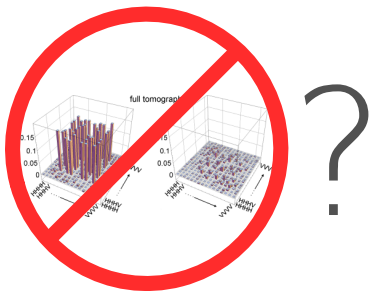


Should I really reconstruct a state from that data?

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Tomography: issues, tweaks & tricks

Reasons for state tomography:

- demonstration of a good control of a quantum system
- experimental verification of interesting states
- most general dataset for later analysis



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Difficulties:

- exponential cost in measurement time
↳ compressed sensing, permutation invariant tomography, ...
- exponential cost of classical computation time
↳ compressed sensing, permutation invariant tomography, ...
- low sampled data with low statistical significance
↳ maximum likelihood estimation, Bayesian methods, maximum entropy principle, Ockham's razor. ...

Why maximum likelihood?

- naïve state reconstruction does not yield a valid quantum state (negative eigenvalues)
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The maximum likelihood methods yields a valid quantum state, even for completely messed up data.

- Why do we get an nonphysical state in the first place?
- What if systematic errors are present?



Simple schemes

Example: for Q qubits, measure locally all Pauli operators,

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More generally:

Experimenter performs measurements $E_{i|\alpha}$ yielding

$$p_{i|\alpha} = \text{tr}(E_{i|\alpha} \rho_{\text{exp}}) \text{ or } \mathbf{p} = M[\rho_{\text{exp}}].$$

Complete tomography, if $M[\rho] \neq M[\rho']$.

Low sampling issues

- in an N_α -fold measurement of setting α , we sample $(n_{\alpha,1}, n_{\alpha,2}, \dots)$ from the multinomial distribution $(p_{\alpha,1}, p_{\alpha,2}, \dots)$
- ↪ frequencies $f_{i|\alpha} = n_{i|\alpha}/N_\alpha$
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Is ϱ_{ls} a bad idea?

- + easy and fast to compute
- + converges to true value
- not appropriate for small N_α
- has negative eigenvalues

Hoeffding beats negative eigenvalues

Suppressed negative expectation values

Choose an arbitrary vector $|\psi\rangle$. Then for $t > 0$,

$$\Pr[\langle\psi|\varrho_{\text{ls}}|\psi\rangle < -t] \leq \exp[-t^2 N/\text{const}_\psi].$$

\hookrightarrow *Negative expectation values occur rarely.*

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A test procedure:

- take two datasets I and II for the same ϱ_{exp}
- choose $|\psi\rangle$ according to $\varrho_{\text{ls}}^{\text{I}}$
- then it is unlikely to observe $\langle\psi|\varrho_{\text{ls}}^{\text{II}}|\psi\rangle < -t$

\hookrightarrow negative eigenvalues are in random directions

The issue of overcomplete tomography

- Usually tomography is performed in an overcomplete setup.
- Example: 3^Q Pauli measurements with $2^Q - 1$ outcomes vs. $4^Q - 1$ entries in the density matrix.
- Reduction from $\sim 6^Q$ to $\sim 4^Q$ dimensions.

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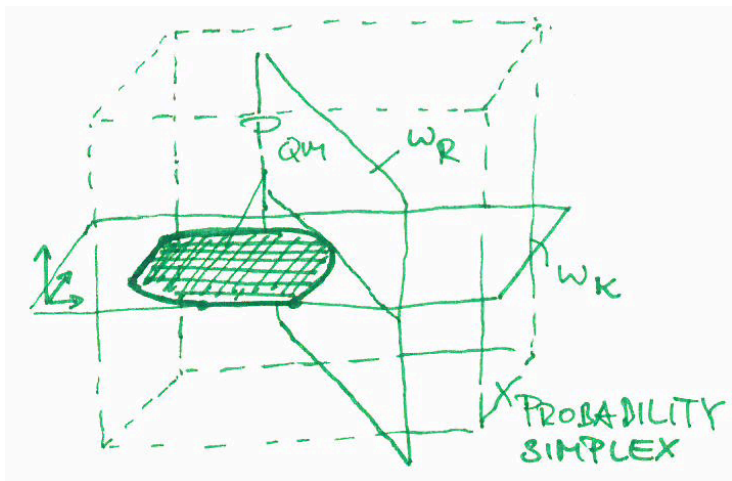
Linear dependencies

Let \mathbf{p}_{ls} be the probabilities predicted from ϱ_{ls} . Then for $t > 0$,

$$\Pr[|(\mathbf{p}_{\text{ls}}^{\text{I}} - \mathbf{f}^{\text{I}}) \cdot (\mathbf{p}_{\text{ls}}^{\text{II}} - \mathbf{f}^{\text{II}})| > t] \leq 2 \exp[-t^2 N / \text{const}_{\mathbf{f}^{\text{I}}}] .$$

\Leftrightarrow *Deviation from predicted probabilities are not systematic.*

Like a witness



We entered the regime of hypothesis testing

Under the assumption $(\text{data}) \sim \mathbf{p} = M[\varrho_{\text{exp}}]$, we arrived at

$$\Pr[T(\text{data}) > t] \leq \exp[-t^2 N / \text{const}_T]$$

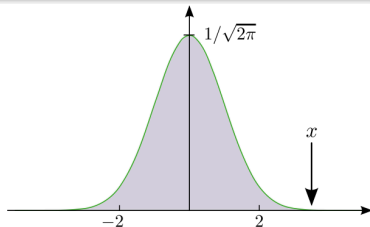
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Sample from Gaussian process, get value x ,
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$$\Pr[|x| > t] = 1 - \int_{-t}^t e^{-x^2/2} / \sqrt{2\pi} dx.$$



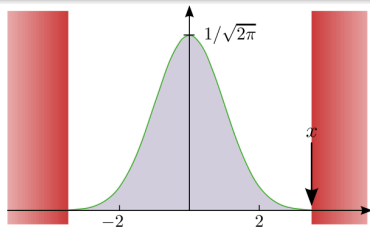
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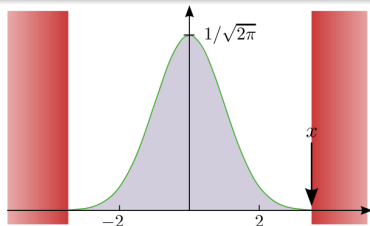
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- We say, x is *excluded* by a x - σ region, e.g. 1σ : 31.7%, 2σ : 4.56%, 3σ : 0.270%, ...
- Let's apply this to our methods – shortly.

Likelihood ratio test

- In overcomplete tomography, M fails to be surjective ($6^Q \rightarrow 4^Q$).
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Likelihood ratio test/Wilks theorem

If $\mathbf{p} \in \text{range } M$, then, as $N \rightarrow \infty$,

$$\Pr[2N \inf\{D(\mathbf{f}||\mathbf{p}) \mid \mathbf{p} \in \text{range } M\} < t] \rightarrow Q(\Delta/2, t/2)$$

where

- $D(\mathbf{f}||\mathbf{p}) = \mathbf{f} \cdot \log \mathbf{f} - \mathbf{f} \cdot \log \mathbf{p}$ is the relative entropy,
- $Q(s, x) = \Gamma(s, x)/\Gamma(s)$ is the regularized Gamma function,
- $\Delta = \dim \text{range}(M)^\perp$ is the dimension deficit.

Empirical results

Given the data, can we exclude, that $\mathbf{p} = M[\rho_{\text{exp}}]$ is a valid model?

state	Q	N	X-talk	w_K	w_R	LR	LR*
GHZ	4	2500	20%	4.0σ	14σ	19σ	$>3.3\sigma$
		750	12%	-	5.0σ	3.6σ	3.3σ
		300	$<3\%$	0.3σ	0.7σ	(2.6σ)	2.0σ
Bell	2	61650	$<3\%$	-	-	0.6σ	0.7σ
$ \uparrow\uparrow\uparrow\uparrow\rangle$	4	250	?	1.6σ	0σ	(3.4σ)	2.8σ
BE	4	5200	$<3\%$	0.08σ	0.8σ	0.9σ	0.9σ
W	5	100	4%	0.6σ	0.1σ	(3.3σ)	1.8σ

Summary

- The negative eigenvalues in a linear reconstruction yield very unlikely a negative expectation value.
- Systematic errors can be distinguished from statistical errors using witness-like structures or the likelihood ratio test.
- Our method works on current experimental data.
- [\[arXiv:very.soon\]](#)