

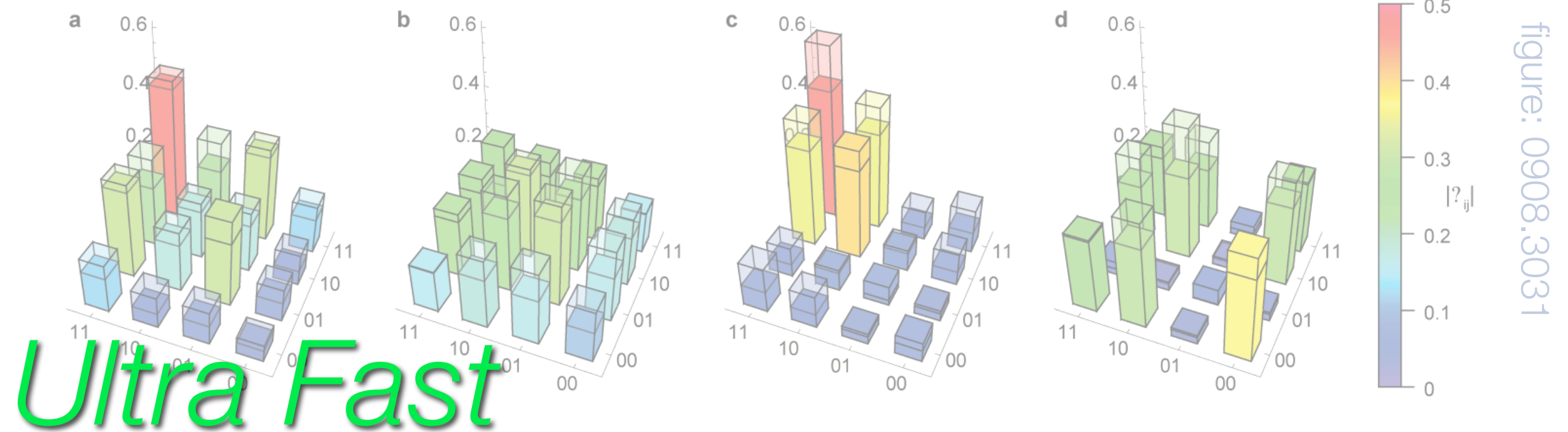
figure: 0908.3031

# Quantum State Tomography

Steve Flammia  
Perimeter Institute  
CQIQC: Aug 24, 2009

Joint work with:

- 1) S Becker, J Eisert, D Gross, Y-K Liu &
- 2) S Bartlett, D Gross, R Somma



# *Ultra Fast* Quantum State Tomography

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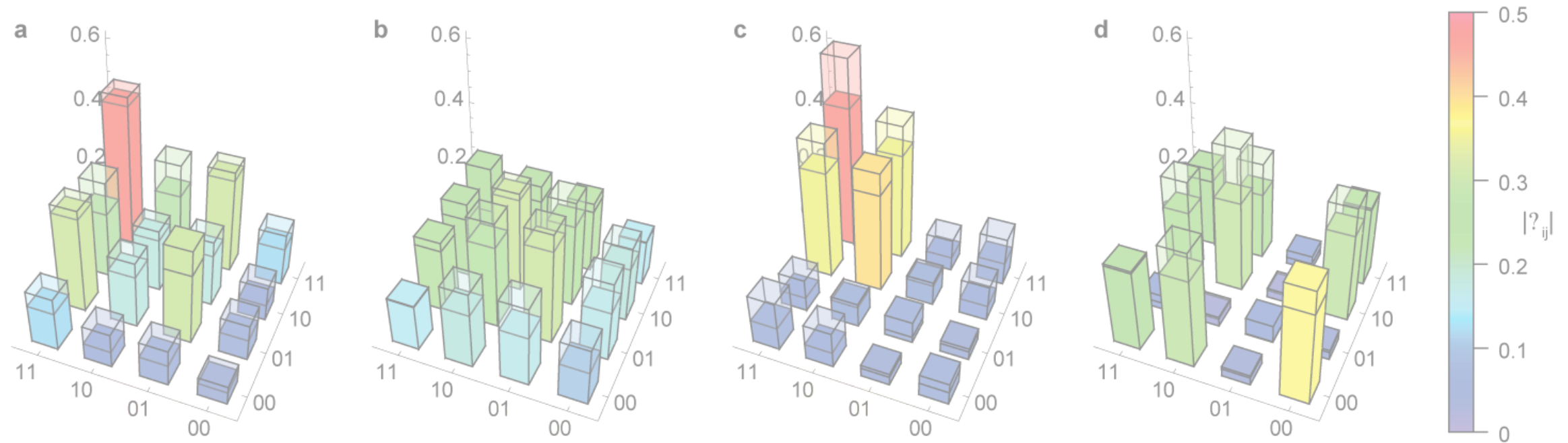


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# Quantum State Tomography

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Q: What is it?

- ✿ Reconstructing a classical description of a quantum state from single-copy measurements and classical post-processing.

Q: Why is it so difficult?

- ✿ the dimension  $d=2^n$  is exponentially large, so one must measure an exponential number of observables....
- ✿ which determine exponentially small parameters, so you need to make exponentially many measurements...
- ✿ and once you have all the data, finding a compatible density operator takes an exponential amount of classical computing!



# Why do we care about tomography?

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- ✿ Short answer: we don't. What we really care about is certifying state preparations...
- ✿ ...but full tomography could still be useful to characterize noise processes (see also quantum process tomography).

# of qubits needed to do something “interesting”

$n < 10$



$n > 100$



# Pure State Tomography

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- ✱ The most interesting quantum states are pure; can we do tomography of pure states with fewer measurement settings? [1]

Yes! For states of rank  $r$ ,  $O(r d \log^2(d))$  measurements suffice.

- ✱ This result is quite robust and similar performance holds even if...
  - ✱ ...the measurements are corrupted by stochastic, deterministic, or adversarial noise.
  - ✱ ...the state is full rank, but has large purity.
  - ✱ ...the efficiency of the classical post-processing is  $O(\text{poly}(d))$ .
- ✱ Moreover, the scheme requires just Pauli measurements.
- ✱ The result is heralded, so no *a priori* promise of large purity or low entropy is necessary.

# What is Compressed Sensing?

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Pioneered by: Candès, Tao, Donoho, ...

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Shoot,  
then  
compress



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# What is Compressed Sensing?

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Shoot,  
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Take only as many samples as  
you need, then post-process  
to reconstruct the image!

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# What is Compressed Sensing?

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Finding the sparsest  
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Use a convex relaxation!

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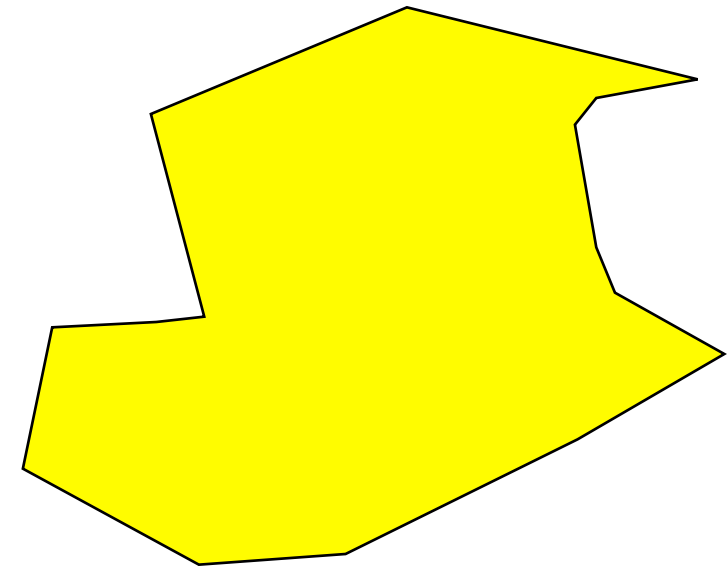
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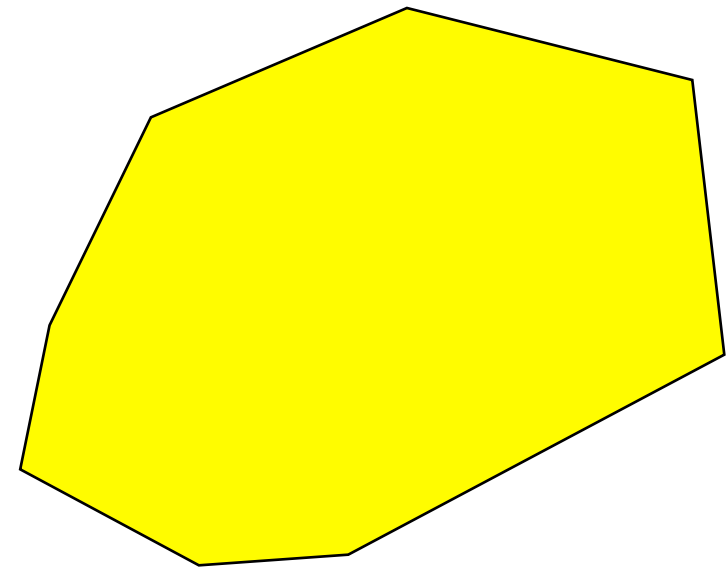
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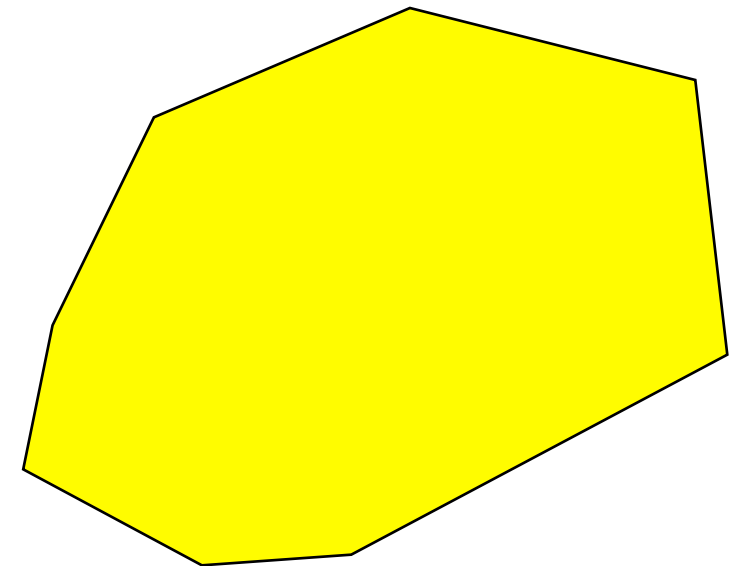
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## Problem:

Unlike classical info., quantum information contains a *basis*.

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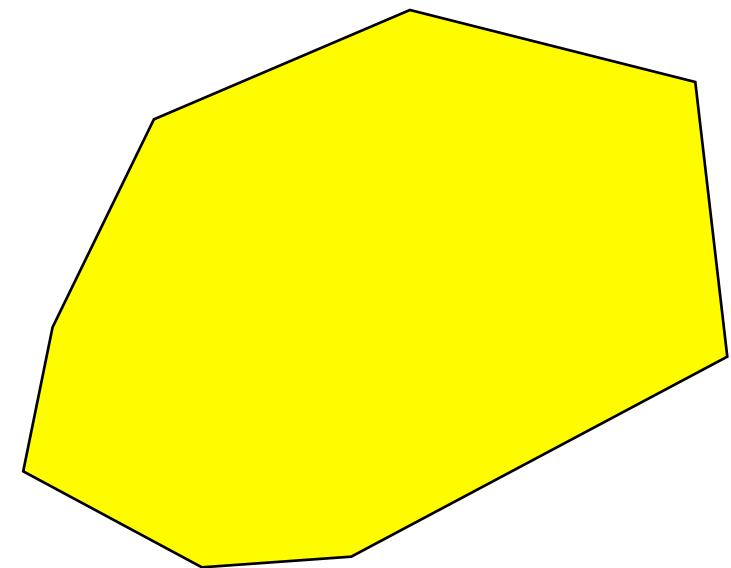
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Finding the sparsest vector consistent with the sampling is NP-hard

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## Problem:

Unlike classical info., quantum information contains a *basis*.

## Solution:

Use *rank minimization* instead (and its convex relaxation, of course.)

# Tomography of low-rank states

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Measure observables  $A_k$  to get expectations  $b_k$ , and collect them each into vectors of length  $m$ , with  **$m = \mathcal{O}(r d \log^2(d))$** .

Now we can solve the convex optimization

$$\min \operatorname{Tr}(X) \quad : \quad \|\mathcal{A}(X) - b\|_2 \leq \delta, \quad X \geq 0$$

**Theorem 1:** If the  $A_k$  are randomly chosen Paulis, then this convex optimization has a unique solution as  $\delta$  vanishes, with overwhelming probability.

This is essentially optimal, since  $m = \Omega(rd)$  follows from simple parameter counting.



## Robustness to noise

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Suppose instead that we measure a different state, but still pretty close (as measured by the 2-norm),  $\|\rho - \sigma\|_2 < \delta$

Solving the same optimization as before,

$$\min \operatorname{Tr}(X) : \|\mathcal{A}(X) - b\|_2 \leq \delta, X \geq 0$$

we obtain **Theorem 2**, a result about robustness

$$\|\rho_{\text{opt}} - \rho\|_2 = \delta O\left(\frac{d}{\log d}\right)$$

As we will see from the numerics, there is probably room for improvement in this bound.

# Heralding the success of the protocol

---

Theorem 1 holds under the promise of low rank, and theorem 2 says that it is robust under small perturbations. Can this be improved to a **heralded** scheme?

Yes!

Using the Pauli measurements, we can estimate the **purity** of the state.

If the purity is large enough, then theorems 1 and 2 apply. Otherwise, we can abort, or do e.g. maximum likelihood estimation.

# Testing the algorithm

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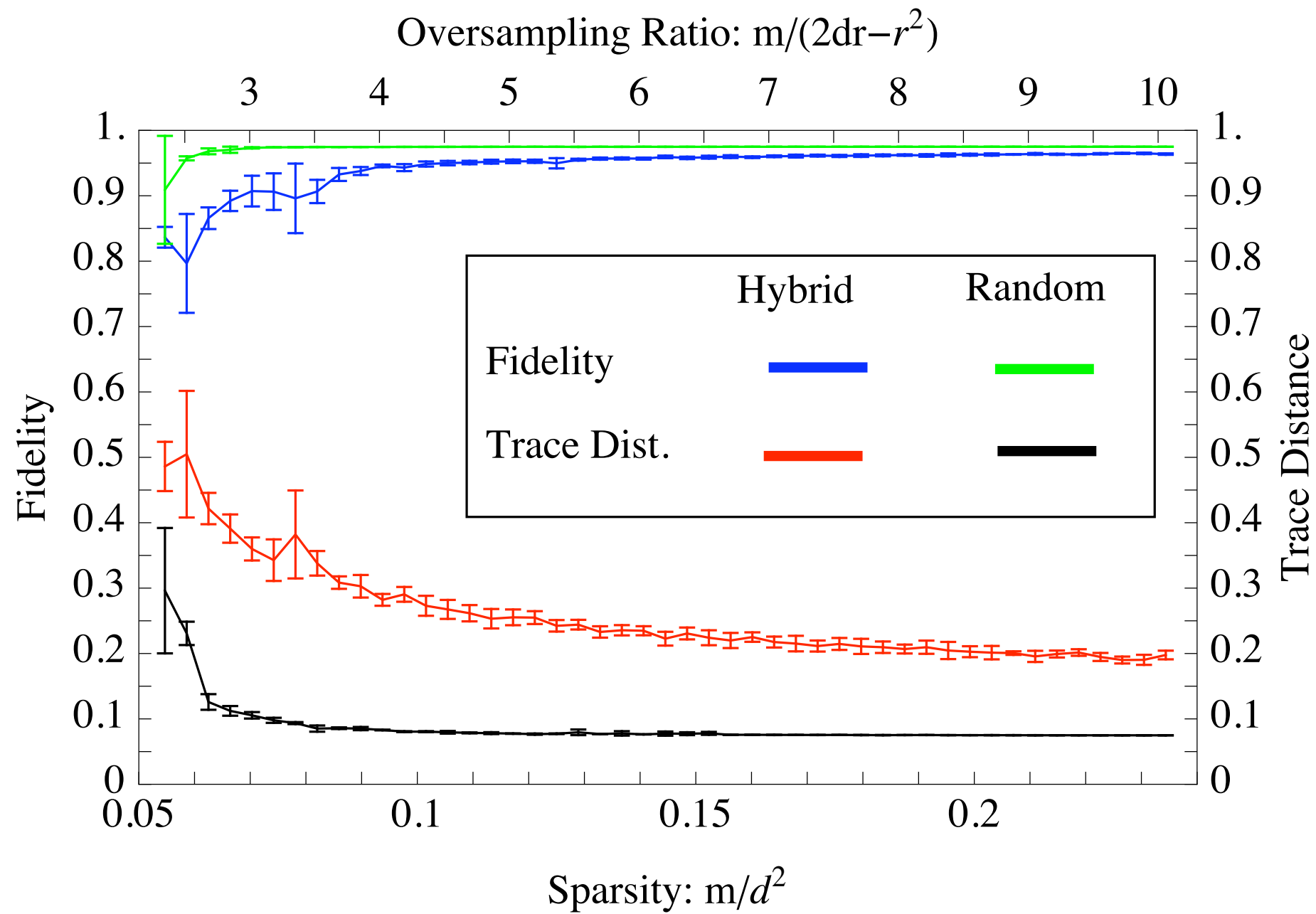
Choose a random pure state from a  $d \times r$  system and trace out the  $r$ -dimensional ancilla.

Add statistical noise and additional depolarizing noise.

Solve the convex program to get the estimate of the state. (ask me about the details!)

Matlab interlude

# Testing the algorithm



# Matrix Product State Tomography

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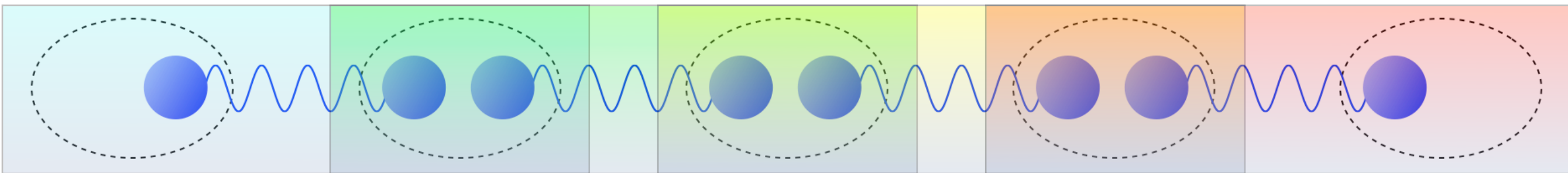
- ✿ Many interesting quantum states can be well approximated by a matrix product state with a small bond dimension. Can we use this to our advantage?

Yes, we can do certified tomography of certain  $n$ -qubit quantum states in time  $\text{poly}(n)$

- ✿ The idea is to produce efficient classical descriptions of large quantum states by doing tomography of the MPS description directly, rather than the exponentially large description as a vector in Hilbert space.

# State certification for non-degenerate MPS

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For each local density operator,  
we do complete local tomography.

$$\rho_j$$

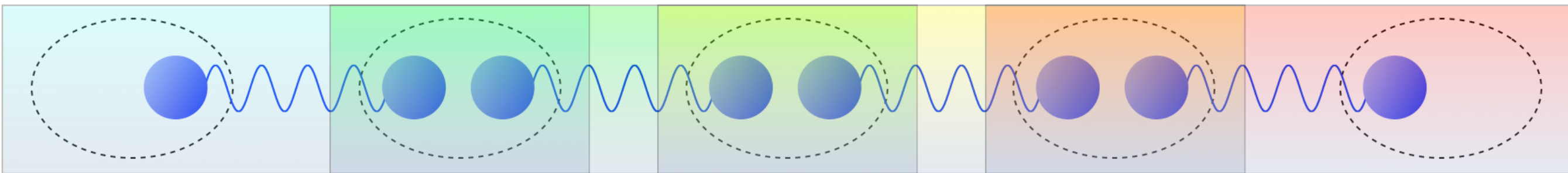
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(Prob :  $1 - \delta_0$ )



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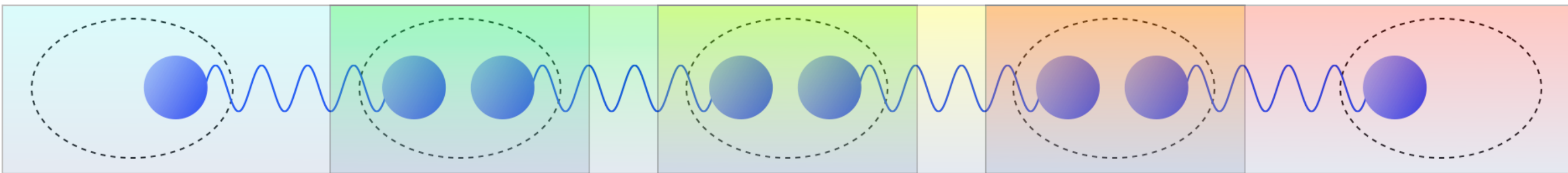
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Build a “fake” parent  
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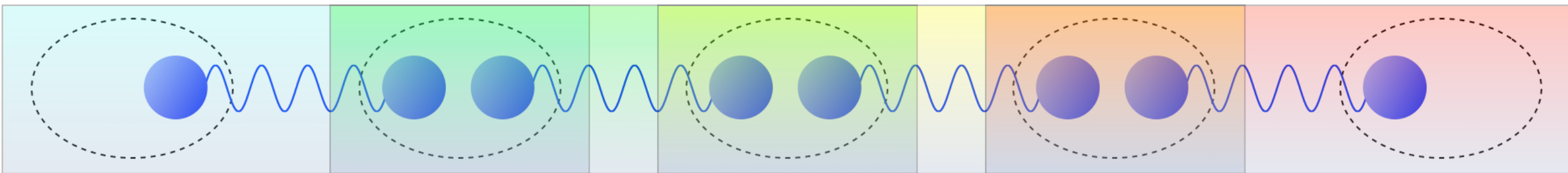
$$\Pi_j = \text{null}(\sigma_j) \quad H = \sum_j \Pi_j$$

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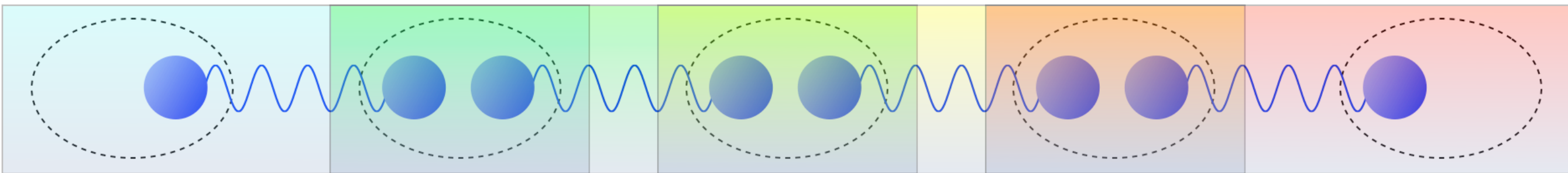
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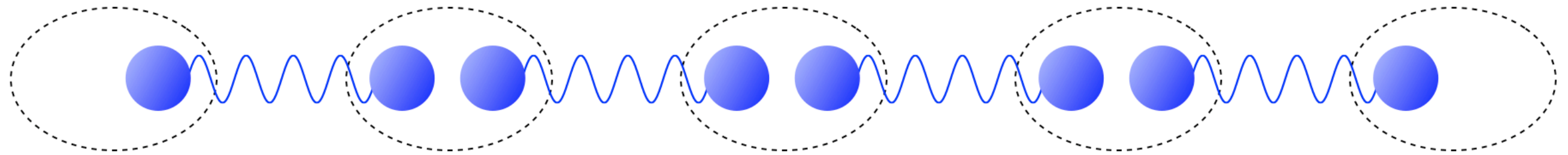
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Ground state of  $H$  should  
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...but how can we find it?

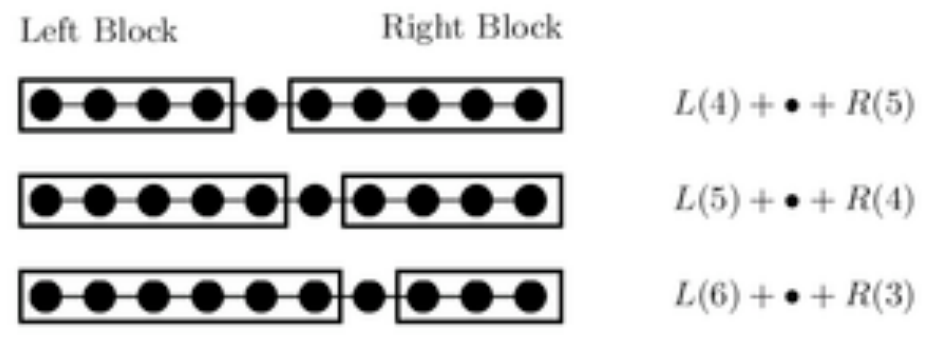
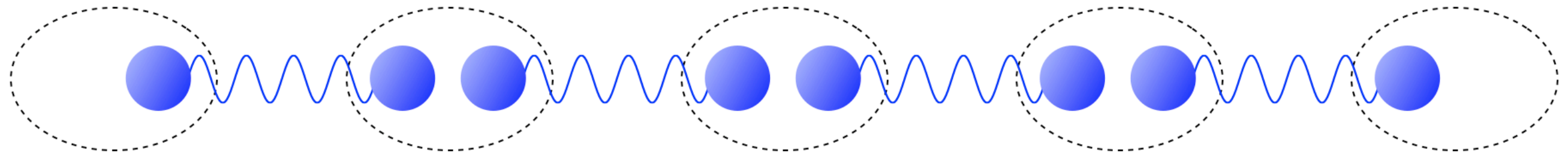
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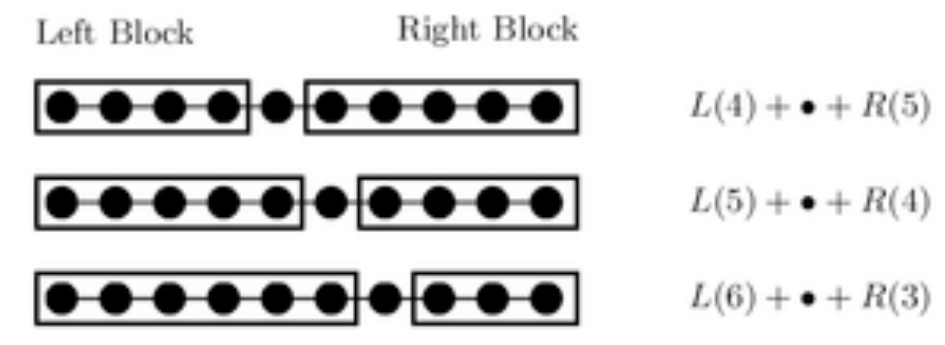
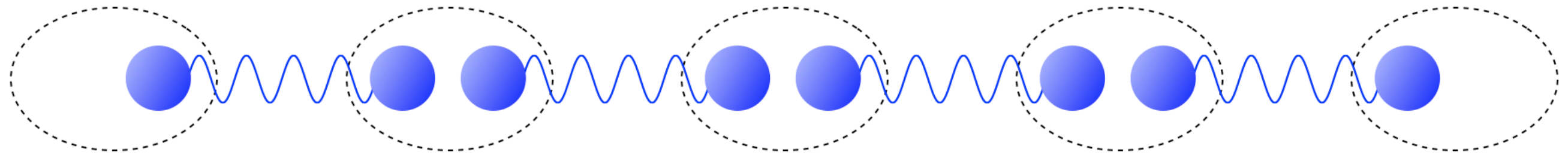
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Use DMRG!



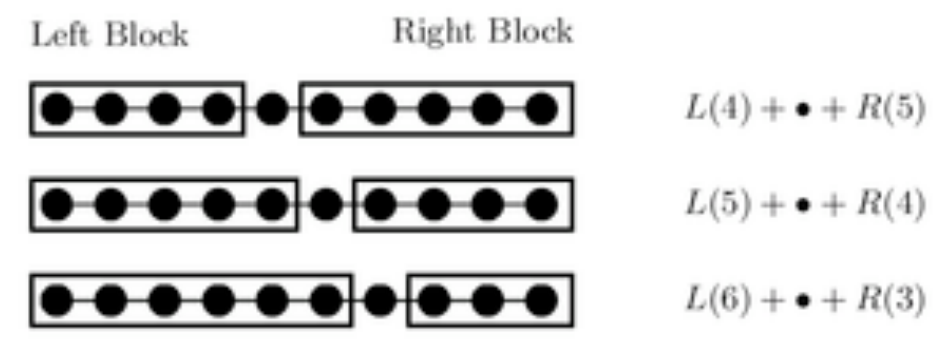
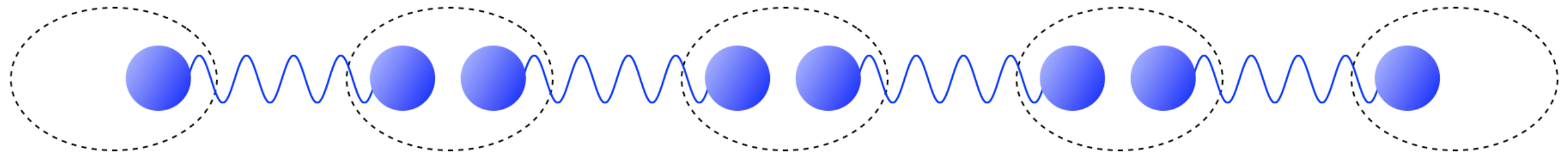
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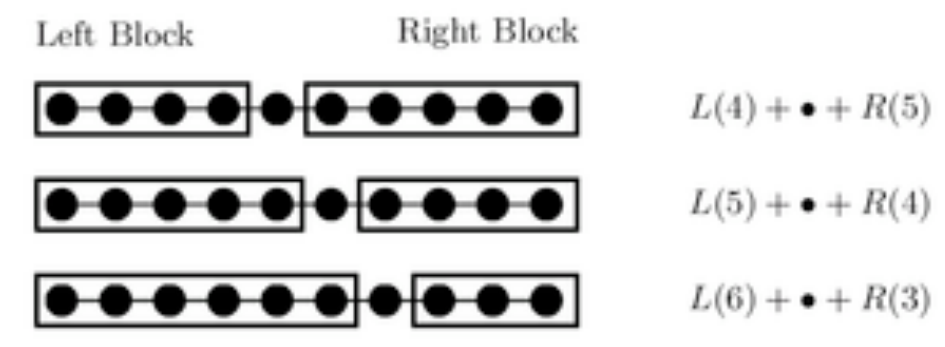
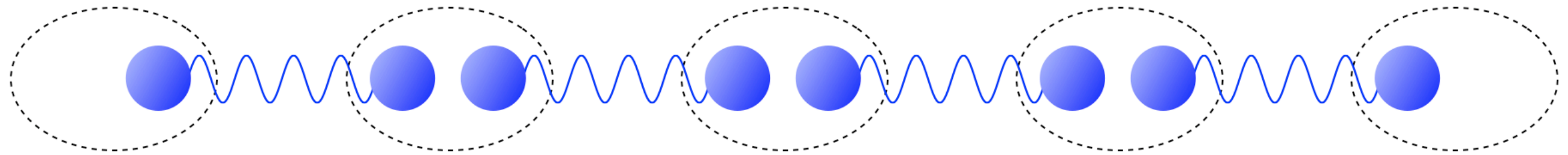


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STF, T Osborne 2009: Parent Hamiltonians can be  
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We can bound the  
fidelity now by first  
introducing a new  
error parameter

$$\xi = \max_j \text{Tr}(Q_j \sigma_j)$$

# Non-degenerate MPS

---

$$\Delta(1 - \langle \phi | \rho | \phi \rangle) \leq \sum_j \lambda_j \langle g_j | \rho | g_j \rangle = \text{Tr}(G\rho)$$

$$\text{Tr}(G\rho) \leq \sum_j [\text{Tr}(Q_j \sigma_j) + \epsilon] \leq \sum_j [\xi + \epsilon] \leq n(\xi + \epsilon)$$

$$1 - \langle \phi | \rho | \phi \rangle \leq n(\epsilon + \xi) / \Delta$$

Thus, we have a fidelity bound with an extensive error in terms of the truncation error and the statistical error of the local estimates

# Conclusions

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- ✱ Tomography of arbitrary states with high purity can be done with roughly  $O(d)$  measurements, and the scheme is fully certified and heralded, is robust to errors, and works very well in practice.
- ✱ States that are well-approximated by a matrix product state can be learned exponentially faster.
- ✱ Lots of open questions. Channels, better error bounds, better code, other types of measurements...