

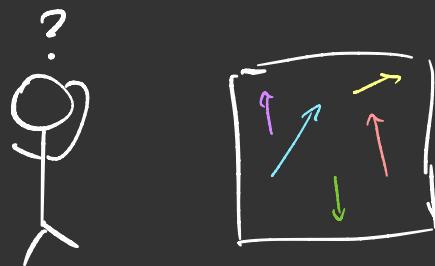
Fundamental Bounds on Quantum State Tomography

Angus Lowe

In the beginning ...

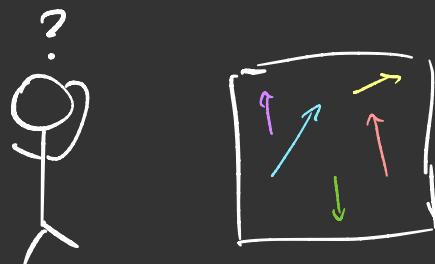
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$$\underbrace{, , \dots}_{\times n} \in (V^{\otimes m})^{\otimes n}$$



Coin Flipping



m correlated coins

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$$P = \epsilon V^{\otimes m}, \rightarrow \mathbb{R}^2$$

m correlated coins

$$= (P_{HHH\dots H}, P_{THH\dots H}, \dots, P_{TTT\dots T})$$
$$\|P\|_1 = \sum_j |P_j| = 1.$$

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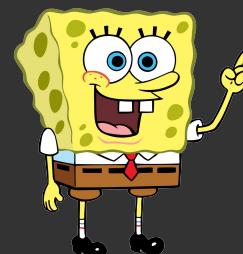


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Prepare P w/prob. $1/2$

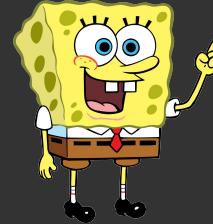
Guess P or q

Prepare q w/prob. $1/2$

Coin Flipping



[?]



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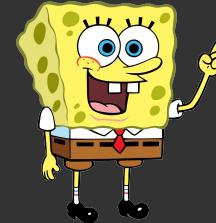
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Fact (Bayes' rule): Optimal prob. of success is

$$p_{\text{succ}} = \frac{1}{2} + \frac{1}{4} \|p - q\|_1$$

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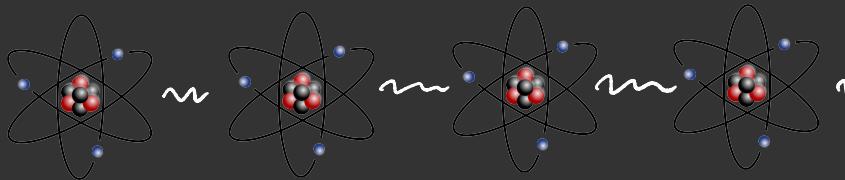
$$p_{\text{succ}} = \frac{1}{2} + \frac{1}{4} \|p - q\|_1$$

Fact: if $\frac{1}{2}\|p - q\|_1 \geq c$, then

$$\frac{1}{2}\|p^{\otimes n} - q^{\otimes n}\|_1 \geq 1 - \left(1 - \frac{c^2}{2}\right)^n$$

$$\Rightarrow p_{\text{succ}} \xrightarrow{n \rightarrow \infty} 1$$

Quantum State Distinguishability

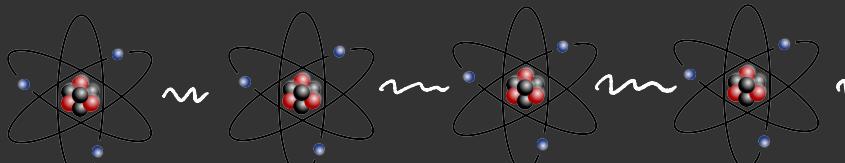
 , $\rho \in V^{\otimes m} \hookrightarrow \mathbb{C}^{2 \times 2}$

m entangled qubits

$$\sigma \in V^{\otimes m}$$

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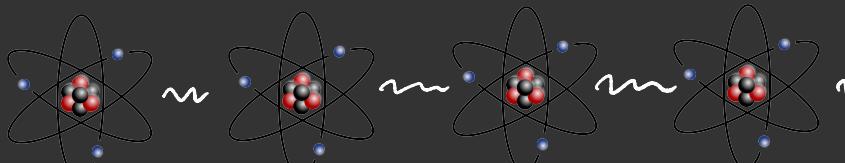
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Meas. w/ $\{E_0, E_1\}$

Prepare σ w/prob. 1/2

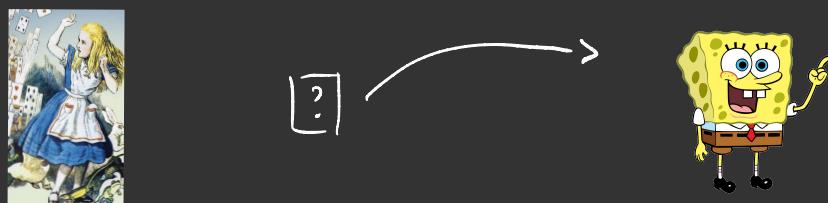
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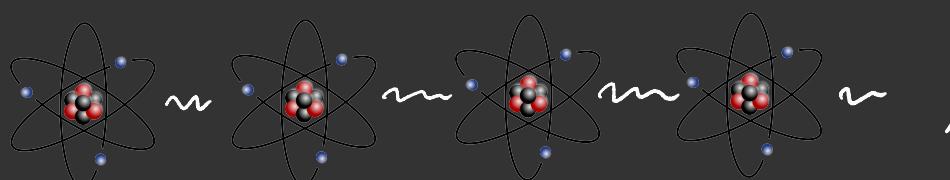
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Thm (Holevo-Helstrom): the optimal meas. $\{E_0, E_1\}$
 succeeds w/ prob.

$$P_{\text{succ}} = \text{Tr}(E_0 \rho) + \text{Tr}(E_1 \sigma) = \frac{1}{2} + \frac{1}{4} \|\rho - \sigma\|_1$$

Quantum State Distinguishability



m entangled qubits

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$$\|\rho\|_1 = \sum_i |\lambda_i| = 1.$$

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Say, $\rho = |\chi\rangle\langle\chi|$

$$\sigma = |\phi\rangle\langle\phi|,$$

then $\frac{1}{2} \|\rho - \sigma\|_1 = \sqrt{1 - |\langle\phi|\chi\rangle|}$

$$\frac{1}{2} \|\rho^{\otimes n} - \sigma^{\otimes n}\|_1 = \sqrt{1 - |\langle\phi|\chi\rangle|^n}$$

$$\xrightarrow{n \rightarrow \infty} 1$$

$$\Rightarrow P_{\text{succ}} \xrightarrow{n \rightarrow \infty} 1.$$

Distribution Learning

If we can distinguish, there is hope to learn.

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Distribution Learning Problem

Given: $(X_1, \dots, X_n) \sim p^{\otimes n}$.

Output: \hat{p}_n s.t. $\|\hat{p}_n - p\|_1 \leq \epsilon$.

Distribution Learning

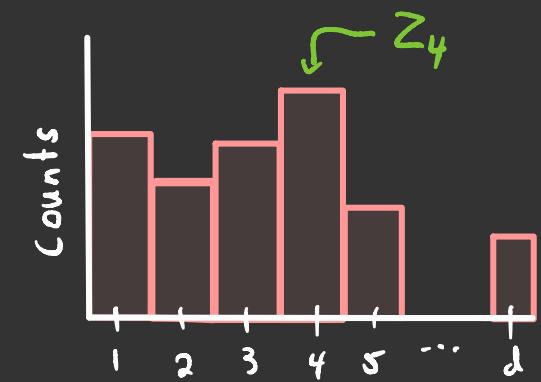
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Solⁿ:

Let $\hat{p}_n(z) = \frac{z}{n}$, $\bar{Z} = (Z_1, \dots, Z_d) \sim \text{Multi}(n, d)$

\rightarrow can show $\mathbb{E} \|\hat{p}_n - p\|_1 \leq \sqrt{d/n}$, so $n \sim d/\epsilon^2$ suffices.

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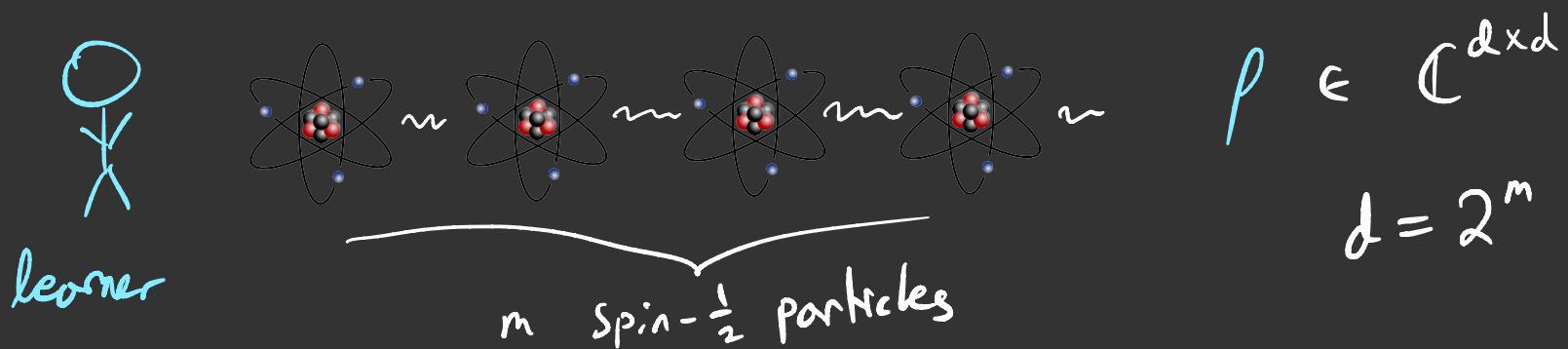
Note: CLT not good enough!

$$\left(\frac{Z_i - np_i}{\sqrt{n}} \right)_{i=1}^d \xrightarrow{n \rightarrow \infty} N(0, \Sigma)$$

$$\Leftrightarrow \forall \epsilon > 0 \exists n_0 \text{ s.t. } |\mathbb{E} Z_i - np_i| \leq \epsilon \sqrt{n} \quad \forall n \geq n_0.$$

What if $n_0 \sim d^2, d^3, e^d$?

Quantum State Tomography

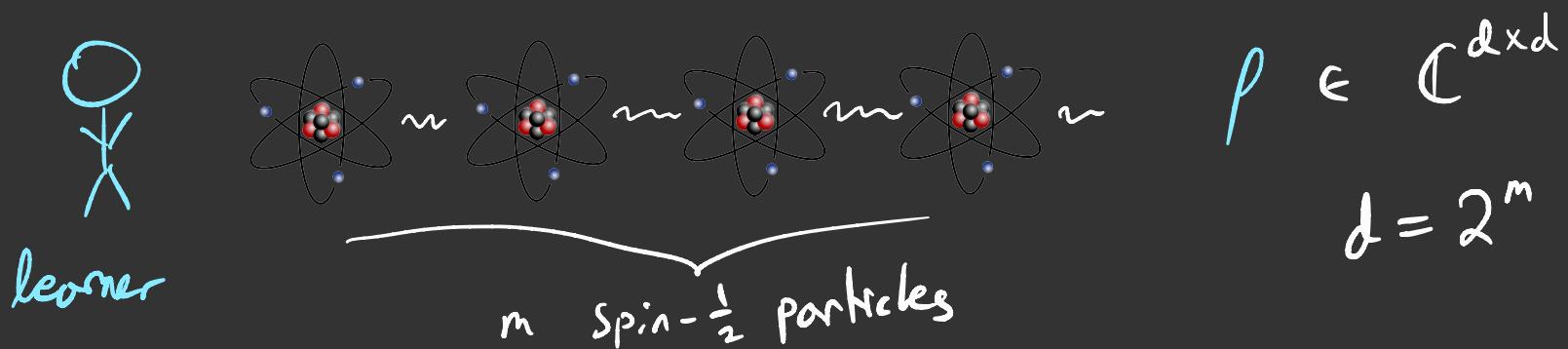


Quantum Tomography Problem

Given: $\rho^{\otimes n}$

Output: $\hat{\rho}_n$ s.t. $\|\hat{\rho}_n - \rho\|_1 \leq \epsilon$

Quantum State Tomography

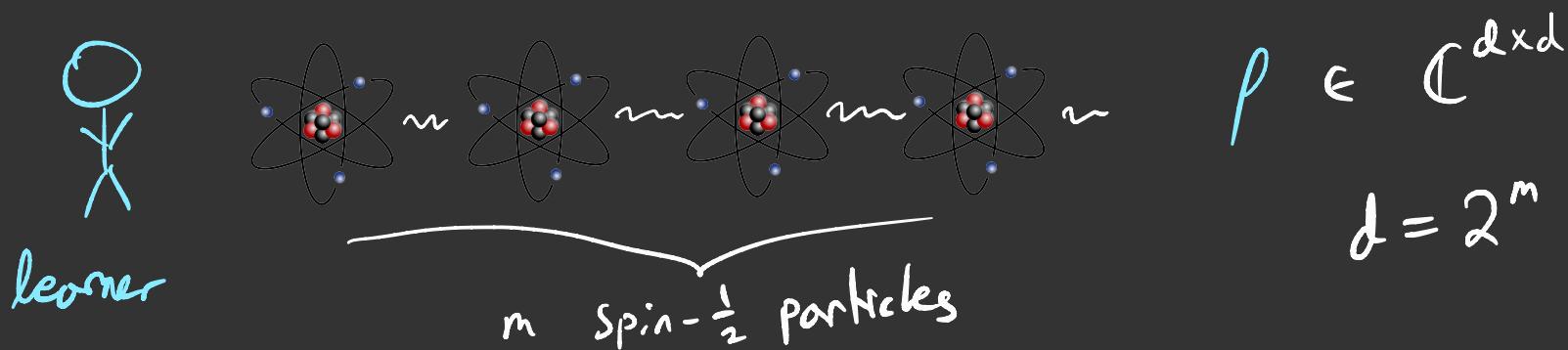


Quantum Tomography Problem

Given: $\rho^{\otimes n} \rightarrow$ measurement?

Output: $\hat{\rho}_n$ s.t. $\|\hat{\rho}_n - \rho\|_1 \leq \epsilon$

Quantum State Tomography



Quantum Tomography Problem

Given: $\rho^{\otimes n} \rightarrow$ measurement?

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- Unlike the classical case, we get different answers depending on the allowed measurements.

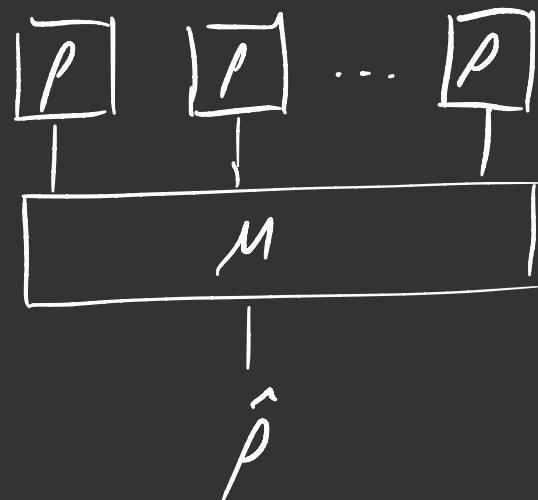
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Entangled	$\sim d^2/\epsilon^2$	c.f. classical distr. learning
Nonadaptive	$\sim d^3/\epsilon^2$	
Adaptive	$\sim d^3/\epsilon^2$	
Binary	$\sim d^4/\epsilon^2$	



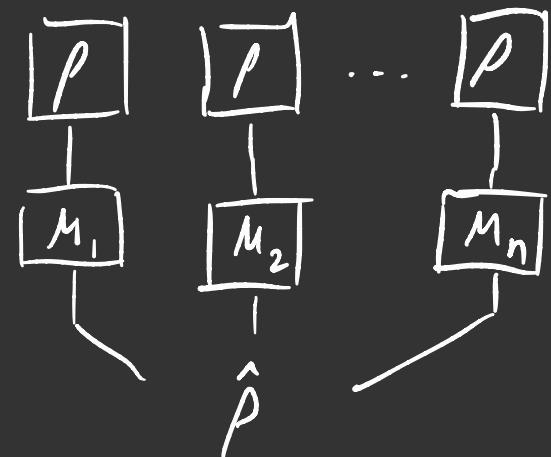
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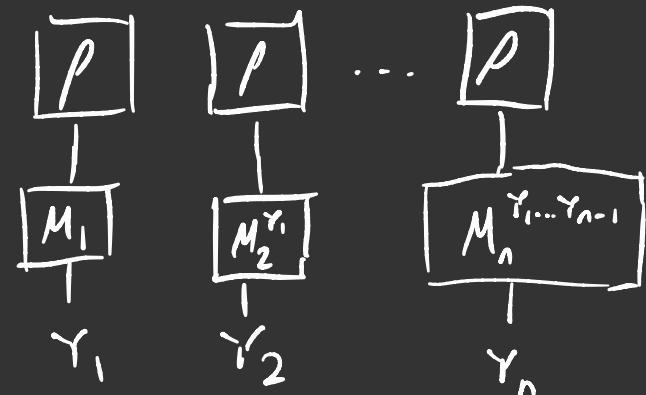
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	$\sim d^2/\epsilon^2$	[HHJWY '17, OW '17]
Entangled	$\sim d^2/\epsilon^2$	c.f. classical distr. learning
Nonadaptive	$\sim d^3/\epsilon^2$	\rightarrow [KRT '15, HHJWY '17]
Adaptive	$\sim d^3/\epsilon^2$	\rightarrow [LN '22, CHLLS '23]
Binary	$\sim d^4/\epsilon^2$	\rightarrow [LN '22] $\gamma \in \{0, 1\}$



...

Lower Bound

(*)

Want to show :

$$n \ll d^2/\epsilon^2 \text{ impossible for quantum tomography}$$

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Not allowed! \times



$$X \in \{1, \dots, 2^n + 1\}$$

$$\rho \in (\mathbb{C}^{2 \times 2})^{\otimes N}$$



Measure ρ , outcome Y
Decode $\hat{X}(Y) = X$.

Lower Bound

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Want to show : $n \ll d^2/\epsilon^2$ impossible for quantum tomography

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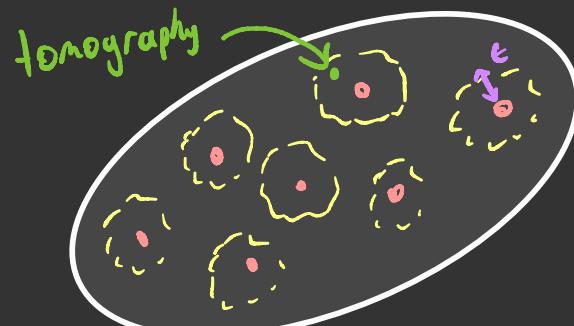


Measure ρ , outcome Y
Decode $\hat{X}(Y) = X$.

Proof of (*): $\rho^{\otimes n} \in (\mathbb{C}^{d \times d})^{\otimes n}$, #qubits \equiv log dim. $= n \log d$.



Contain $\exp(d^2)$ states, d^2 bits



Upper Bound

Recall state discrimination



Prepare ρ w/prob. $1/2$

Meas. w/ $\{E_0, E_1\}$

Prepare σ w/prob. $1/2$

Guess ρ or σ



Prepare ρ_j w/prob. $1/M$

$j = 1, \dots, M$

Meas. w/ $\{E_1, \dots, E_M\}$

Guess j .

Upper Bound



?



Prepare p_j w/ prob. $1/M$
 $j=1, \dots, M$

Meas. w/ $\{E_1, \dots, E_M\}$
Guess j .

Thm (Pretty Good Measurement): Let $\rho := \frac{1}{M} \sum_{j=1}^M p_j$,

$$E_j = \rho^{-\frac{1}{2}} p_j \rho^{-\frac{1}{2}} / M \quad . \quad \text{Then}$$

$$P_{\text{succ}} \geq \underbrace{(\text{optimal prob. of success})^2}_{P_{\text{opt}}}$$

Proof:

$$\begin{aligned} P_{\text{opt}} &= \frac{1}{M} \sum_{j=1}^M \text{Tr}(F_j P_j) \\ &= \frac{1}{M} \sum_{j=1}^M \text{Tr}\left(\rho^{\frac{1}{4}} F_j \rho^{\frac{1}{4}} \rho^{-\frac{1}{4}} P_j \rho^{-\frac{1}{4}}\right) \\ (\text{Cauchy-Schwarz}) &\leq \frac{1}{M} \sum_{j=1}^M \left\| \rho^{\frac{1}{4}} F_j \rho^{\frac{1}{4}} \right\|_2 \left\| \rho^{-\frac{1}{4}} P_j \rho^{-\frac{1}{4}} \right\|_2 \\ (\text{Cauchy-Schwarz}) &\leq \frac{1}{M} \underbrace{\sqrt{\sum_{j=1}^M \left\| \rho^{\frac{1}{4}} F_j \rho^{\frac{1}{4}} \right\|_2^2} \sqrt{\sum_{j=1}^M \left\| \rho^{-\frac{1}{4}} P_j \rho^{-\frac{1}{4}} \right\|_2^2}}_{\leq 1} \\ \left(\sum_j F_j = \mathbb{1} \right) &\leq \sqrt{\sum_{j=1}^M \left\| \rho^{-\frac{1}{4}} P_j / M \rho^{-\frac{1}{4}} \right\|_2^2} \\ &= \sqrt{P_{\text{succ}}} \end{aligned}$$

Upper Bound



?



Prepare p_j w/ prob. $1/M$

$j = 1, \dots, M$

Meas. w/ $\{E_1, \dots, E_m\}$

Guess j .

• PGM works well. Silver bullet?

Upper Bound



?



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• PGM works well. Silver bullet?

• Idea:

$$\rho = \frac{1}{M} \sum_{j=1}^M p_j \rightarrow \int d\rho \rho^{\otimes n}$$

↓
what measure?

Upper Bound



?



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Guess j .

- PGM works well. Silver bullet?

- Idea: $\rho = \frac{1}{M} \sum_{j=1}^M \rho_j \rightarrow \int d\rho \rho^{\otimes n}$

↓
what measure?

- Symmetries to exploit: $[\int d\rho \rho^{\otimes n}, U^{\otimes n}] = 0$.

- Schur-Weyl duality $\int dp \rho^{\otimes n} = \bigoplus_{\lambda \vdash n} c_\lambda \underbrace{\mathbb{1}_{g_\lambda}}_{\text{irrep of } GL(\delta)} \otimes \underbrace{\mathbb{1}_{p_\lambda}}_{\text{irrep of } S_n}$

$$= \sum_\lambda c_\lambda \prod_\lambda$$

- Schur-Weyl duality $\int dp \rho^{\otimes n} = \bigoplus_{\lambda \vdash n} c_\lambda \underbrace{\mathbb{1}_{g_\lambda}}_{\text{irrep of } GL(d)} \otimes \underbrace{\mathbb{1}_{p_\lambda}}_{\text{irrep of } S_n}$

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The PGM becomes

$$M_\sigma d\sigma = \left(\int dp \rho^{\otimes n} \right)^{-\frac{1}{2}} \sigma \left(\int dp \rho^{\otimes n} \right)^{-\frac{1}{2}}$$

$$M_\sigma d\sigma = \left(\int d\sigma P^{\otimes n} \right)^{-\frac{1}{2}} \sigma \left(\int d\sigma P^{\otimes n} \right)^{-\frac{1}{2}} d\sigma$$

$$= \sum_{\lambda} \frac{1}{c_\lambda} \prod_\lambda \sigma^{\otimes n} \prod_\lambda d\sigma$$

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- If we believe in the PGM, and we believe tomography
double w/ $\sim d^2/\epsilon^2$ copies, this will do it!

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If we believe in the PGM, and we believe tomography
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$\text{Tr}(M_\sigma \rho^{\otimes n}) d\sigma$ should be low when σ far from ρ .

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$\text{Tr}(M_\sigma \rho^{\otimes n}) d\sigma$ should be low when σ far from ρ .

$$\underbrace{\lesssim e^{\delta^2} (1 - \|\rho - \sigma\|_1^2)^n}_{d\sigma}$$

$$M_\sigma d\sigma = \left(\int d\rho \rho^{\otimes n} \right)^{-\frac{1}{2}} \sigma \left(\int d\rho \rho^{\otimes n} \right)^{-\frac{1}{2}} d\sigma$$

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$$\underbrace{\quad}_{\leq} e^{d^2} (1 - \|\rho - \sigma\|_1^2)^n d\sigma$$

$$\Rightarrow n \sim d^2 / \|\rho - \sigma\|_1^2 \text{ suffices.}$$

Final Thoughts

- We showed $\sim \delta^2/\epsilon^2$ necessary and sufficient for tomography in most generous setting.
↳ tight bounds recently shown in more realistic ones
- "Shadow tomography" → learning full state not necessary.
↳ what is the copy complexity?
- Other learning / testing problems.

→ Quantum \approx classical often, but showing is hard. Is this true for computational problems as well?