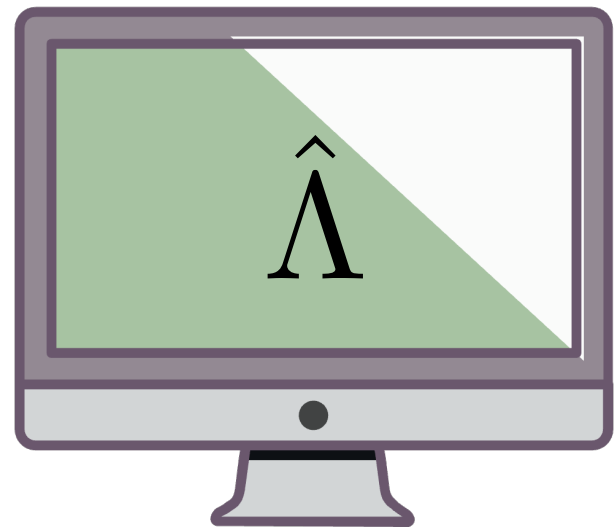


Optimal tomography of quantum channels



$$\|\hat{\Lambda} - \Lambda\|_{\diamond} \leq \varepsilon$$

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Quantum Physics

[Submitted on 11 Dec 2025 (v1), last revised 16 Jan 2026 (this version, v2)]

Optimal learning of quantum channels in diamond distance

Antonio Anna Mele, Lennart Bittel

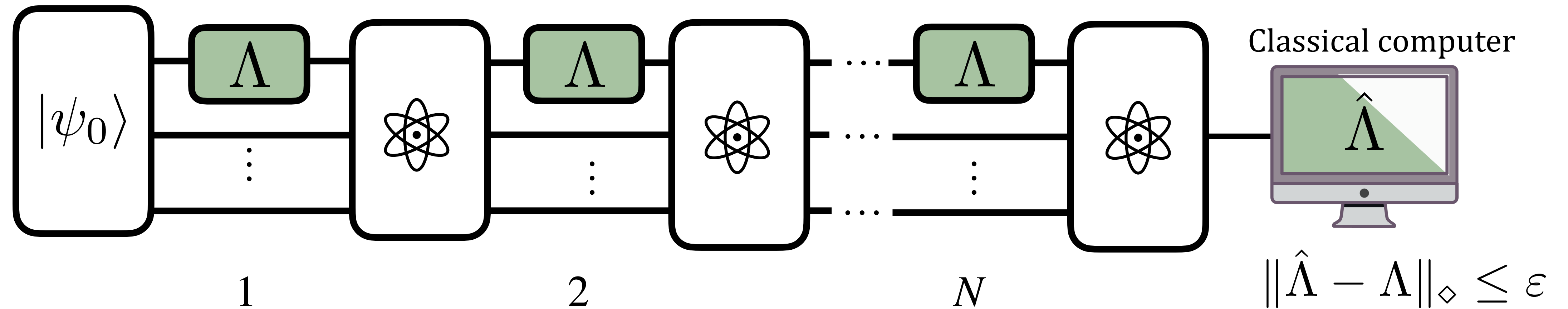
Joint work with
Lennart Bittel



Photo taken during a road trip in South Africa (Nov 2025), where we started thinking about the problem.

Quantum process tomography

Learn/estimate an unknown quantum channel Λ from black-box queries.



The **diamond distance** $\|\hat{\Lambda} - \Lambda\|_{\diamond}$ is the standard error metric for channels.

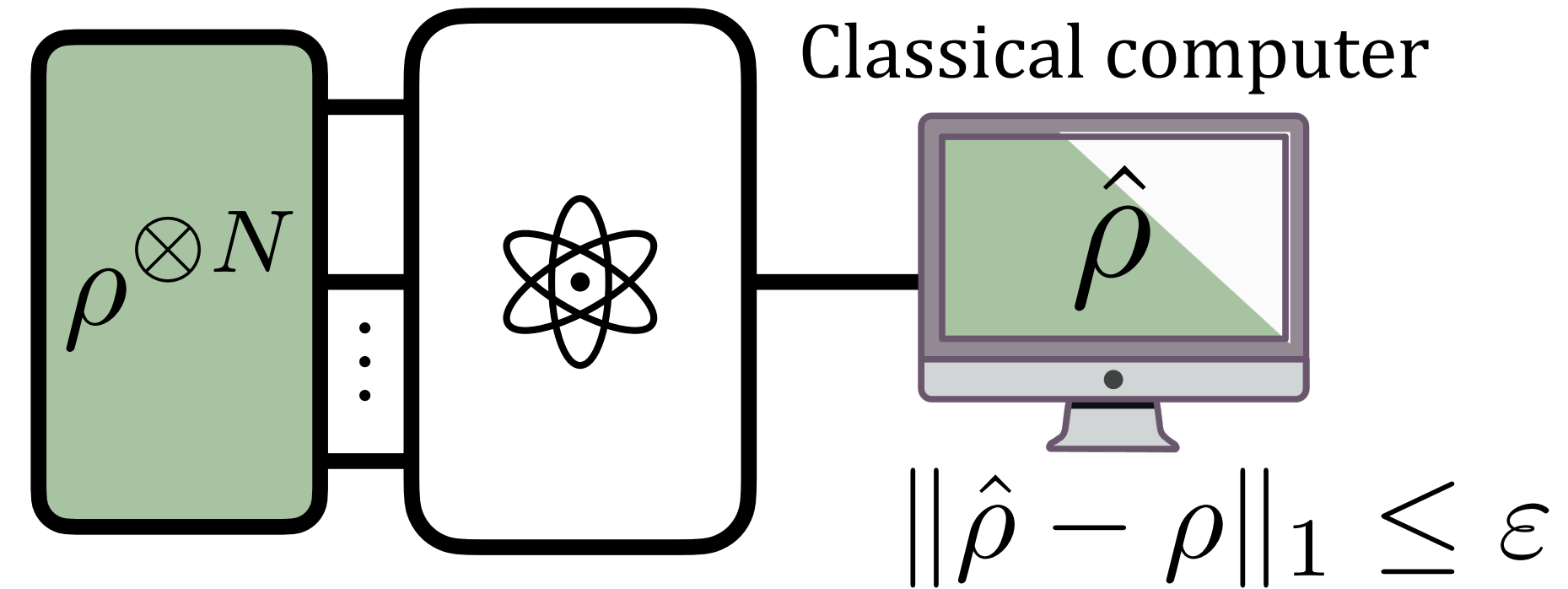
Open problem

Optimal *query complexity* for channel tomography.

How many uses of Λ are sufficient/necessary to estimate it within ϵ -accuracy in diamond distance?

Optimal number of queries for tomography

- This has been settled only for the special case of **quantum state tomography**.



- $N = \Theta(rd/\varepsilon^2)$ sufficient and necessary to learn an unknown r -rank state ρ .

Pure states ($r = 1$)	Mixed states ($r \leq d$)
[Hayashi (1998)] [Guta, Kahn, Kueng, Tropp (2018)]	[O'Donnell, Wright (2015)] [Haah, Harrow, Ji, Wu, Yu (2015)]

- For channels, this was solved only for unitaries (Kraus rank $r = 1$).

Unitaries ($r = 1$)	General channels ($r \leq d^2$)
[Haah, Kothari, O'Donnell, Tang (2022)] $N = O(d^2/\varepsilon)$??? [This work] $N = O(rd^2/\varepsilon^2)$

Quantum channel Λ : CPTP map

$$\Lambda : \mathcal{L}(\mathbb{C}^{d_{\text{in}}}) \rightarrow \mathcal{L}(\mathbb{C}^{d_{\text{out}}})$$

Kraus rank r : minimum number of Kraus operators $\{K_i\}_{i=1}^r \subset \mathbb{C}^{d_{\text{out}} \times d_{\text{in}}}$.

$$\Lambda(\rho) = \sum_{i=1}^r K_i \rho K_i^\dagger, \quad \sum_{i=1}^r K_i^\dagger K_i = I_{d_{\text{in}}}. \quad \lceil d_{\text{in}}/d_{\text{out}} \rceil \leq r \leq d_{\text{in}}d_{\text{out}}$$

Diamond distance:

$$\begin{aligned} \|\hat{\Lambda} - \Lambda\|_\diamond &= \sup_{\rho \in \mathcal{D}(\mathbb{C}^{d_{\text{in}}} \otimes \mathbb{C}^{d_{\text{in}}})} \left\| ((\hat{\Lambda} - \Lambda) \otimes \text{id})(\rho) \right\|_1 \\ &= 4 \cdot \max_{\text{protocol}} \left(\Pr[\text{correctly distinguish } \hat{\Lambda} \text{ vs. } \Lambda] - \frac{1}{2} \right) \end{aligned}$$

- **Previous channel-tomography query-complexity bounds were not tight.**

(e.g. $d_{\text{in}} = d_{\text{out}} = d$, constant ε):

Upper bound: $N = O(d^6)$, [Surawy-Stepney, Kahn, Kueng, Guta, 2021]
[Oufkir, 2023]

↓ quadratic gap ↓

Lower bound: $N = \tilde{\Omega}(d^4)$. [Rosenthal, Aaronson, Subramanian, Datta, Gur, 2024]

- A Kraus rank- r channel can be described by $\Theta(r d_{\text{in}} d_{\text{out}})$ parameters.

- **Natural conjecture** (parameter counting):

$$N \stackrel{?}{=} \Theta(r d_{\text{in}} d_{\text{out}}) = O(r d^2) = O(d^4)$$

We prove this and so close the gap.

Main Theorem (informal)

Let $\Lambda : \mathcal{L}(\mathbb{C}^{d_{\text{in}}}) \rightarrow \mathcal{L}(\mathbb{C}^{d_{\text{out}}})$ be an unknown channel of Kraus rank $\leq r$.

There exists a process-tomography algorithm that uses

$$N = O\left(\frac{d_{\text{in}} d_{\text{out}} r}{\varepsilon^2}\right)$$

queries to Λ , and outputs $\hat{\Lambda}$ satisfying

$$\|\hat{\Lambda} - \Lambda\|_{\diamond} \leq \varepsilon$$

- For any fixed ε constant, $N = \Omega(d_{\text{in}} d_{\text{out}} r)$ queries are necessary.

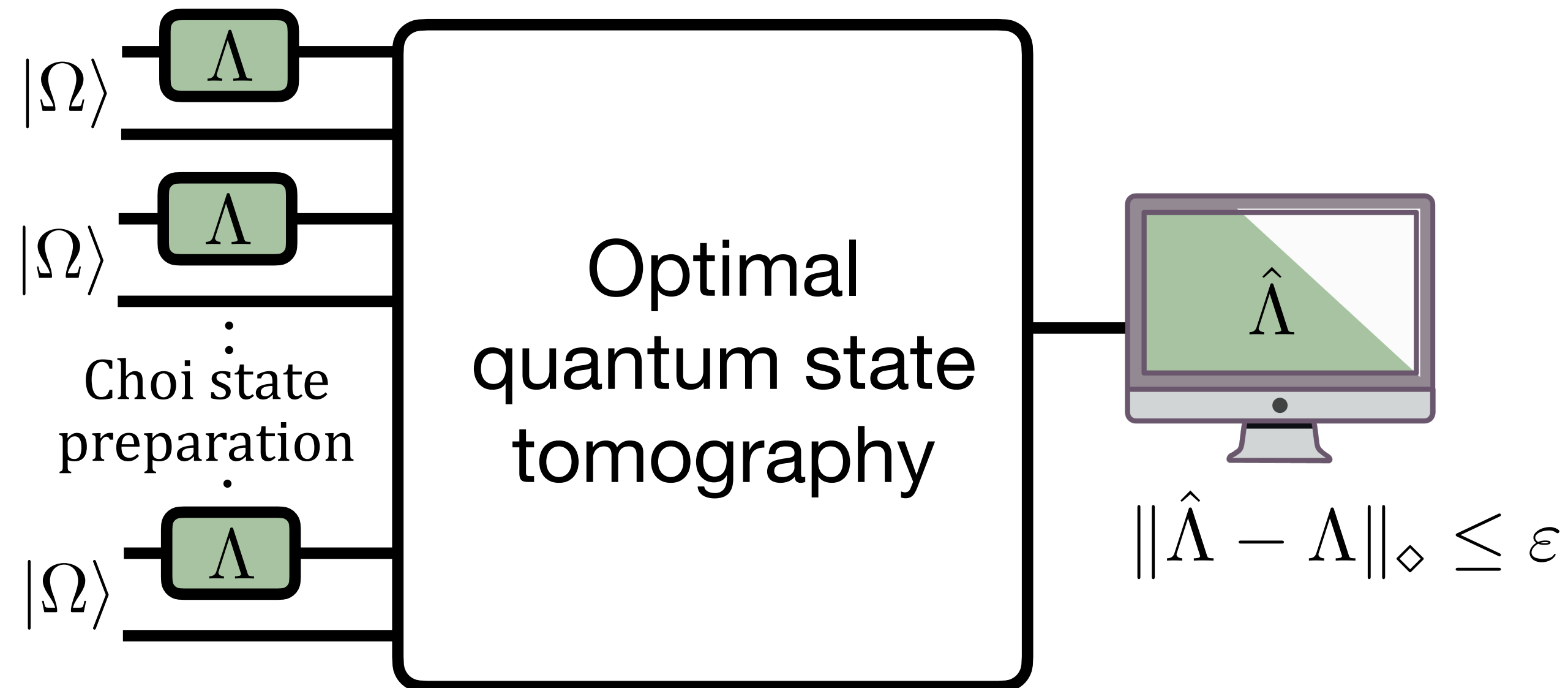
Our idea for the optimal algorithm

What is the first guess you'd make as a student learning basics of quantum channels?

(Assuming the existence of an optimal state tomography algorithm)

(HINT: Choi isomorphism)

Let's run **optimal** state tomography on the Choi state.

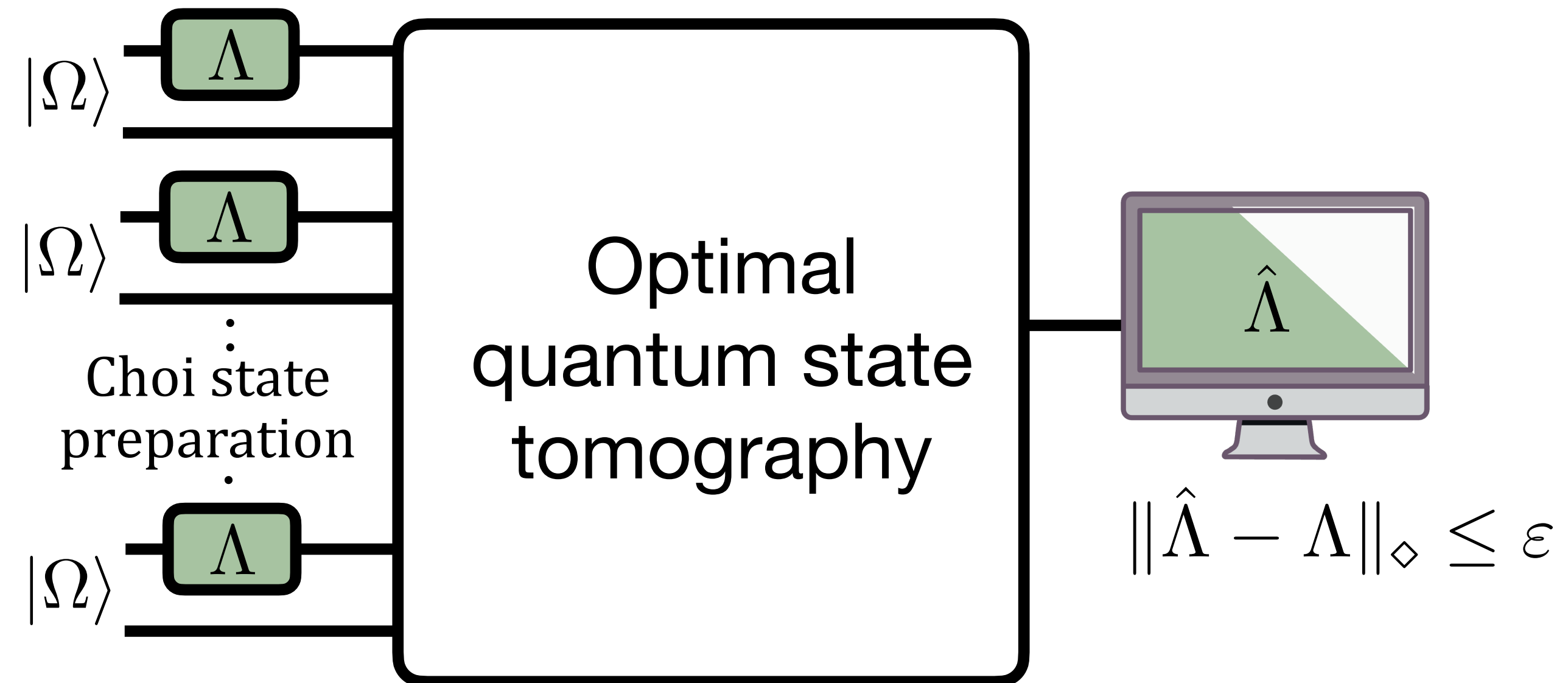


Spoiler, it suffices:

Optimal channel tomography = Optimal state tomography on Choi-state

Take-home message

Choi-state tomography suffices.



Common objection: doesn't this incur a dimensional overhead?

Answer: No (if you analyze the 'error state' directly in diamond norm)

The optimal state-tomography algorithm we use

Mixed state tomography reduces to pure state tomography

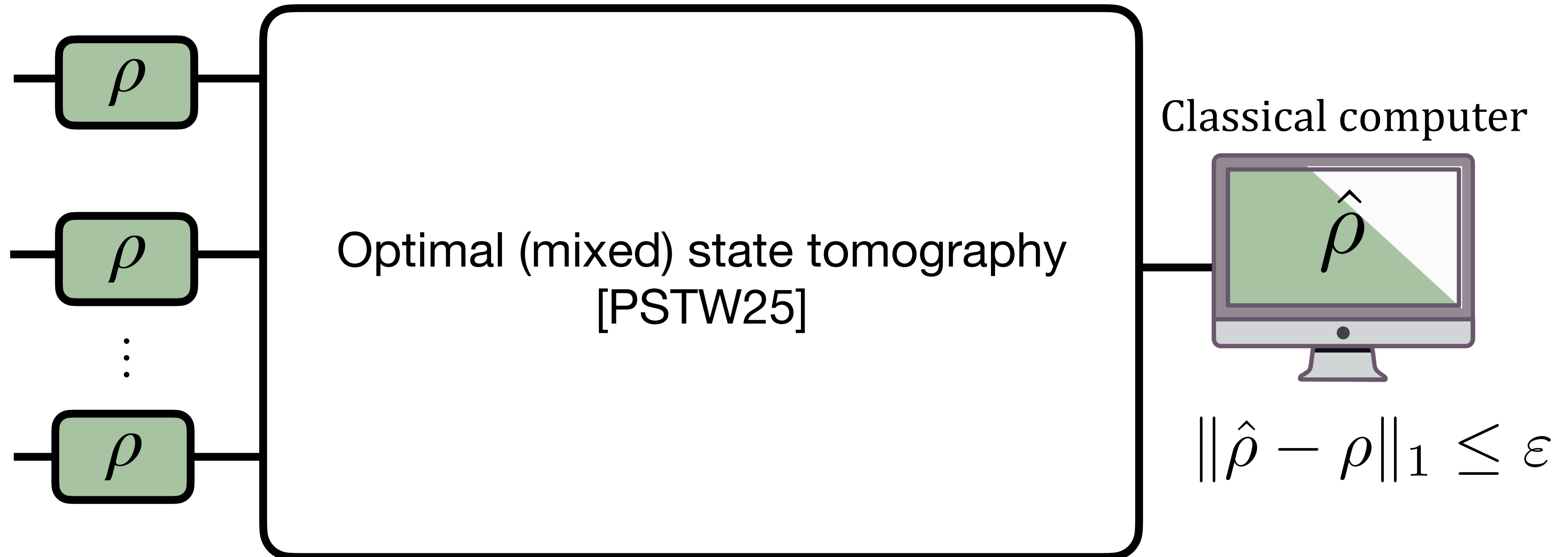
Angelos Pelecanos*

Jack Spilecki*

Ewin Tang*

John Wright*

arXiv:2511.15806



The optimal state-tomography algorithm we use

Mixed state tomography reduces to pure state tomography

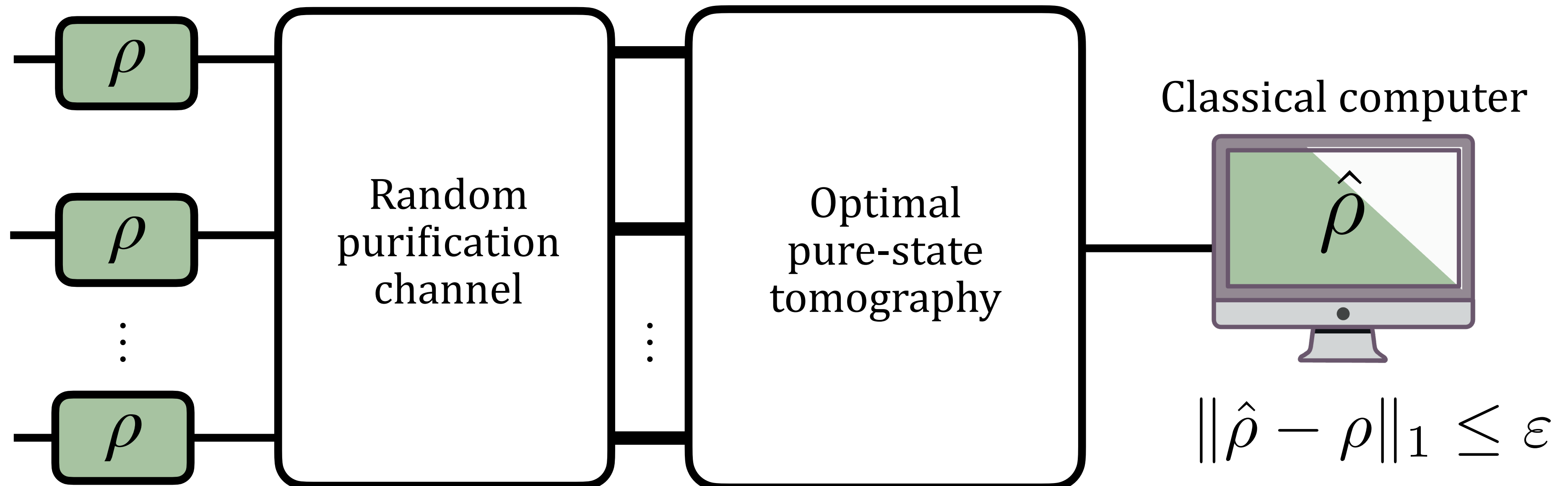
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The optimal state-tomography algorithm we use

Mixed state tomography reduces to pure state tomography

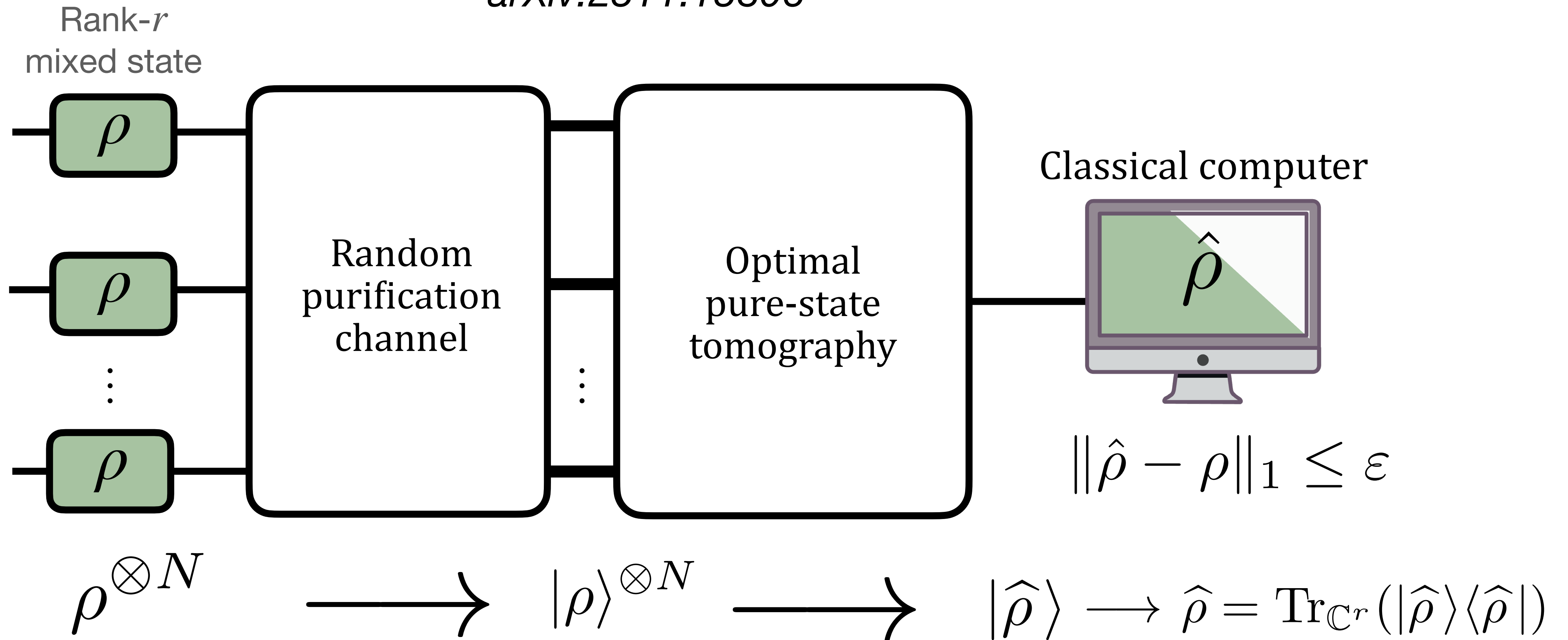
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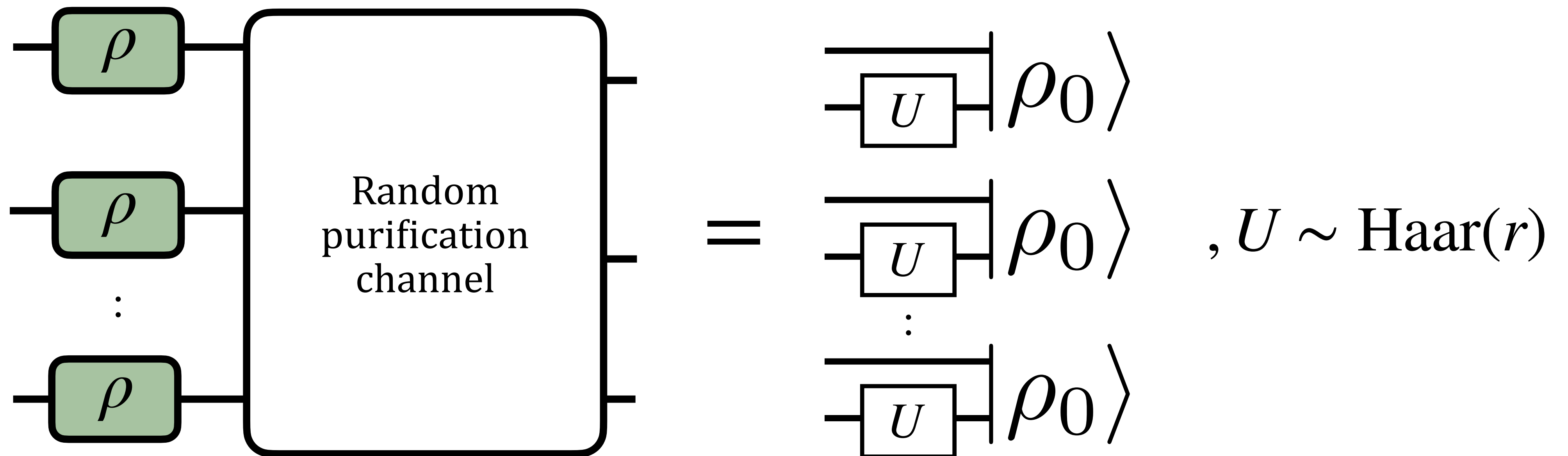
arXiv:2511.15806



Random purification channel

[Tang, Wright, Zhandry (2025)]

$$\Phi_{\text{Purify}}^{d,r}(\rho^{\otimes n}) = \mathbb{E}_{|\rho\rangle} [|\rho\rangle\langle\rho|^{\otimes n}] \quad |\rho\rangle \in \mathbb{C}^d \otimes \mathbb{C}^r$$



Proof of correctness

(mixed state tomography)

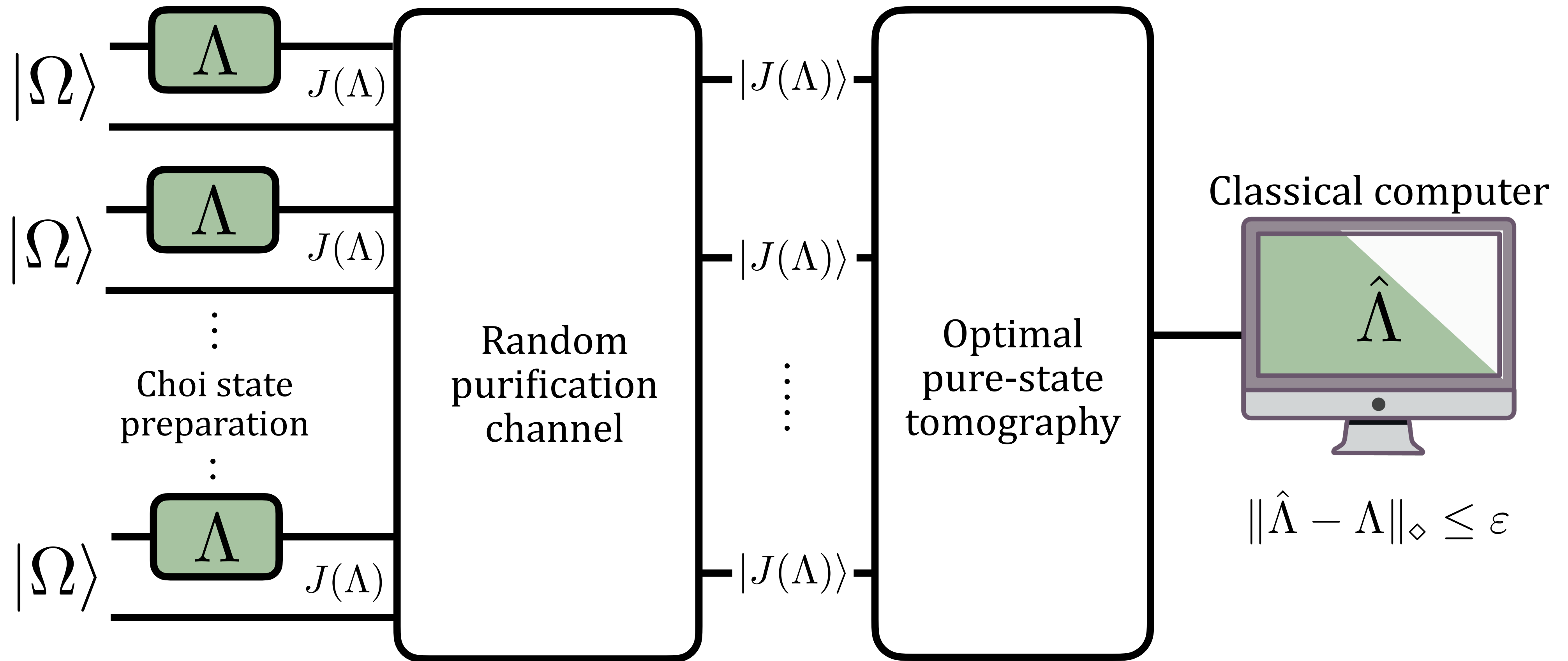
$$\begin{array}{l} \rho \text{ rank-}r \\ \text{mixed state} \end{array} \longrightarrow |\rho\rangle \in \mathbb{C}^d \otimes \mathbb{C}^r \longrightarrow N = O\left(\frac{rd}{\varepsilon^2}\right)$$

(Optimal pure-state
tomography guarantees) ↓

$$\|\hat{\rho} - \rho\|_1 = \left\| \text{Tr}_{\mathbb{C}^r}(|\hat{\rho}\rangle\langle\hat{\rho}| - |\rho\rangle\langle\rho|) \right\|_1 \leq \left\| |\hat{\rho}\rangle\langle\hat{\rho}| - |\rho\rangle\langle\rho| \right\|_1 \leq \varepsilon.$$

$$N = O\left(\frac{rd}{\varepsilon^2}\right) \implies \|\hat{\rho} - \rho\|_1 \leq \varepsilon.$$

Our channel-tomography algorithm



Channel tomography reduces to pure state tomography!

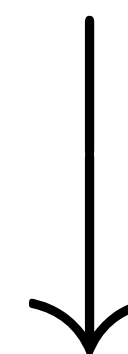
Proof of correctness

(Not as easy as for mixed state tomography)

$$\begin{aligned} \Lambda \text{ Kraus rank-}r &\longrightarrow \text{Choi state } J(\Lambda) \text{ rank-}r &\longrightarrow |J(\Lambda)\rangle \in \mathbb{C}^{d_{\text{in}}d_{\text{out}}} \otimes \mathbb{C}^r \\ \Lambda : \mathcal{L}(\mathbb{C}^{d_{\text{in}}}) \rightarrow \mathcal{L}(\mathbb{C}^{d_{\text{out}}}) & & J(\Lambda) \in \mathbb{C}^{d_{\text{in}}d_{\text{out}} \times d_{\text{in}}d_{\text{out}}} & \longrightarrow N = O\left(\frac{r d_{\text{in}}d_{\text{out}}}{\varepsilon^2}\right). \end{aligned}$$

Naive approach fails for channels:

(Optimal pure-state tomography guarantees)



$$\|\hat{J} - J(\Lambda)\|_1 = \left\| \text{Tr}_r \left(|\hat{J}\rangle\langle\hat{J}| - |J(\Lambda)\rangle\langle J(\Lambda)| \right) \right\|_1 \leq \left\| |\hat{J}\rangle\langle\hat{J}| - |J(\Lambda)\rangle\langle J(\Lambda)| \right\|_1 \leq \varepsilon$$

$$\|\hat{\Lambda} - \Lambda\|_{\diamond} \leq d_{\text{in}} \|\hat{J} - J(\Lambda)\|_1 \leq d_{\text{in}} \varepsilon$$

The naive reduction introduces a d_{in} overhead.

Proof of correctness

(Done well: via SDP equivalent definition of the diamond norm)

For Hermitian Choi matrices:

$$\begin{aligned} \|\Phi\|_{\diamond} &= d_{\text{in}} \max_{\substack{Y=Y^{\dagger} \\ \sigma \in \mathcal{D}(\mathcal{H}_{\text{in}})}} \text{tr}(J(\Phi) Y) \\ &\text{subject to } Y \in \mathcal{L}(\mathcal{H}_{\text{out}} \otimes \mathcal{H}_{\text{in}}), \\ &\quad -\mathbb{I}_{\text{out}} \otimes \sigma \leq Y \leq \mathbb{I}_{\text{out}} \otimes \sigma. \end{aligned}$$

For positive Choi matrices:

$$\begin{aligned} \|\Phi\|_{\diamond} &= d_{\text{in}} \max_{\sigma \in \mathcal{D}(\mathcal{H}_{\text{in}})} \text{tr}(J(\Phi) (\mathbb{I}_{\text{out}} \otimes \sigma)) \\ &= d_{\text{in}} \left\| \text{tr}_{\text{out}} J(\Phi) \right\|_{\infty}. \end{aligned}$$

Proof of correctness (Why no dimensional overhead)

Key decomposition for pure-state tomography:

[Hayashi (1998)]
[Guta, Kahn, Kueng, Tropp, (2018)]

$$|\hat{J}\rangle = \sqrt{1 - \varepsilon^2} |J(\Lambda)\rangle + \varepsilon |\psi_{\text{err}}\rangle, \quad |\psi_{\text{err}}\rangle \sim \text{Haar}.$$

Convert to Choi operators:

$$\hat{J} = \text{Tr}_{\mathbb{C}^r} (|\hat{J}\rangle\langle\hat{J}|), \quad J(\Lambda) = \text{Tr}_{\mathbb{C}^r} (|J(\Lambda)\rangle\langle J(\Lambda)|).$$

Choi difference:

$$\hat{J} - J(\Lambda) = \varepsilon \sqrt{1 - \varepsilon^2} (K_{12} + K_{21}) + \varepsilon^2 (K_{22} - K_{11}),$$

$$K_{ij} := \text{Tr}_{\mathbb{C}^r} (|\Phi_i\rangle\langle\Phi_j|), \quad \Phi_1 := J(\Lambda), \quad \Phi_2 := \psi_{\text{err}}.$$

Control the diamond norm of the Choi difference:

$$\|\hat{\Lambda} - \Lambda\|_{\diamond} \equiv \|\hat{J} - J(\Lambda)\|_{\diamond} \leq O(\varepsilon) \left(\underbrace{\|K_{11}\|_{\diamond}}_{= 1} + \underbrace{\|K_{12} + K_{21}\|_{\diamond}}_{\leq 2\sqrt{\|K_{22}\|_{\diamond}} = O(1)} + \underbrace{\|K_{22}\|_{\diamond}}_{= O(1)} \right) \leq O(\varepsilon)$$

Diamond SDP definition + Haar concentration:

$$\left(\begin{array}{l} \|K_{22}\|_{\diamond} \stackrel{\downarrow}{=} d_{\text{in}} \|\text{Tr}_{\text{out}}(K_{22})\|_{\infty} \stackrel{\downarrow}{=} O(1), \\ \|\text{Tr}_{\text{out},r}(|\psi_{\text{err}}\rangle\langle\psi_{\text{err}}|)\|_{\infty} \stackrel{\uparrow}{=} O(1/d_{\text{in}}). \end{array} \right)$$

$$\|\hat{\Lambda} - \Lambda\|_{\diamond} \leq O(\varepsilon)$$

Theorem (upper bound)

Let $\Lambda : \mathcal{L}(\mathbb{C}^{d_{\text{in}}}) \rightarrow \mathcal{L}(\mathbb{C}^{d_{\text{out}}})$ be an unknown channel of Kraus rank $\leq r$.

There exists a channel-tomography protocol that uses

$$N = O\left(\frac{d_{\text{out}} d_{\text{in}} r}{\varepsilon^2} + \frac{d_{\text{in}}}{\varepsilon^2} \log \frac{1}{\delta}\right)$$

queries to Λ , and outputs $\hat{\Lambda}$ satisfying $\|\hat{\Lambda} - \Lambda\|_{\diamond} \leq \varepsilon$ with probability at least $1 - \delta$.

Any channel tomography algorithm must use:

(with Kraus rank r larger than minimal)

$$N = \Omega\left(\frac{d_{\text{out}} d_{\text{in}} r}{\varepsilon^{\beta}} + \frac{d_{\text{in}}}{\varepsilon^2} \log \frac{1}{\delta}\right)$$

↑
Reduction to
channel identification
+ packing-net argument

↑
Reduction to
learning the bias of
many classical coins

The unification of optimal tomography protocols

$$N = O\left(\frac{d_{\text{in}} d_{\text{out}} r}{\varepsilon^2}\right) \implies \|\hat{\Lambda} - \Lambda\|_{\diamond} \leq \varepsilon$$

Optimal state tomography ($d_{\text{in}} = 1$, $\|\cdot\|_{\diamond} = \|\cdot\|_1$)

Optimal tomography of isometries (previously not known) and unitaries ($r = 1$)

Choi-state already pure (purification step not needed).

Thus, our protocol = single-copy Choi-state tomography. [Guta, Kahn, Kueng, Tropp (2018)]

Heisenberg scaling ε^{-1} achievable for unitaries as well. [Haah, Kothari, O'Donnell, Tang (2022)]

Optimal binary POVMs tomography (previously not known) ($d_{\text{out}} = 2$, $d_{\text{in}} \equiv d$, $r = d$)
(i.e., channels diagonal in the computational basis)

Concurrent work:

[Submitted on 15 Dec 2025 (v1), last revised 2 Feb 2026 (this version, v2)]

Quantum channel tomography and estimation by local test

Kean Chen, Nengkun Yu, Zhicheng Zhang

Follow-up:

[Submitted on 15 Jan 2026]

Optimal lower bound for quantum channel tomography in away-from-boundary regime

Kean Chen, Zhicheng Zhang, Nengkun Yu

Minimal Kraus rank:
 $r_{\min} = \lceil d_{\text{in}}/d_{\text{out}} \rceil$

$$r \geq 2r_{\min} \implies N = \Omega\left(\frac{d_{\text{out}}d_{\text{in}}^r}{\varepsilon^2}\right) \xrightarrow{\text{(together with our result)}} N = \Theta\left(\frac{d_{\text{out}}d_{\text{in}}^r}{\varepsilon^2}\right)$$

- In the relevant case $d_{\text{in}} = d_{\text{out}} = d$, this implies:

$r \geq 2$:	$r = 1$:
$N = \Theta\left(\frac{d^2 r}{\varepsilon^2}\right)$ <p>(Classical scaling)</p>	$N = \Theta\left(\frac{d^2}{\varepsilon}\right)$ <p>(Heisenberg scaling) [Haah, Kothari, O'Donnell, Tang (2022)]</p>

Conclusions

- **Channel tomography** with query-**optimal** dimensional-scaling.
- **Choi-state learning suffices.**
- **Adaptivity does not help** for optimal channel tomography.
(We use the channel only to prepare Choi state copies)
- **Entangled/Coherent measurements here help** (coherence is needed only to purify).
(Single-copy non-adaptive protocols are suboptimal [Oufkir, 2023])
(Entangled measurements provably necessary for non-adaptive protocols)

Open problems

- **Adaptivity helps for single-copy channel tomography?** (For states: [Chen, Huang, Li, Liu, Sellke, 2023])
- **Close the gap in the ε -dependence** left open near the minimal Kraus-regime. **THANK YOU!**

$$r_{\min} \leq r \leq 2r_{\min} \implies N = \Theta \left(\frac{d_{\text{out}} d_{\text{in}}^r}{\varepsilon^\beta} \right), \quad \beta = ?$$

- **Is adaptivity necessary for optimal unitary tomography** with Heisenberg scaling ε^{-1} ?
- **Limited entangled measurements trade-off.** (For states: [Chen, Li, Liu, 2025])
- Generalise to arbitrary **process tensors** (combs): **“A complete and unified theory of tomography”**.