \lambda_1(x) = \max_{y \in \mathbb{C}^k, \|y\|=1} \|P_V P_y \otimes I_n P_V\|_\infty.$$

- For **fixed** y , P_V and $P_y \otimes I_n$ are independent projectors of relative ranks t and $1/k$ respectively.
- Thus,

$$\begin{aligned} \|P_V \cdot P_y \otimes I_n \cdot P_V\|_\infty &\rightarrow \|((1-t)\delta_0 + t\delta_1) \boxtimes ((1-1/k)\delta_0 + 1/k\delta_1)\| \\ &= \varphi(t, 1/k) = \|(1, 0, \dots, 0)\|_{(t)}. \end{aligned}$$

- We can take the max over y at no cost, by considering a **finite** net of y 's, since k is **fixed**.
- To get the full result $\limsup_{n \rightarrow \infty} K_{V_n} \subset K_{k,t}$, use $\langle \lambda, x \rangle$ (for all directions x) instead of λ_1 .
- The inclusion $\liminf_{n \rightarrow \infty} K_{V_n} \supset K_{k,t}$, is much easier, and follows from the convergence in distribution.

Additivity violations

$$H_{\min}^p(\Phi \otimes \bar{\Phi}) \leq B_2 < 2B_1 \leq 2H_{\min}^p(\Phi).$$

Theorem (Collins + N. '09)

For all k, t , almost surely as $n \rightarrow \infty$, if $Z_n = (\Phi \otimes \bar{\Phi})(E_{tnk})$

$$\text{spec}(Z_n) \rightarrow \left(t + \frac{1-t}{k^2}, \underbrace{\frac{1-t}{k^2}, \dots, \frac{1-t}{k^2}}_{k^2-1 \text{ times}} \right) \in \Delta_{k^2}.$$

Theorem (Belinschi, Collins, N. '13)

For all $p \geq 1$,

$$\lim_{n \rightarrow \infty} H_p^{\min}(\Phi) = H_p(a, b, b, \dots, b),$$

where $b = (1-a)/(k-1)$ and $a = \varphi(1/k, t)$ with

$$\varphi(s, t) = \begin{cases} s + t - 2st + 2\sqrt{st(1-s)(1-t)} & \text{if } s + t < 1; \\ 1 & \text{if } s + t \geq 1. \end{cases}$$

Theorem (Belinschi, Collins, N. '13)

Using the limit for $H^{\min}(\Phi)$ and the upper bound for $H^{\min}(\Phi)$, the lowest dimension for which a violation of the additivity can be observed is $k = 183$. For large k , violations of size $1 - \varepsilon$ bits can be obtained.

How to improve this ?

- 1 Other asymptotic regimes
- 2 Use $\Psi \neq \bar{\Phi}$
- 3 For $\Phi \otimes \bar{\Phi}$, compute the actual limit of $H^{\min}(\Phi \otimes \bar{\Phi})$, and not just an upper bound.

The End

thank you for your attention