

Quantum state estimation

Challenges and solutions, methods of
statistical inference

—

Arthur Vesperini

Quantum statistics: generalities

Density operators and measurement
basis



Quantum statistics: 2-level systems

Classical coin

$$x = 0, 1$$

Quantum coin

$$|x\rangle = \alpha|0\rangle + \beta|1\rangle$$



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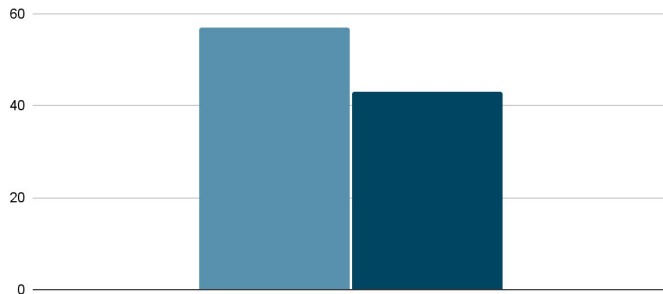
$$P(x) = p_0\delta(x) + p_1\delta(x - 1)$$

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Population sample

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Quantum statistics: 2-level systems

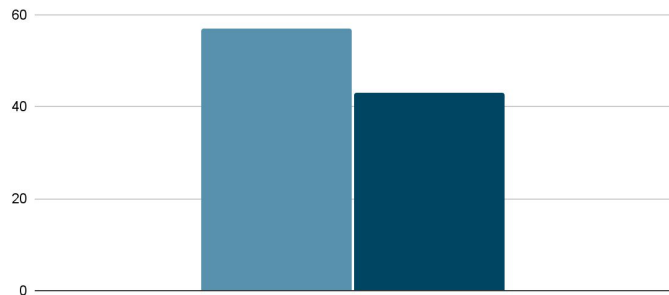
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Quantum coin

$$|x\rangle = \alpha|0\rangle + \beta|1\rangle$$

$$P(|x\rangle) = \text{Tr}(\rho|x\rangle\langle x|)$$

$$\rho = \begin{pmatrix} p_0 & p_{01} \\ p_{01}^* & p_1 \end{pmatrix}$$



Quantum statistics: 2-level systems

Qubit / spin-1/2

“Classical” probabilistic mixture
(diagonal matrix) of $|0\rangle$ and $|1\rangle$

$$\rho_c = \frac{1}{2}(|0\rangle\langle 0| + |1\rangle\langle 1|)$$

Pure quantum superposition
(off-diagonal “interferences”)

$$|+\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)$$

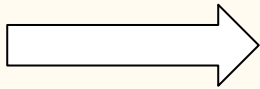
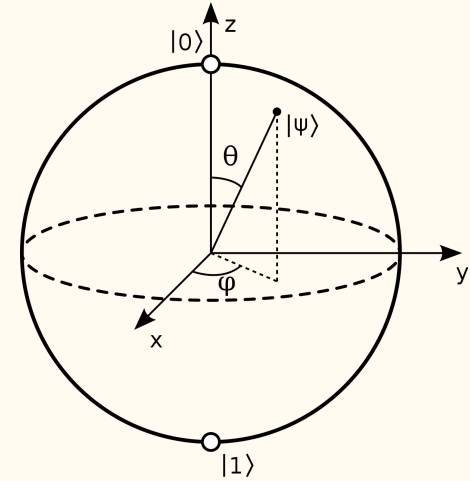
$$\rho_+ = \frac{1}{2}(|0\rangle\langle 0| + |1\rangle\langle 1| + |0\rangle\langle 1| + |1\rangle\langle 0|)$$

Quantum statistics: 2-level systems

Qubit / spin-1/2

$$\rho = \begin{pmatrix} p_0 & p_{01} \\ p_{01}^* & p_1 \end{pmatrix}$$

$$= \frac{1}{2} (\mathbb{I} + n_x \sigma_x + n_y \sigma_y + n_z \sigma_z)$$



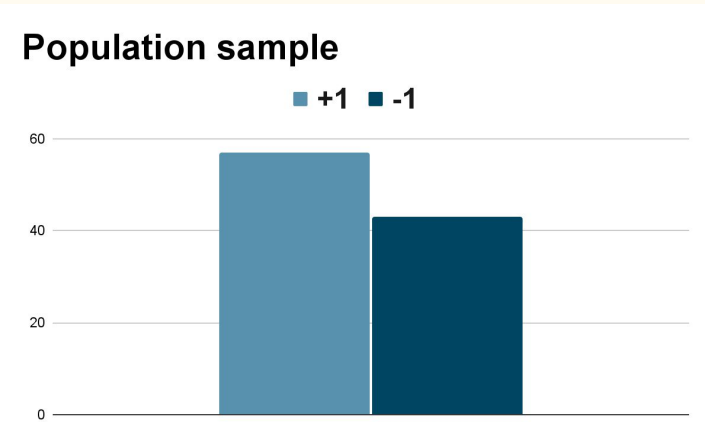
Tomographic completeness: 3 independent basis of measurement (“*quorum*” of observables).

Quantum statistics: 2-level systems

Classical coin

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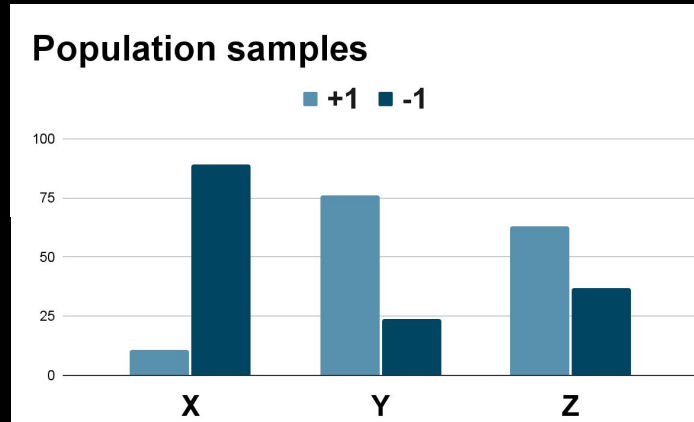
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Standard scheme: Linear Inversion

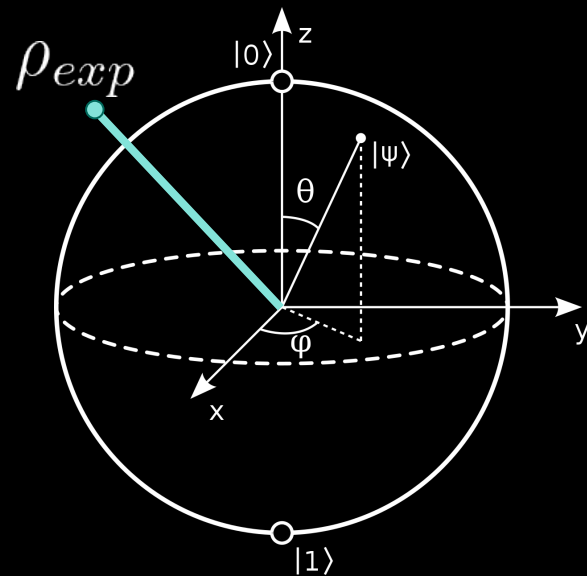
The problem of positivity

Main references:

[Řeháček et al., 2001;

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Linear inversion: principle

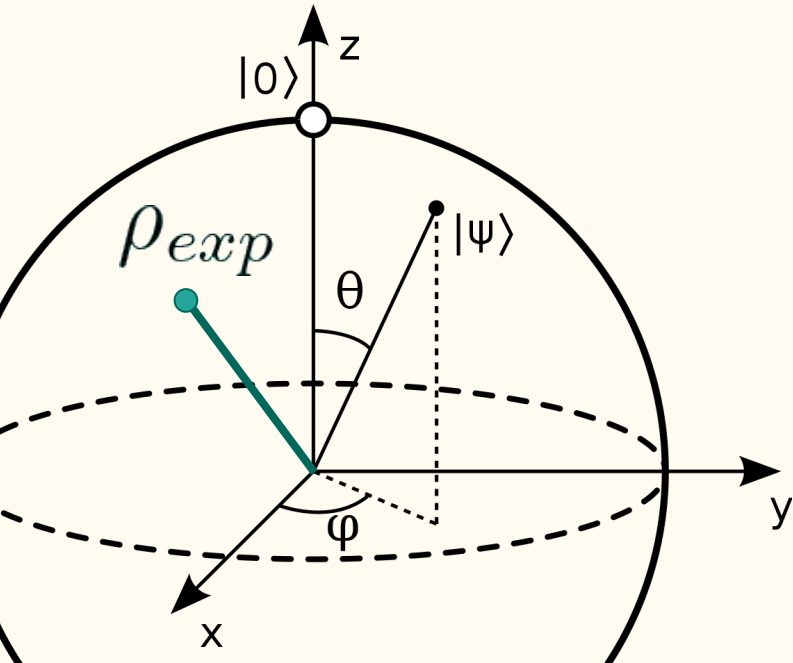
Direct linear inversion

$$f_j = \text{Tr}(\rho_{exp} |x_j\rangle \langle x_j|) = \langle x_j | \rho_{exp} |x_j\rangle$$

$$\longrightarrow \rho_{exp} = \sum_j f_j |x_j\rangle \langle x_j|$$

Linear inversion: 2-level systems (qubit/spin-1/2)

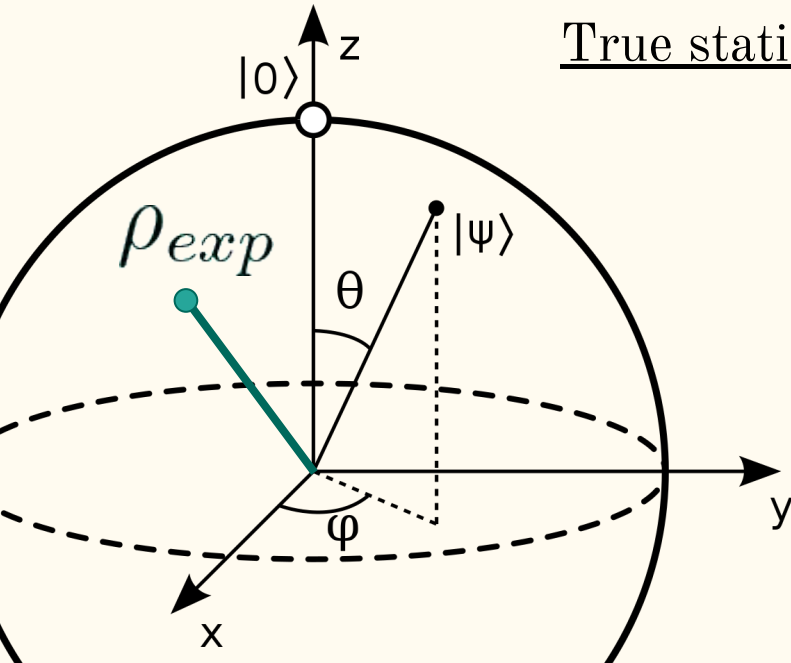
$$\rho_{exp} = \frac{1}{2} (\mathbb{I} + \vec{n}_{exp} \cdot \vec{\sigma})$$



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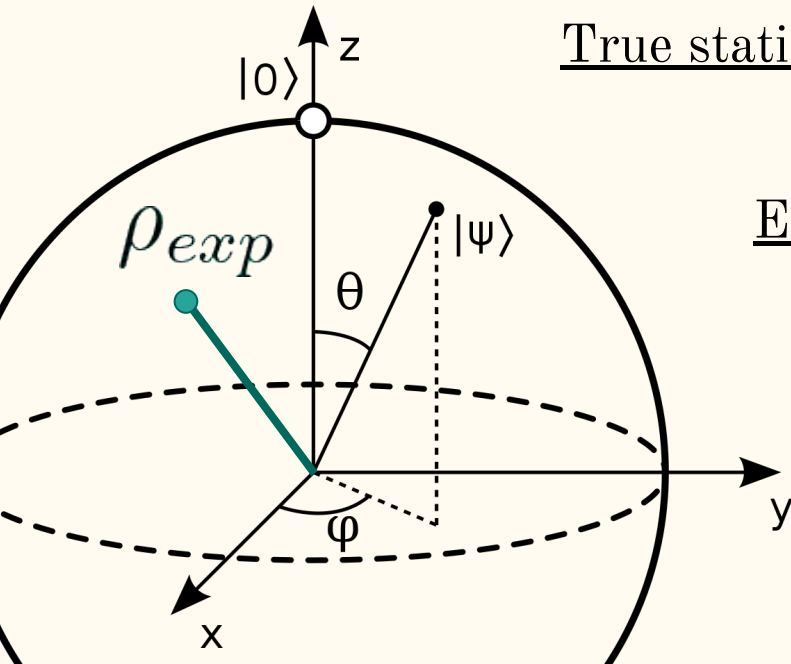
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True statistics: $\rho = |0\rangle\langle 0| = \frac{1}{2}(\mathbb{I} + \sigma_z)$



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Experimental outcome:

$$\vec{n}_{exp} = \begin{pmatrix} \langle \sigma_x \rangle_{exp} \\ \langle \sigma_y \rangle_{exp} \\ \langle \sigma_z \rangle_{exp} \end{pmatrix} = \begin{pmatrix} f_x - f_{-x} \\ f_y - f_{-y} \\ 1 \end{pmatrix}$$

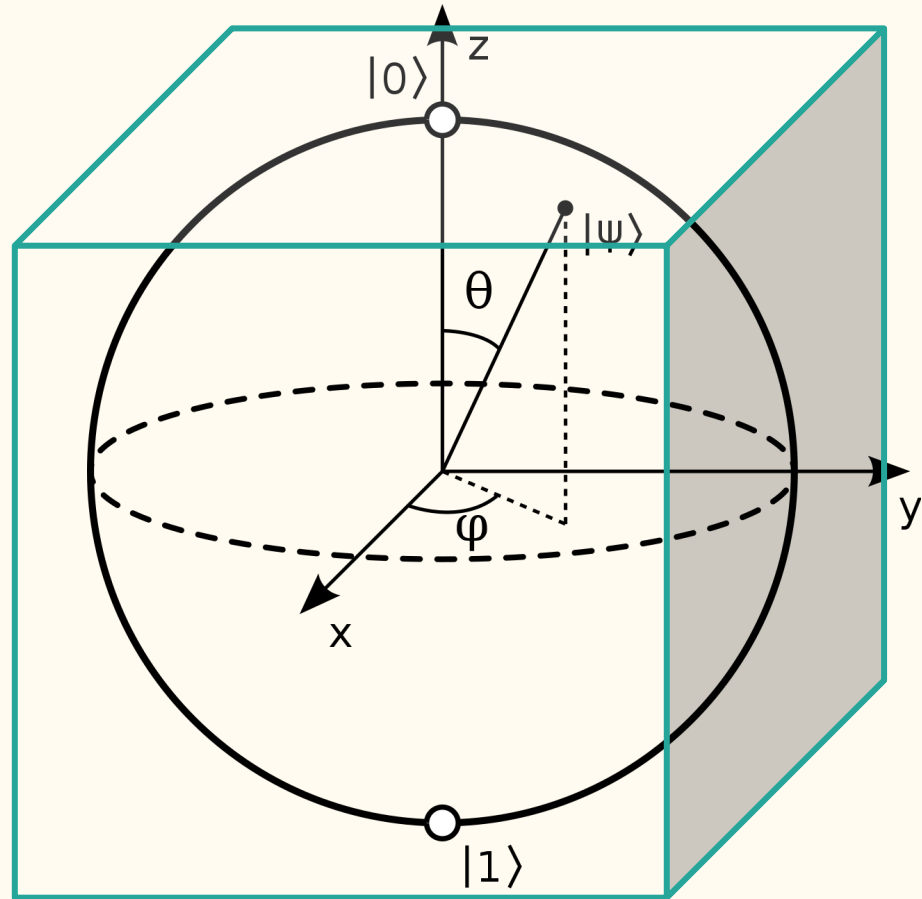
Linear inversion: 2-level systems (qubit/spin-1/2)

$$|n_i^{exp}| \leq 1$$

hence \vec{n}_{exp} belongs to a “Bloch cube”.

In particular, we might have:

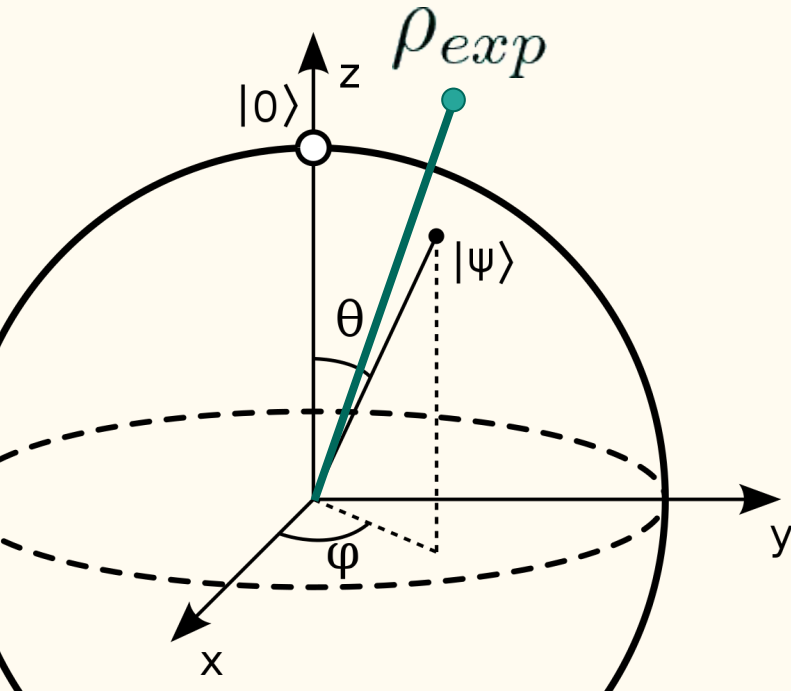
$$|\vec{n}_{exp}| > 1$$



Linear inversion: 2-level systems (qubit/spin-1/2)

$$p_{\pm n}^{exp} = \lambda_{\pm n} = \frac{1 \pm |\vec{n}|}{2}$$

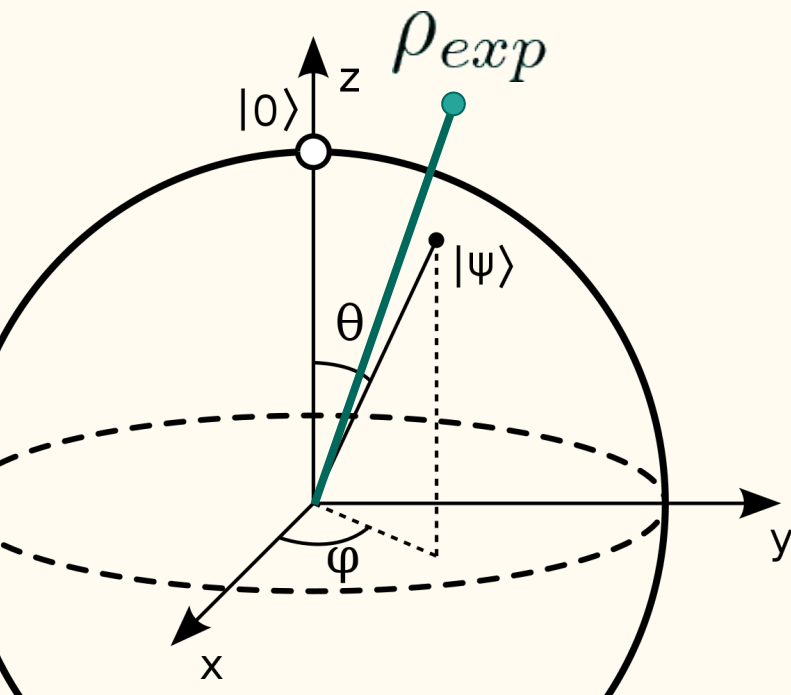
is negative or greater than one if $|\vec{n}| > 1$
(outside of the Bloch ball).



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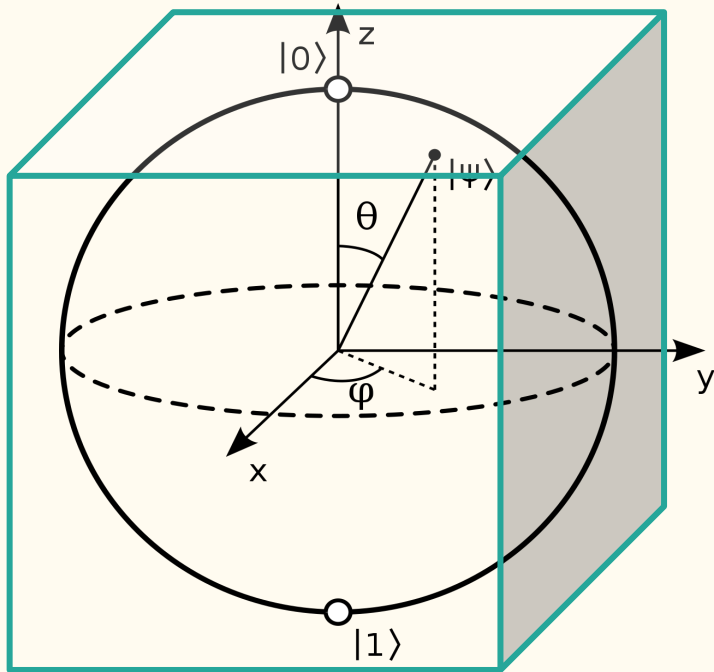
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**Generally ill-defined
probability weights**

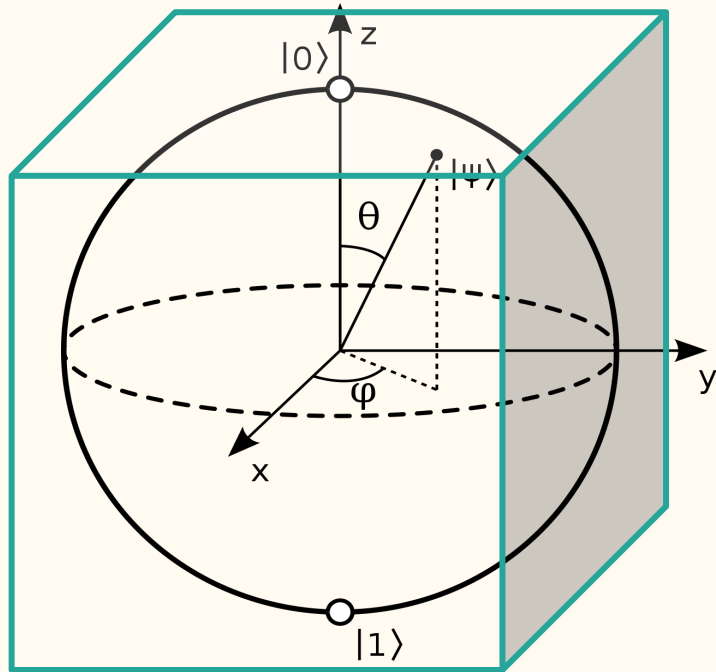
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In quantum systems, there always exists an *infinite number of measurement basis*.



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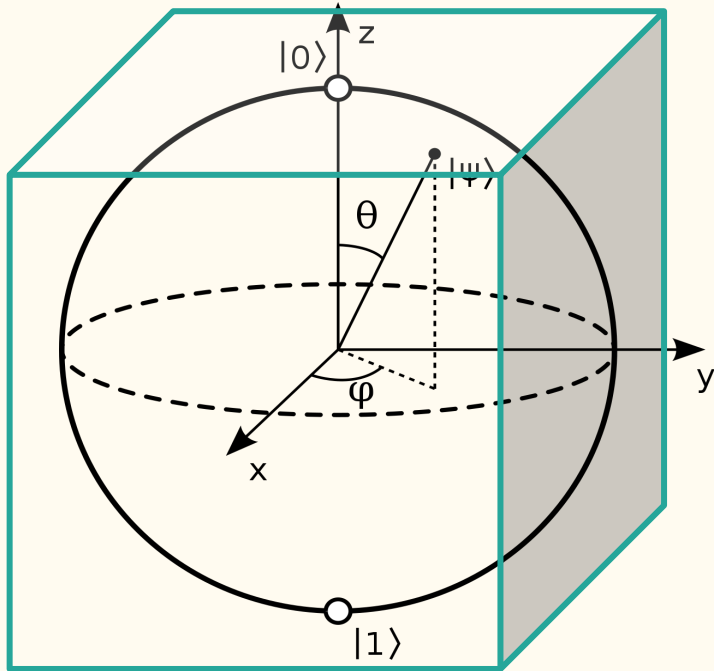
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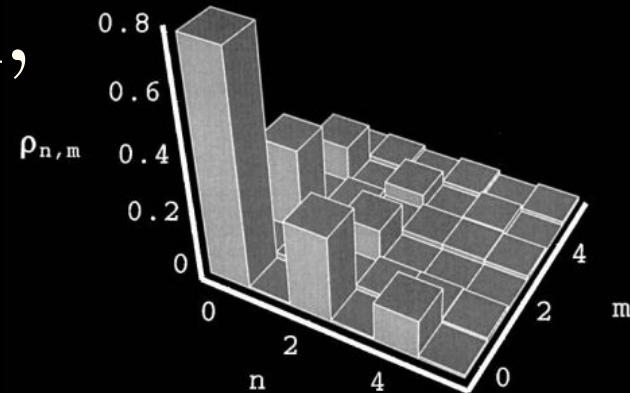
One can get negative probabilities for observables that have not been measured.

Maximum Likelihood Estimation

Principle, extremal equation,
algorithm and flaws

Main references:

[Banaszek, 1999;
Řeháček et al., 2001;
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MLE reconstruction of the density matrix
of a single-mode radiation field.
Image from K. Banaszek et al, 1999

MLE of quantum states: principle

What state is most likely to have produced a given experimental outcome $\mathcal{M} = \{|x_1\rangle, \dots, |x_N\rangle\}$?

Maximizing the likelihood functional:

$$(\mathcal{L}(\rho))^{1/n} = \prod_j^M p(|x_j\rangle|\rho)^{f_j}$$

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equivalent to minimizing the statistical distance (Kullback-Leibler divergence):

$$D(\mathbf{f}, \mathbf{p}) = - \sum_j^M f_j \ln(p_j)$$

MLE of quantum states: extremal equation

General solution: extremal equation

$$R\rho_e = \rho_e$$

with $R = \sum_j^M \frac{f_j}{\langle x_j | \rho_e | x_j \rangle} |x_j\rangle \langle x_j|$

MLE of quantum states: a simple algorithm

If eigenbasis is known, we have $\rho = \sum_k r_k |\psi_k\rangle\langle\psi_k|$,

hence $f_j = \langle x_j | \rho | x_j \rangle = \sum_k r_k |\langle x_j | \psi_k \rangle|^2$

Finding the eigenvalues is a linear positive problem for which converging algorithms are available (e.g. expectation-maximization algorithm).

MLE of quantum states: a simple algorithm

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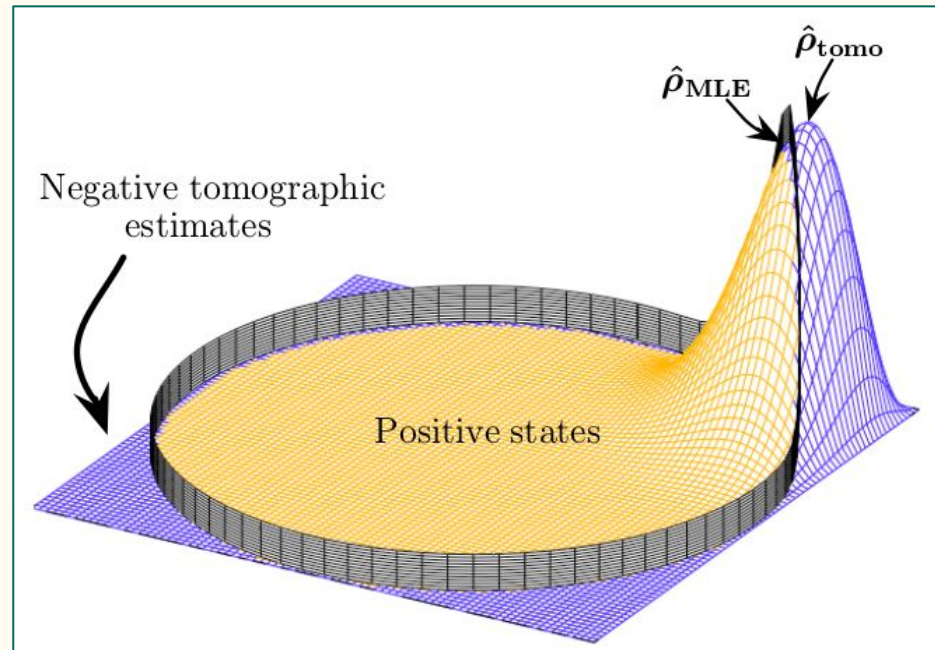
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MLE of quantum states: relation to linear inversion

Linear inversion \sim unconstrained MLE

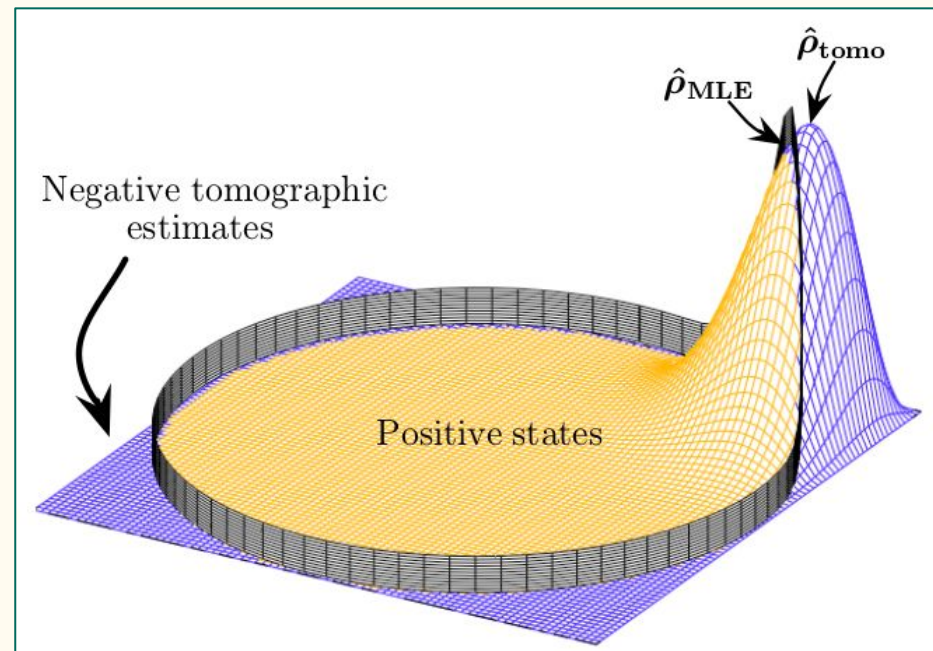


Example of likelihood function, with both constrained and unconstrained maxima shown, in a cross-section of the single-spin Bloch sphere. Figure from [Blume-Kohout, 2010]

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- $\rho_{\text{tomo}} \geq 0 \implies \max(\mathcal{L}(\rho)) = \mathcal{L}(\rho_{\text{tomo}})$

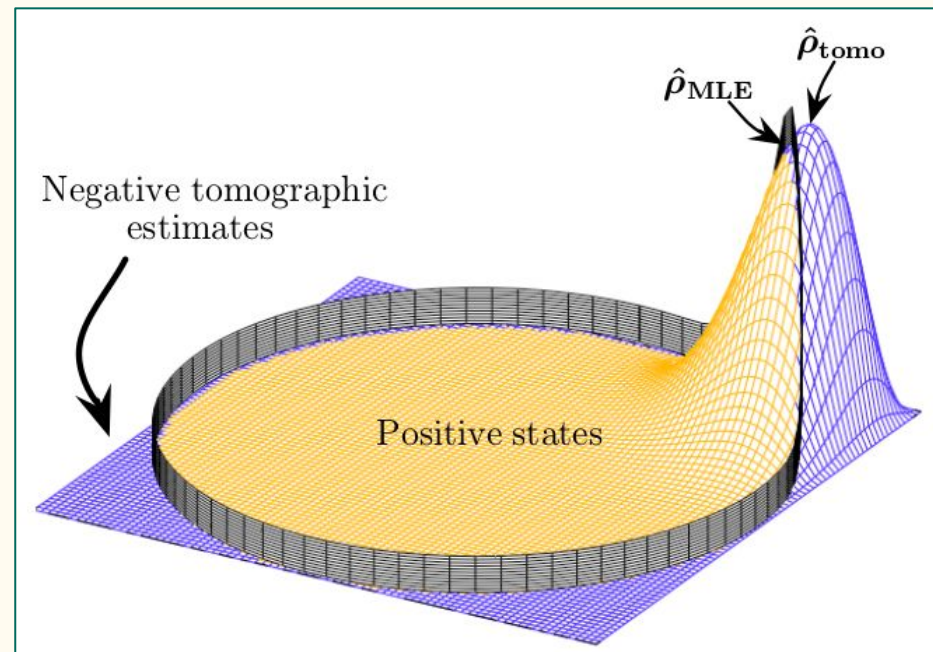


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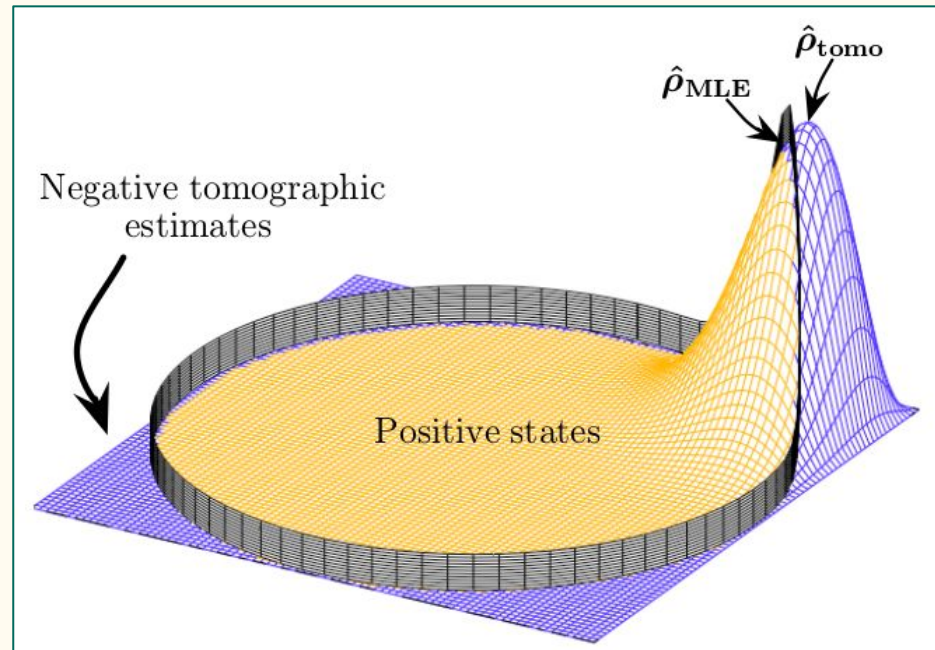
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A finite number of measurement cannot justify a vanishing probability!



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Flaws inherent to frequentist approaches to quantum state reconstruction.

Somehow related to the infinite number of observables: the estimation closest from observed frequencies fails to describe unmeasured events.

Bayesian Mean Estimation

Principle, extensivity, pros and cons

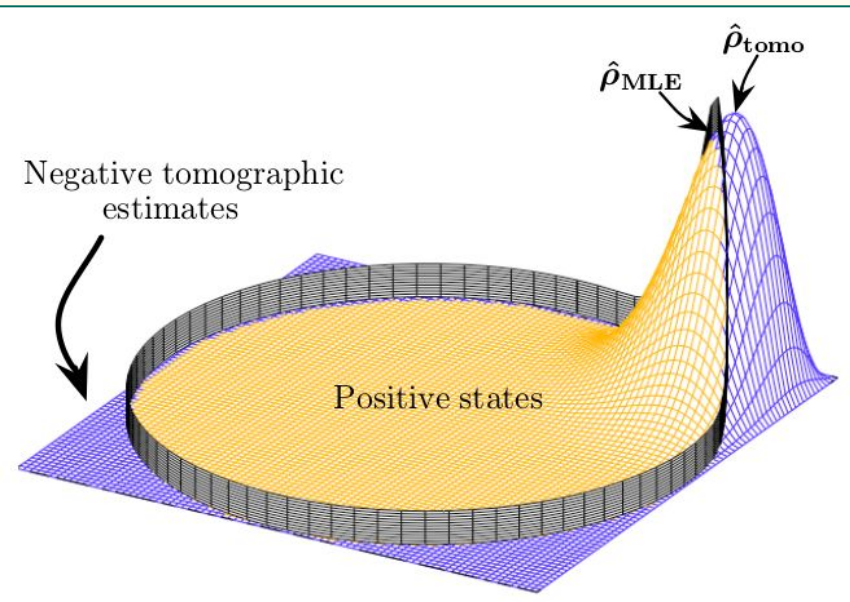
Main reference:
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$$p(\rho|\mathcal{M}) = \frac{p(\mathcal{M}|\rho)p(\rho)}{p(\mathcal{M})}$$

Bayes theorem

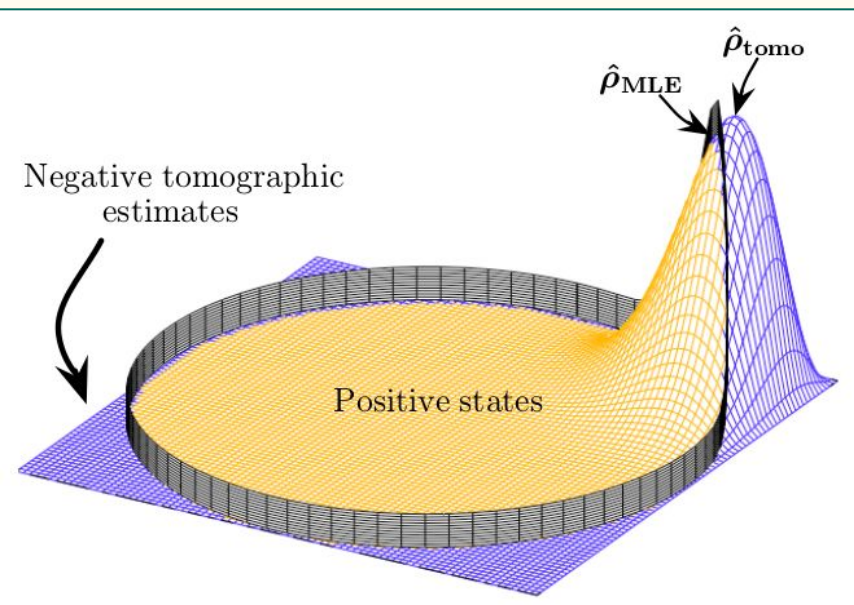
BME of quantum states: principle

$$\rho_B = \int \rho \pi_f(\rho) d\rho = \frac{\int \rho \mathcal{L}(\rho) \pi_0(\rho) d\rho}{\int \mathcal{L}(\rho) \pi_0(\rho) d\rho}$$



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Provided a reasonable (*robust*) prior was chosen:

- no vanishing probabilities
- extensive with respect to sample size

BME of quantum states is extensive

Extreme example: $\mathcal{M} = \{|0\rangle, \dots, |0\rangle\}$, $\pi_0 = c$

$$\begin{aligned}\rho_B &= \int \pi_f(\rho) \rho d\rho \\ &= \frac{\int (\rho_{00})^N \rho d\rho}{\int (\rho_{00})^N d\rho_{00}} = (N + 1) \int (\rho_{00})^N \rho d\rho\end{aligned}$$

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$$(\rho_B)_{00} = (N + 1) \int (\rho_{00})^{N+1} d\rho_{00} = \frac{N + 1}{N + 2}$$

$$(\rho_B)_{11} = \frac{1}{N + 2}$$

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- Sensitive to choice of prior.
 - Relatively low performance (numerical integration, e.g. Metropolis-Hastings algorithm).

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- BME: most rigorous and computationally costly, preferred when accurate *prediction* and well-defined error bars are crucial (e.g. quantum computers).

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Frequentist approach:
accurate description
of the past

Bayesian approach:
fitted for
predictions of the
future

Bibliography

- Banaszek, K., G. M. D'Ariano, M. G. A. Paris, and M. F. Sacchi. 1999. "Maximum-Likelihood Estimation of the Density Matrix." *Physical Review A* 61 (1): 010304. <https://doi.org/10.1103/PhysRevA.61.010304>.
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Thanks for your attention.

