

Random Subspaces of a Tensor Product and the Additivity Problem

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joint work with

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Eigen- and singular values

- ▶ **Singular value decomposition** of a matrix $X \in \mathcal{M}_{k \times n}(\mathbb{C})$

$$X = \sum \sqrt{\lambda_i(X)} e_i \otimes f_i^*,$$

where e_i, f_i are orthonormal families in $\mathbb{C}^k, \mathbb{C}^n$, and $\lambda_1(X) \geq \lambda_2(X) \geq \dots \geq 0$ are the singular values of X .

- ▶ The eigenvalues of the matrix XX^* are $\lambda_i(x)$.
- ▶ **Question:** What are the singular values of a **random** matrix ?

Singular values of random matrices, k fixed, $n \rightarrow \infty$ regime

- ▶ Let X be a $k \times n$ **Ginibre** random matrix, i.e. $\{X_{ij}\}$ are i.i.d. complex Gaussian random variables.
- ▶ We are interested in **long matrices**: k fixed, $n \rightarrow \infty$.
- ▶ We normalize our matrices, by taking them on the unit Euclidean sphere $\text{Tr}XX^* = 1$.
- ▶ Thus, the singular values vector $\lambda(X)$ is a probability vector

$$\lambda(X) \in \Delta_k^\downarrow = \{y \in \mathbb{R}^k : y_i \geq 0, \sum_i y_i = 1, y_1 \geq \dots \geq y_k\}.$$

- ▶ It is an easy exercise to show that, almost surely,

$$\forall i, \quad \lambda_i(X) \rightarrow 1/k.$$

Vector formulation

- ▶ Recall: SVD of $X \in \mathcal{M}_{k \times n}(\mathbb{C})$

$$X = \sum \sqrt{\lambda_i(X)} e_i \otimes f_i^*.$$

- ▶ Using the isomorphism $\mathbb{C}^k \otimes \mathbb{C}^n \simeq \mathcal{M}_{k \times n}(\mathbb{C})$, X can be seen as a vector in a tensor product $x \in \mathbb{C}^k \otimes \mathbb{C}^n$.
- ▶ The vector x admits a **Schmidt decomposition**
 $x = \sum_i \sqrt{\lambda_i(x)} e_i \otimes f_i$.
- ▶ The eigenvalues of the matrix $XX^* = [\text{id}_k \otimes \text{Tr}_n] P_x$ are $\lambda_i(x)$.
- ▶ **Problem:** What are the singular values of **ALL** vectors [matrices] inside a (random) subspace V of a tensor product [matrix space] ?
- ▶ This is a **simple question** about subspaces of tensor products (equivalently, about the singular values of matrices inside a given subspace V).

Singular values of vectors from a subspace

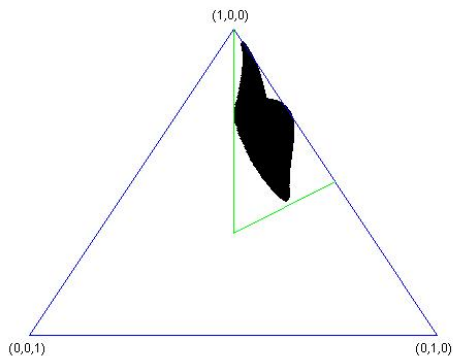
- ▶ For a subspace $V \subset \mathbb{C}^k \otimes \mathbb{C}^n$ of dimension d , define the set eigen-/singular values or Schmidt coefficients

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

- ▶ 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 $K_{V_1} \subset K_{V_2}$.
- ▶ Example: $d = 1$, $V = \mathbb{C}x$. We have $K_V = \{\lambda(x)\}$.
- ▶ Example: if $d > (k-1)(n-1)$, then $(1, 0, \dots, 0) \in K_V$.

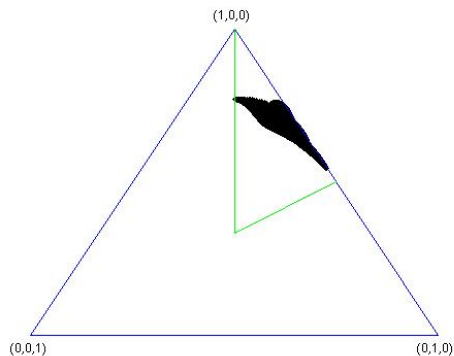
Examples

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



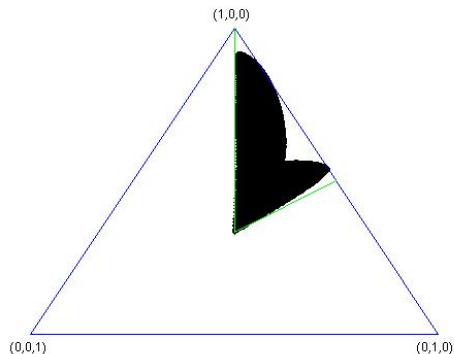
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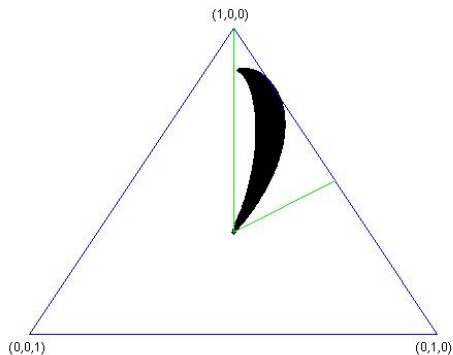
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Why do we care ? Quantum channels !

- ▶ **Quantum states** with d degrees of freedom are described by density matrices

$$X \in \mathcal{M}^{\text{sa}}(\mathbb{C}^d); \quad \text{Tr}X = 1 \text{ and } X \geq 0.$$

- ▶ **Quantum channels** $\mathcal{N} : \mathcal{M}(\mathbb{C}^d) \rightarrow \mathcal{M}(\mathbb{C}^k)$ are completely positive, trace-preserving maps. In particular, they send quantum states to quantum states.
 - ▶ Complete positivity **CP**: $\mathcal{N} \otimes \text{id}_n$ preserves positivity.
 - ▶ Trace preservation **TP**: $\text{Tr}[\mathcal{N}(X)] = \text{Tr}(X)$ for all X .
- ▶ Quantum channels describe the most general physical transformations a quantum system can undergo.

The additivity problem

- ▶ The **von Neumann entropy** of $X \in \mathcal{M}^{1,+}(\mathbb{C}^d)$

$$H(X) = -\text{Tr}(X \log X).$$

- ▶ The entropy is **additive**: $H(X_1 \otimes X_2) = H(X_1) + H(X_2)$.
- ▶ The **minimum output entropy** of a quantum channel is

$$H^{\min}(\mathcal{N}) = \min_{X \in \mathcal{M}^{1,+}(\mathbb{C}^d)} H(\mathcal{N}(X)).$$

Conjecture (Amosov, Holevo, and Werner '00)

For any channels $\mathcal{N}_1, \mathcal{N}_2$

$$H^{\min}(\mathcal{N}_1 \otimes \mathcal{N}_2) = H^{\min}(\mathcal{N}_1) + H^{\min}(\mathcal{N}_2).$$

- ▶ Given $\mathcal{N}_1, \mathcal{N}_2$, the \leq direction of the equality is trivial, take $X_{12} = X_1 \otimes X_2$.

An important problem

- ▶ Additivity of MOE is equivalent to other additivity conjectures in quantum information theory [Shor '03], such as the additivity of Holevo capacity for quantum channels or the additivity of entanglement of formation for quantum bipartite states.
- ▶ Additivity has been shown to hold for a large class of channels: unitary, unital qubit, depolarizing, dephasing, entanglement breaking, ...
- ▶ But ... **the Additivity Conjecture is false!**
- ▶ How to get counter-examples:
 1. Lower bound $H^{\min}(\mathcal{N}_{1,2})$;
 2. Upper bound $H^{\min}(\mathcal{N}_1 \otimes \mathcal{N}_2)$, eg. by finding a particular input X_{12} with low entropy;
 3. Conclude by
$$H^{\min}(\mathcal{N}_1 \otimes \mathcal{N}_2) \leq UB < LB_1 + LB_2 \leq H^{\min}(\mathcal{N}_1) + H^{\min}(\mathcal{N}_2).$$
- ▶ In this talk, we focus on lower bounding the MOE of a quantum channel.

Channels as subspaces

- ▶ Let $V \subset \mathbb{C}^k \otimes \mathbb{C}^n$ be a subspace of dimension d and consider an **isometry** $W : \mathbb{C}^d \rightarrow \mathbb{C}^k \otimes \mathbb{C}^n$ with $\text{Im}W = V$.
- ▶ One can define a channel $\mathcal{N} : \mathcal{M}(\mathbb{C}^d) \rightarrow \mathcal{M}(\mathbb{C}^k)$ by

$$\mathcal{N}(X) = [\text{id}_k \otimes \text{Tr}_n](WXW^*).$$

- ▶ Every channel can be defined in this way (by choosing n large enough).
- ▶ By convexity properties, the MOE is attained on **pure states** i.e. rank one projectors.
- ▶ Since $\mathcal{N}(P_x) = [\text{id}_k \otimes \text{Tr}_n](WP_x W^*) = [\text{id}_k \otimes \text{Tr}_n]P_{Wx}$, the minimal entropies of the channel \mathcal{N} are determined by the image subspace $V = \text{Im}W$.

Random subspaces

- ▶ **Idea**: when you do not know how to find a subspace having some nice properties, **pick one at random!**
- ▶ There is an **uniform** (or Haar) measure on the set of d -dimensional subspaces of \mathbb{C}^{kn} .
- ▶ Take a $kn \times kn$ Haar distributed random unitary matrix $U \in \mathcal{U}(kn)$ and take V to be the span of its first d columns.
- ▶ Alternatively, if W is a $kn \times d$ truncation of U , then $V = \text{Im}W$ is uniform.
- ▶ From such a random isometry W , one can construct **random quantum channels** $\mathcal{N}(X) = [\text{id}_k \otimes \text{Tr}_n](WXW^*)$.
- ▶ There are other measures on the Grassmannian one can consider, the one above being the simplest and the most natural.

Main result

- ▶ For the rest of the talk, we consider 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 MOE

- ▶ 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:

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

- ▶ It is not just a bound, the **exact limit value** is obtained 😊
- ▶ However, the set $K_{k,t}$ is not explicit, and minimizing entropy functions is difficult 😞

Idea of the proof

- ▶ **Question:** what is the maximum singular value $\max_{x \in V, \|x\|=1} \lambda_1(x)$ of a unit vector from V ?
- ▶ Compute

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

- ▶ Limit of $\|P_V \cdot P_y \otimes I_n\|_\infty$ for **fixed** y and **random** V ?

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 $q_n \sim \alpha n$ and $q'_n \sim \beta n$ where $\alpha, \beta \in (0, 1)$. Let P_n (resp. P'_n) be the orthogonal projection onto V_n (resp. V'_n). Then,

$$\lim_n \|P_n P'_n P_n\|_\infty = \varphi(\alpha, \beta).$$

- ▶ One can compute

$\varphi(\alpha, \beta) = \alpha + \beta - 2\alpha\beta + 2\sqrt{\alpha\beta(1-\alpha)(1-\beta)}$ if $\alpha + \beta \leq 1$
and $\varphi(\alpha, \beta) = 1$ if $\alpha + \beta > 1$ (subspaces V_n and V'_n have non-trivial intersection).

t -norms

Definition

For a positive integer k , embed \mathbb{R}^k as a self-adjoint real subalgebra \mathcal{R} of a II_1 factor (\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} .

Proposition

The distribution $\mu_{t^{-1}p_t x p_t}$ of the (non-commutative) random variable $t^{-1}p_t x p_t$ in the II_1 factor reduced by the projection p_t is $\mu_{t^{-1}p_t x p_t} = \mu_x^{\boxplus 1/t}$, $t \in (0, 1]$, where \boxplus denotes the **free additive convolution** of Voiculescu.

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

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

- ▶ For **fixed** y , P_V and $P_y \otimes I_n$ are independent projectors of relative ranks t and $1/k$ respectively.
- ▶ Thus, $\|P_V \cdot P_y \otimes I_n\|_\infty \rightarrow \varphi(t, 1/k) = \|(1, 0, \dots, 0)\|_{(t)}$.
- ▶ 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, use $\langle \lambda, x \rangle$ (for all directions x) instead of λ_1 .
- ▶ Unfortunately, it is difficult to compute (t) -norms, so we do not have an **explicit** formula for $K_{k,t}$.

Thank you !

Collins, N. - *Random quantum channels II: Entanglement of random subspaces, Rényi entropy estimates and additivity problems.*

Belinschi, Collins, N. - *Laws of large numbers for eigenvectors and eigenvalues associated to random subspaces in a tensor product.*