

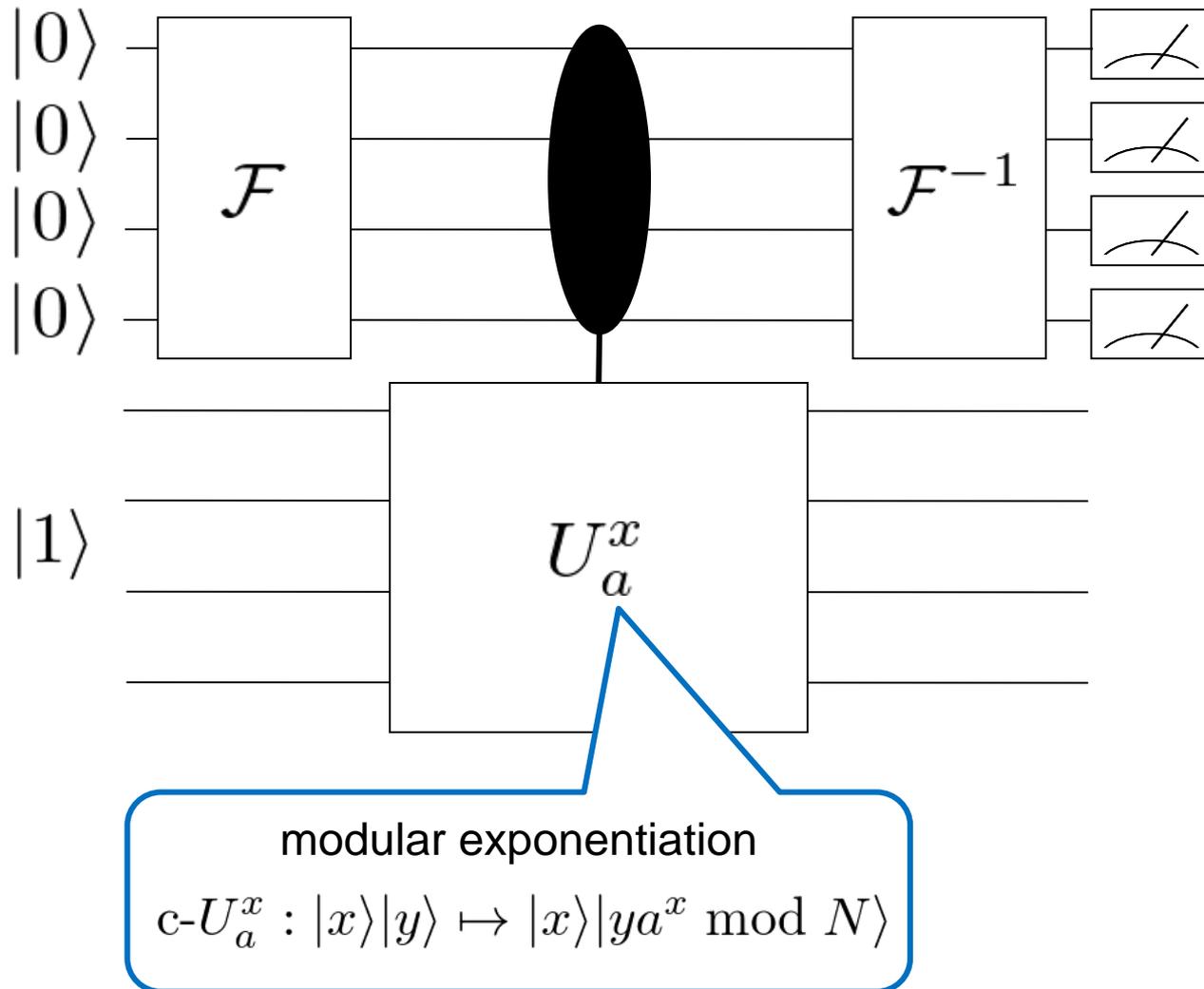
Simulating Quantum Circuits with Sparse Output Distributions

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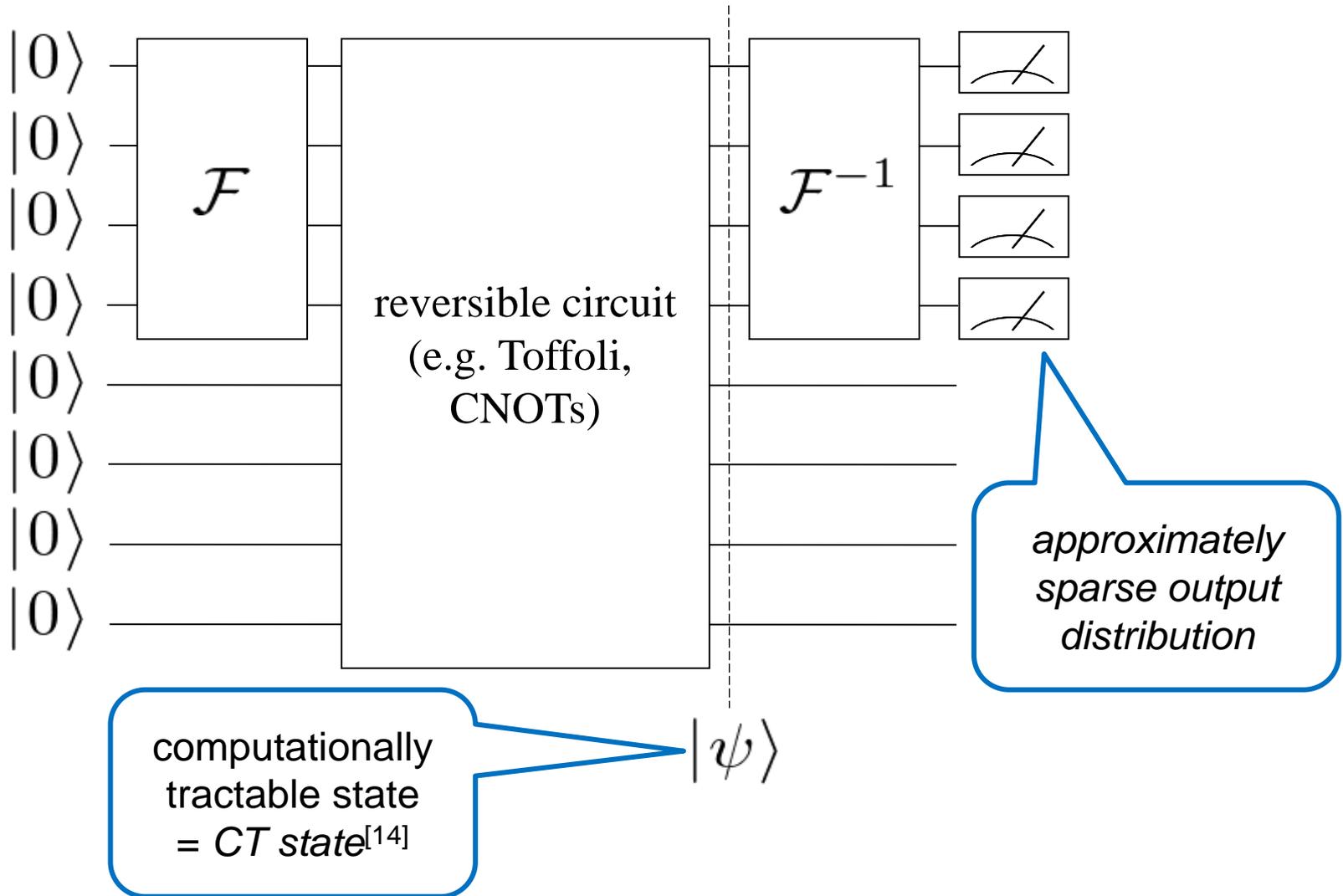
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Shor's algorithm (quantum part)



Simulating quantum circuits classically



Computationally Tractable (CT) states



Definition^[14]: A state is called *computationally tractable (CT)*, if

- (a) $p_x = |\langle x|\psi\rangle|^2$ can be sampled efficiently classically, and if
- (b) $\langle x|\psi\rangle$ can be computed efficiently (polynomial in the bit size)

CT states capture two key properties of several important families of simulable quantum states, such as

- MPS with polynomial bond dimension,
- states generated by poly-size Clifford circuits,
- nearest-neighbor matchgate circuits,
- bounded tree-width circuits,
- normalizer circuits over finite Abelian groups (acting on coset states)
- ...

Useful lemmas on CT states

CT states have the remarkable property, that overlaps CT states and expectation values of certain operators on CT states can be efficiently computed:

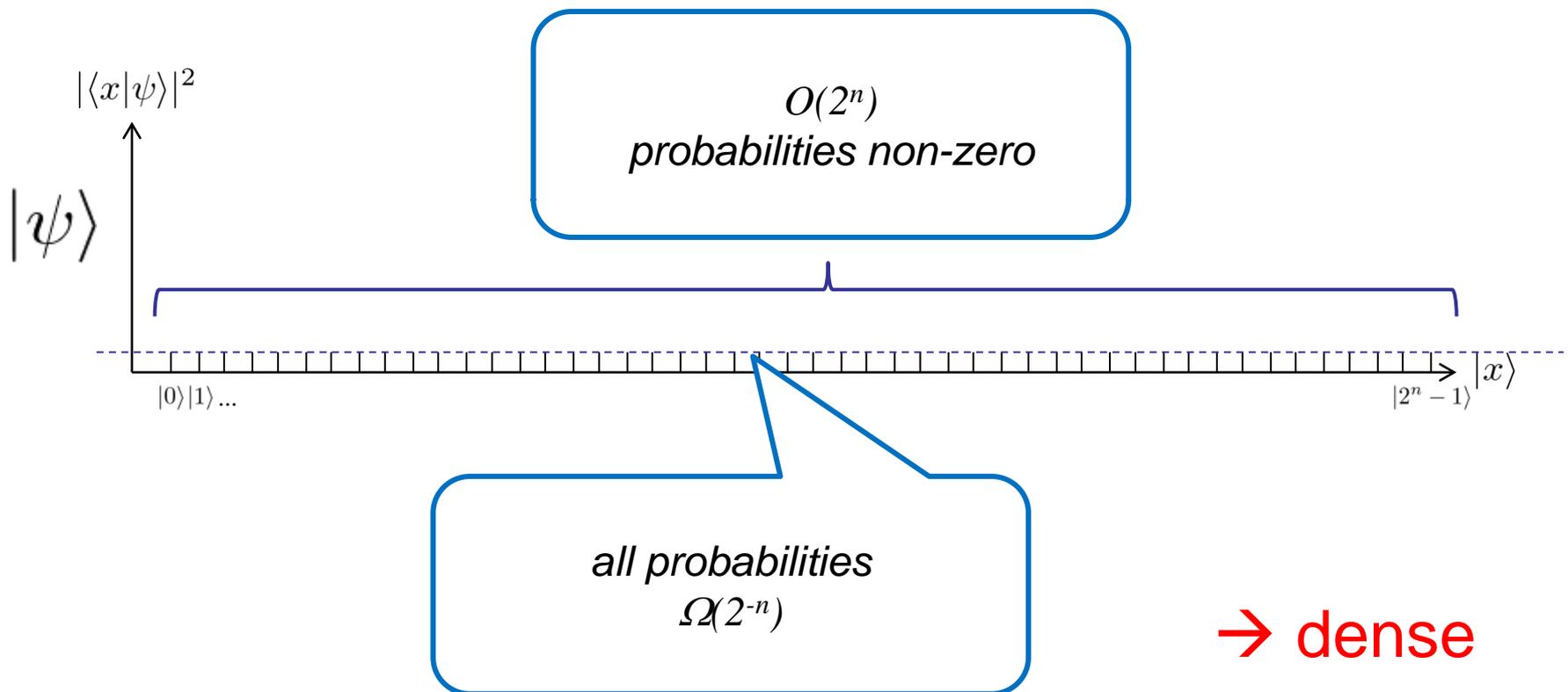
Lemma 12 ([VdN11]). *Let $|\psi\rangle$ and $|\varphi\rangle$ be CT n -qubit states and let A be an efficiently computable basis-preserving n -qubit operation. Then there exists a randomized classical algorithm with runtime $\text{poly}(n, 1/\varepsilon, \log \frac{1}{\delta})$ which outputs an approximation of $\langle\psi|A|\varphi\rangle$ with accuracy ε and success probability at least $1 - \delta$.*

Lemma 13 ([VdN11]). *Let $|\psi\rangle$ and $|\varphi\rangle$ be CT n -qubit states, let $|\xi\rangle$ and $|\chi\rangle$ be CT k -qubit states with $k \leq n$. Then there exists a randomized classical algorithm with runtime $\text{poly}(n, 1/\varepsilon, \log \frac{1}{\delta})$ which outputs an approximation of $\langle\varphi|[\langle\xi\rangle\langle\chi| \otimes \mathbb{1}]|\psi\rangle$ with accuracy ε and success probability at least $1 - \delta$.*

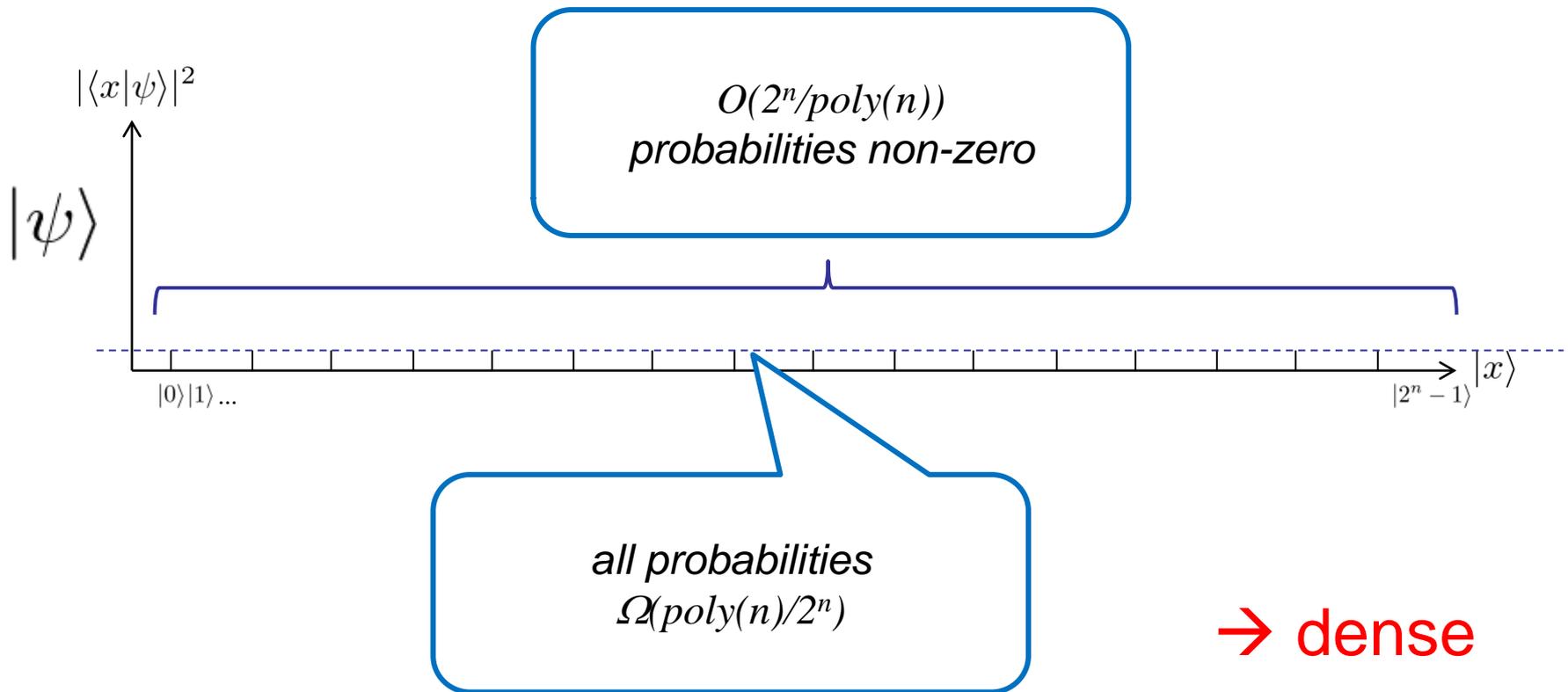
Proof: by a sampling argument using a complex-valued Chernoff-bound

Note, that this is exponentially more accurate than estimating the overlap of two explicitly given general state vectors by sampling.

Approximate sparseness



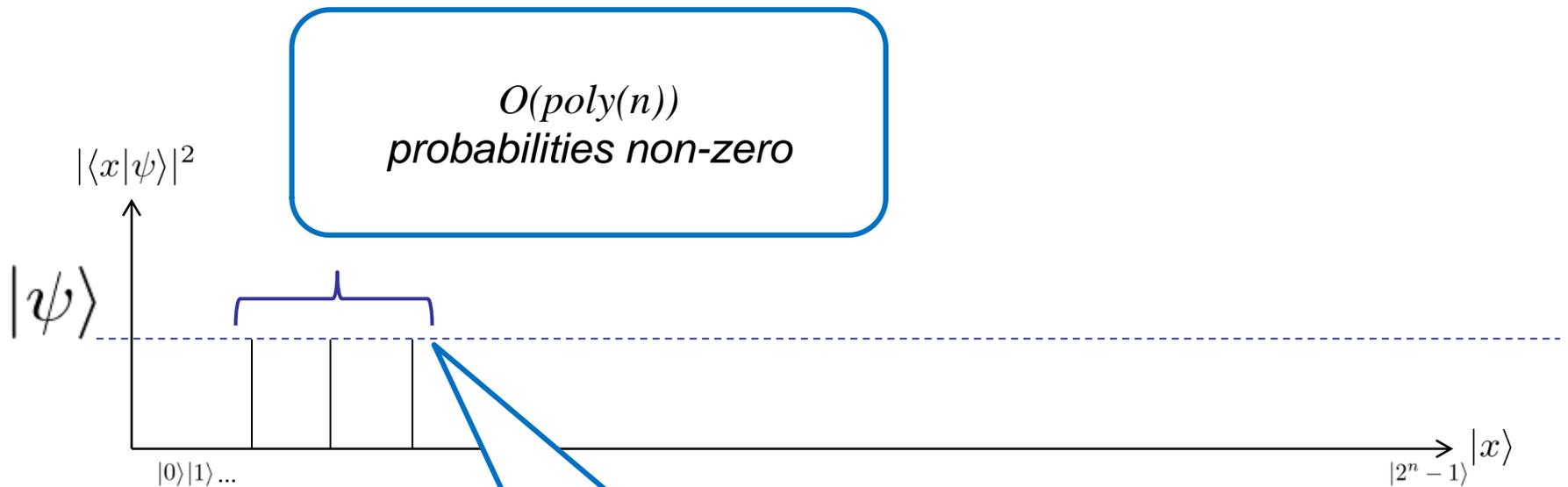
Approximate sparseness



Shor's algorithm: $\Omega(N/\log(N))$ amplitudes non-zero.

non-zero elements
cannot be identified

Approximate sparseness



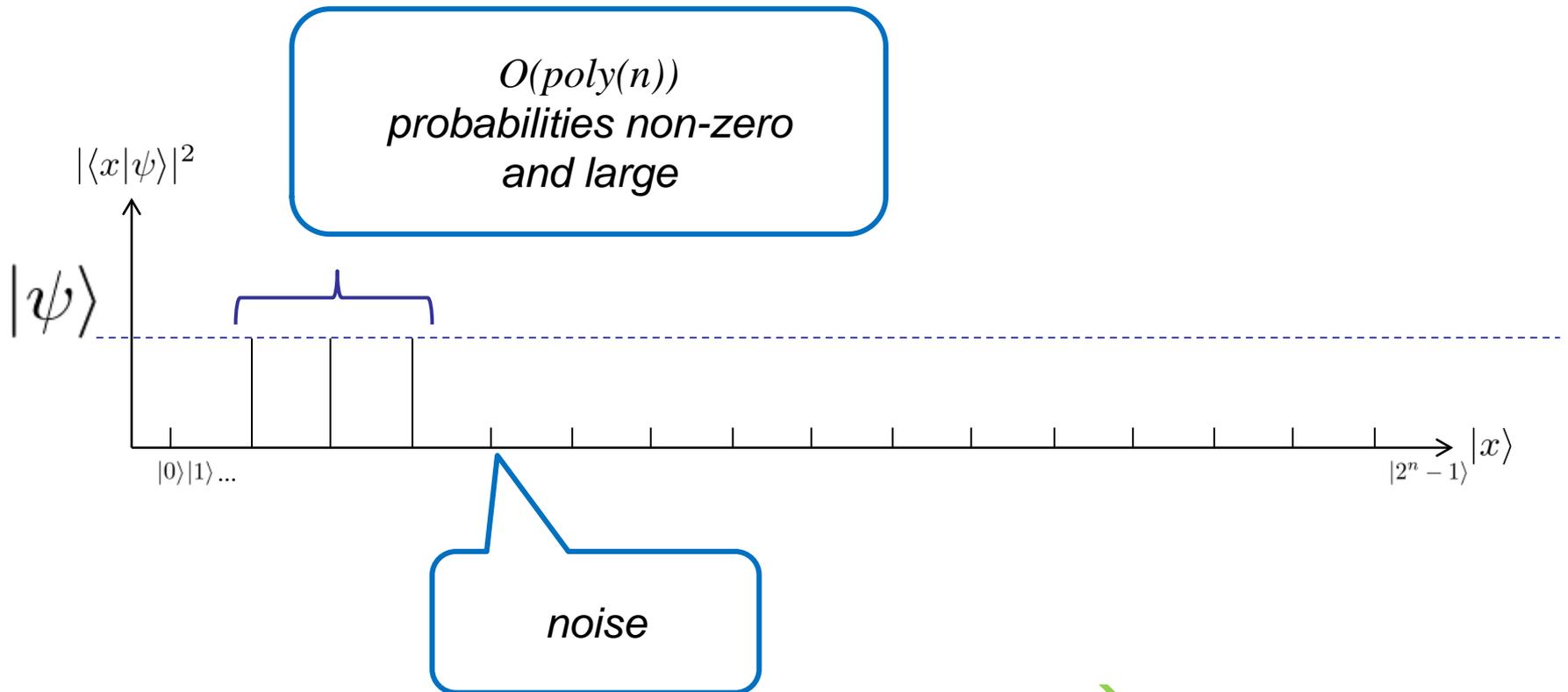
$O(\text{poly}(n))$
probabilities non-zero

all non-zero probabilities
 $\Omega(1/\text{poly}(n))$

→ sparse!

elements can be identified!
probabilities can be estimated!

Approximate sparseness



→ approx. sparse
still works with noise

Main result

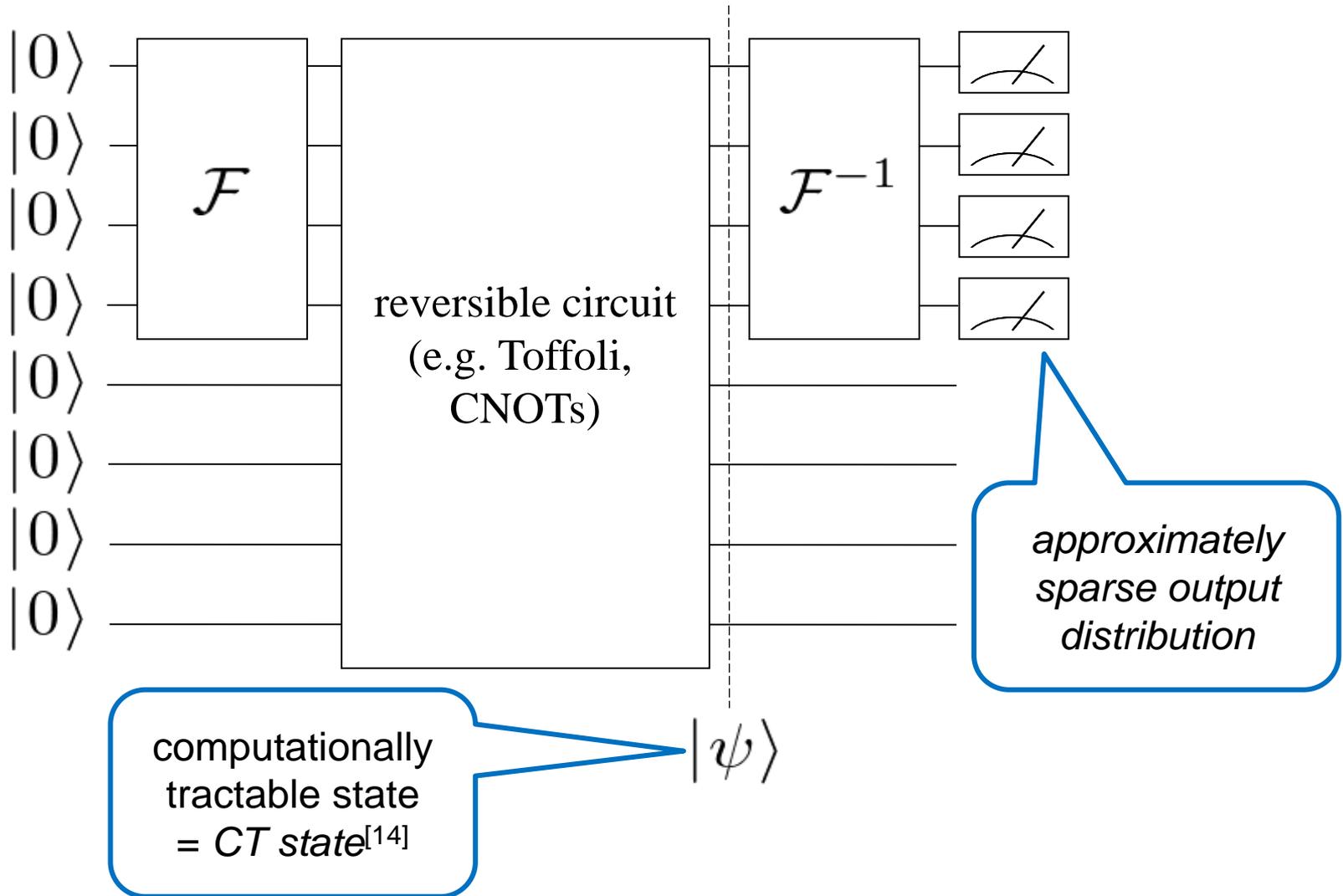
Theorem. Consider a unitary n -qubit quantum circuit composed of two blocks $C = U_2 U_1$ with input state $|\psi_{in}\rangle$. Suppose that the following conditions are fulfilled:

- (a) the state $U_1|\psi_{in}\rangle$ obtained after applying the first block is CT,
- (b1) the second block U_2 is the QFT modulo 2^n or its inverse, or
- (b2) the second block U_2 is a tensor product of unitaries $u_1 \otimes \cdots \otimes u_n$
- (c) the final state $|\psi_{out}\rangle = C|\psi_{in}\rangle$ is promised to be $\sqrt{\varepsilon}$ -approximately t -sparse for some $\varepsilon \leq 1/6$ and some t .

Then there exists a randomized classical algorithm with runtime $\text{poly}(n, t, 1/\varepsilon, \log \frac{1}{\delta})$ which outputs (by means of listing all nonzero amplitudes) an s -sparse state $|\psi\rangle$ which, with probability at least $1 - \delta$, is $O(\sqrt{\varepsilon})$ -close to $|\psi_{out}\rangle$, where $s = O(t/\varepsilon)$.

(Theorem is stated for case of amplitudes and 2-norm.
Analogous theorem is true for probabilities and 1-norm.)

Simulating quantum circuits classically



Proof sketch (QFT case)

Lemma (coefficients estimation): Given an approx.-sparse probability distribution over n bits, where *all marginal distributions* of the first m bits are efficiently sampleable. Then w.h.p. a list of the $O(\text{poly}(n))$ -many bit strings with non-zero probability can be efficiently computed and the probabilities can be efficiently estimated to $O(\text{poly}(1/\varepsilon))$ accuracy.

Proof: by a binary search / branch-and-bound argument.

The main theorem follows from the coefficient estimation lemma and the next lemma.

Main theorem is a generalization of the **Kushilevitz-Mansour** algorithm or **Goldreich-Levin** theorem to quantum states.

Proof sketch (QFT case)

Lemma (marginal distribution): The m -bit marginals of the probability distribution produced by the quantum circuit satisfying the assumptions of the main theorem are efficiently approximable.

Proof sketch:

Generalized Pauli operators:

$$\begin{aligned} X_d|x\rangle &= |x+1\rangle & \mathcal{F}_d^\dagger Z_d \mathcal{F}_d &= X_d \\ Z_d|x\rangle &= e^{\frac{2\pi i}{d}x}|x\rangle & \mathcal{F}_d Z_d \mathcal{F}_d^\dagger &= X_d^\dagger \end{aligned}$$

We want to estimate $|y_1 \cdots y_m\rangle\langle y_1 \cdots y_m| \otimes I \equiv P(y)$ on a CT state.

Note: $\hat{x} \bmod 2^m = \hat{y}$ iff $\alpha^{\hat{y}} Z^{2^{k-m}} |\hat{x}\rangle = |\hat{x}\rangle$ with $\alpha := e^{-\frac{2\pi i}{2^m}}$.

Therefore, $P(y)$ is the 1-eigenspace of $M := \alpha^{\hat{y}} Z^{2^{k-m}}$, which can be obtained by the average:

$$P(y) = \frac{1}{2^m} \sum_{u=0}^{2^m-1} M^u.$$

Proof sketch (QFT case)

Thus the marginal probability distribution of the first m qubits of the quantum circuit can be written as

$$p(y) = \langle \text{CT} | [\mathcal{F}^\dagger P(y) \mathcal{F}] \otimes I | \text{CT} \rangle$$

Using $\mathcal{F}^\dagger P(y) \mathcal{F} = \frac{1}{2^m} \sum_{u=0}^{2^m-1} N^u$ where $N := \alpha^{\hat{y}} X^{2^k-m}$ we find that

$$p(y_1 \cdots y_m) = \frac{1}{2^m} \sum_{u=0}^{2^m-1} \langle \text{CT} | N^u \otimes I | \text{CT} \rangle$$

But each term of the sum is additively approximable by the CT state lemma, thus the sum is additively approximable as well and the lemma follows.

Consequences

- The dense output distribution of Shor's algorithm (or its generalizations) is a ***necessary feature*** for the (conjectured) exponential speed-up over classical computers.
- In order to extract meaningful information out of a dense superposition, additional structure (e.g. group structure) must be present, such that $O(\text{poly}(n))$ samples suffice to efficiently identify the structure.

Thank you.