

Random positive maps

Benoît Collins

Kyoto University & University of Ottawa

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Overview

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Joint work in preparation with Patrick Hayden (Stanford) and Ion Nechita (CNRS, Toulouse)

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2. Random positive maps with random matrices (convergence of the largest eigenvalue).
3. Application: almost optimal entanglement witnesses.

Quantum information: a primer

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- ▶ *Entangled states* $Ent(n_1, n_2) := D(n_1 n_2) - Sep(n_1, n_2)$. A very important set (resource for quantum computing, etc).

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- ▶ Φ_1 is k -positive for all k whereas Φ_2 is 'only' 1-positive.
- ▶ A map that is positive for all k is called *completely positive*.

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(roughly speaking) The only final results available are: maps from $\mathcal{M}_{n_1}(\mathbb{C}) \rightarrow \mathcal{M}_{n_2}(\mathbb{C})$ with

$(n_1, n_2) = \{(1, n); (n, 1); (2, 2); (2, 3); (3, 2)\}$ are positive iff they are CP.

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So, trying to find *positive but not completely positive maps* is a strategy to *witness* entanglement.

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- ▶ The following state is not separable in $\mathcal{M}_2(\mathbb{C}) \otimes \mathcal{M}_2(\mathbb{C})$

$$\begin{pmatrix} 0.2 & 0 & 0 & 0 \\ 0 & 0.3 & 0.3 & 0 \\ 0 & 0.3 & 0.3 & 0 \\ 0 & 0 & 0 & 0.2 \end{pmatrix}$$

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We need to find more examples of 'more' positive maps. This is a hard task.
- ▶ RMT can help.

Choi matrix

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- ▶ More recently: Φ is *positive* iff $\rho \otimes 1_{n_2} C_\Phi \rho \otimes 1_{n_2}$ is positive.

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That is, the eigenvalues are located in a semi-circle distribution on the interval $[1 - 2\sqrt{a/k}, 1 + 2\sqrt{a/k}]$.

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with probability tending to 1 as n becomes large, we obtain a *random positive map*.
[largest eigenvalue convergence + ε -net + union bound argument]
- ▶ In addition, if $1 - 2\sqrt{a} < 0$, Φ is not completely positive with probability tending to 1 as n becomes large, therefore it 'detects' many entangled states.

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[uses C & Male's strong convergence for random unitaries]

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Let X be a *GUE* as above: $X \in \mathcal{M}_k(\mathbb{C}) \otimes \mathcal{M}_n(\mathbb{C})$ its eigenvalues are approximately in $[1 - 2\sqrt{a}, 1 + 2\sqrt{a}]$. k is fixed, n tends to ∞ .

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- ▶ X is positive with probability one as $n \rightarrow \infty$ as soon as $0 < \alpha < 1$.
- ▶ X is PPT with probability one as $n \rightarrow \infty$ as soon as $0 < \alpha < 1$.
- ▶ PPT states and general states have typical size – i.e. PPT is not so efficient in large dimension to detect entanglement.

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- ▶ If $\alpha < \sqrt{k}/(2(k-1) + \sqrt{k})$, X is separable.
- ▶ The criterion starts to become useful when $k > 16$.
- ▶ In both cases, α is of order C/\sqrt{k} . The order C/\sqrt{k} is optimal. We use the non-centered GUE random positive maps exhibited earlier to prove this result.
- ▶ Conclusion: Random maps are much more efficient than PPT.

Thank you!

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