

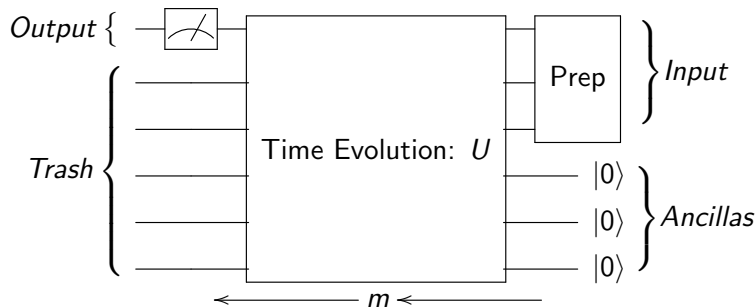
“Improved classical simulation of quantum circuits dominated by Clifford gates”

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Model of Computation



- ▶ System: input qubits + ancilla qubits = n qubits
- ▶ Hilbert space: \mathbb{C}^{2^n} - exponentially large
- ▶ Intuition: Quantum computation efficiently calculates matrix multiplication $U|\Psi_{in}\rangle$

Model of Computation: Output Probabilities

- 1 Introduction
- Model of Computation
- Gottesmann-Knill Theorem
- Magic States
- 2 Algorithm
- 3 Appendix

- ▶ Simplification: assume input state $|0^{\otimes n}\rangle$
- ▶ Measure output qubit: some probability distribution
- ▶ Goal: sample from this distribution

$$c_x = \langle x|U|0^{\otimes n}\rangle$$

$$P_x = |c_x|^2 = \langle 0^{\otimes n}|U^\dagger|x\rangle\langle x|U|0^{\otimes n}\rangle$$

- ▶ Here $|x\rangle\langle x|$ is a projector onto an output state

Classical simulation: naïve approach

- ▶ Given $U = U_m \dots U_3 U_2 U_1$, calculate P_x :

$$P_x = \langle 0^{\otimes n} | U_1^\dagger U_2^\dagger \dots U_m^\dagger | x \rangle \langle x | U_m \dots U_2 U_1 | 0^{\otimes n} \rangle$$

- ▶ Calculate m matrix multiplications in \mathbb{C}^{2^n}
- ▶ Naïve runtime: $m(2^n)^\omega$, best known¹ $\omega = 2.3737$

Interpretation

- ▶ Exponential in n : always intractable for large enough n
- ▶ Getting rid of exponentiality? Would imply:
Quantum computing = Classical computing
- ▶ Algorithm 'moves' exponent in n to other parameter

¹Coppersmith-Winograd algorithm

Stabilizer States

- ▶ D. Gottesman (1998): “The Heisenberg Representation of Quantum Computers”²
- ▶ Consider an Abelian subgroup $G \subset \mathcal{P}_n$ with $-I \notin G$.
- ▶ Def: $|\phi\rangle$ is **stabilized** by G if $P|\phi\rangle = |\phi\rangle, \forall P \in G$.
- ▶ Def: $|\phi\rangle$ is a **stabilizer state** if stabilized by some G
- ▶ Clifford gates map stabilizer states to stabilizer states

- ▶ Example: $G = \langle X \otimes X, Z \otimes Z \rangle \subset \mathcal{P}_2$
- ▶ Unique stabilizer state:

$$|\phi\rangle = \frac{|00\rangle + |11\rangle}{\sqrt{2}}$$

- ▶ Degrees of freedom: G defined by k **stabilizer generators** stabilizes 2^{n-k} states.

The Gottesmann-Knill Theorem

- ▶ Stabilizer projector: project onto space stabilized by G

$$\Pi_G = \prod_{P \in G} \frac{1 + P}{2}$$

- ▶ Clifford gates can act on stabilizer projectors: act on each generator of G
- ▶ Want to calculate:

$$P_x = \langle 0^{\otimes n} | U_1^\dagger U_2^\dagger \dots U_m^\dagger | x \rangle \langle x | U_m \dots U_2 U_1 | 0^{\otimes n} \rangle$$

- ▶ What if $U_i \in \mathcal{C}_n$? Then, given $|x\rangle\langle x| = \Pi_x$:

$$\begin{aligned} P_x &= \langle 0^{\otimes n} | U_1^\dagger U_2^\dagger \dots \Pi_{G_m(x)} \dots U_2 U_1 | 0^{\otimes n} \rangle \\ &= \langle 0^{\otimes n} | U_1^\dagger \Pi_{G_2(x)} U_1 | 0^{\otimes n} \rangle = \langle 0^{\otimes n} | \Pi_{G(x)} | 0^{\otimes n} \rangle \end{aligned}$$

- ▶ Result: can calculate circuit in polynomial time!

Clifford + T: A universal gate set

- ▶ Circuits composed of Cliffords, i.e. $H, S, CNOT$, can be simulated efficiently
- ▶ $\{H, S, CNOT\}$ **acting on stabilizer states only** is not universal
 - ▶ Mathematical fact
 - ▶ Otherwise: quantum computing = classical computing
- ▶ Add the T gate to obtain universal gate set:

$$T = \begin{bmatrix} 1 & 0 \\ 0 & e^{i\pi/4} \end{bmatrix}$$

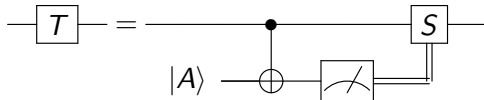
- ▶ Problem: T gate hard to simulate classically
- ▶ Incidentally: T gate also hard to build in experiment

Gadgetization with Magic States

- ▶ Consider a 'magic state':

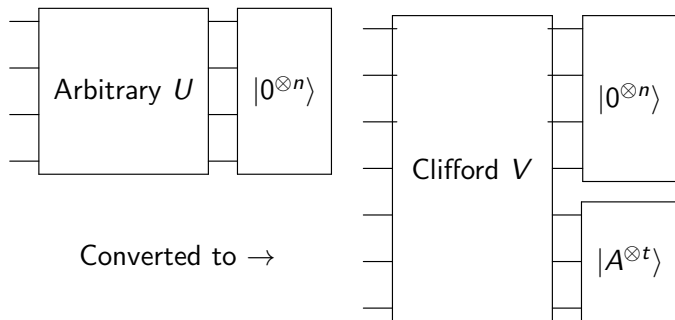
$$|A\rangle = \frac{1}{\sqrt{2}}(|0\rangle + e^{i\pi/4}|1\rangle)$$

- ▶ Use $|A\rangle$ as a resource to write T in terms of Cliffords:



- ▶ Measurement destroys magic state in the process.
- ▶ Input to circuit was $|0^{\otimes n}\rangle$, now is $|0^{\otimes n}A^{\otimes t}\rangle$.
- ▶ t = number of T gates in circuit

Converting the problem



The challenge

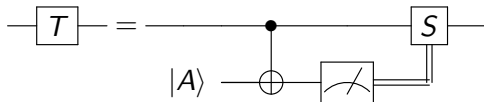
- ▶ Before: non-Clifford circuit with T gates
- ▶ After: non-stabilizer 'magic' resource state $|A^{\otimes t}\rangle$

Algorithm Goal

- ▶ Goal: Sample from probability distribution

$$P_x = \langle 0^{\otimes n} | U^\dagger \Pi_x U | 0^{\otimes n} \rangle$$

- ▶ Gadgetize non-Clifford unitary U to Clifford V with 'magic' resource state $|A^{\otimes t}\rangle$



- ▶ How to deal with measurement? Post-select measurement outcomes into string y . Calculate:

$$P_x = \frac{\langle A^{\otimes t} 0^{\otimes n} | V^\dagger (\Pi_x \otimes \Pi_y) V | 0^{\otimes n} A^{\otimes t} \rangle}{\langle A^{\otimes t} 0^{\otimes n} | V^\dagger (\mathbb{I} \otimes \Pi_y) V | 0^{\otimes n} A^{\otimes t} \rangle}$$

- ▶ Works for any y !

Concept: Stabilizer Rank χ

- ▶ Remaining problem: non-stabilizer state $|A^{\otimes t}\rangle$
- ▶ Write as a linear combination of χ stabilizer states $|\phi_a\rangle$

$$|A^{\otimes t}\rangle \approx \sum_a^{\chi} z_a |\phi_a\rangle = |\Psi\rangle$$

- ▶ 2^t stabilizer states: Naïve upper bound $\chi \leq 2^t$
- ▶ Clever trick 1: Recognize that $|A^{\otimes 2}\rangle$ is a sum of two stabilizer states. Divide $|A^{\otimes t}\rangle$ into pairs: $\chi \leq 2^{t/2}$
- ▶ Clever trick 2 (see appendix): Achieve $\chi \sim O(2^{0.23t})$. Authors conjecture that this is optimal.

Summary Calculation

$$\begin{aligned} P_x &= \langle 0^{\otimes n} | U^\dagger \Pi_x U | 0^{\otimes n} \rangle \\ &= \frac{\langle A^{\otimes t} 0^{\otimes n} | V^\dagger (\Pi_x \otimes \Pi_y) V | 0^{\otimes n} A^{\otimes t} \rangle}{\langle A^{\otimes t} 0^{\otimes n} | V^\dagger (\mathbb{I} \otimes \Pi_y) V | 0^{\otimes n} A^{\otimes t} \rangle} \\ &= \frac{\langle A^{\otimes t} 0^{\otimes n} | \Pi_{G(x,y)} | 0^{\otimes n} A^{\otimes t} \rangle}{\langle A^{\otimes t} 0^{\otimes n} | \Pi_{H(y)} | 0^{\otimes n} A^{\otimes t} \rangle} = \frac{1}{2^u} \frac{\langle A^{\otimes t} | \Pi_{\bar{G}(x,y)} | A^{\otimes t} \rangle}{\langle \bar{A}^{\otimes t} | \Pi_{H(y)} | A^{\otimes t} \rangle} \\ &= \frac{1}{2^u} \frac{|\langle \Pi_{\bar{G}(x,y)} | A^{\otimes t} \rangle|^2}{|\langle \Pi_{\bar{H}(y)} | A^{\otimes t} \rangle|^2} \approx \frac{1}{2^u} \frac{|\langle \Pi_{\bar{G}(x,y)} | \Psi \rangle|^2}{|\langle \Pi_{\bar{H}(y)} | \Psi \rangle|^2} \end{aligned}$$

- ▶ Calculation boils down to $|\langle \Pi_{\bar{G}(x,y)} | \Psi \rangle|^2$ and $|\langle \Pi_{\bar{H}(y)} | \Psi \rangle|^2$
- ▶ Approx. requires random y , rather than arbitrary y

The Algorithm

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1. Choose random y , evaluate projectors $\Pi_{\bar{G}(x,y)}$, $\Pi_{\bar{H}(y)}$
2. Compute $|A^{\otimes t}\rangle \approx \sum_a^\chi z_a |\phi_a\rangle = |\Psi\rangle$
such that $|\langle A^{\otimes t} | \Psi \rangle|^2 \geq 1 - \delta$,
where $|\phi_a\rangle$ are stabilizer states, $\chi = O(2^{0.23t} \delta^{-1})$
3. Evaluate inner products $|\Pi_{\bar{G}(x,y)} |\Psi\rangle|^2$ and $|\Pi_{\bar{H}(y)} |\Psi\rangle|^2$

$$|\Pi |\Psi\rangle|^2 = \left| \Pi \sum_a^\chi z_a |\phi_a\rangle \right|^2 = \left| \sum_a^\chi z_a \Pi |\phi_a\rangle \right|^2$$

4. Compute distribution $P_{x=0}$, $P_{x=1} = 1 - P_{x=0}$, and sample from distribution.

- ▶ Sample from output distribution for a string x :

$$\text{poly}(n, m) + 2^{0.23t} t^3 w^4$$

- ▶ Exponential: number of T gates t
- ▶ Polynomial: n qubits, width m circuit
- ▶ Length of output string $|x| = w$
- ▶ Projector Π_x has 2^w generators: w^4 via trick (appendix)
- ▶ Exponential part is highly parallelizable
 - ▶ Each term in $\sum_a^x z_a \Pi|\phi_a\rangle$ can be calculated independently

Conclusions, next steps

Implementation

- ▶ MATLAB implementation by Bravyi, Gosset
 - ▶ Hidden shift algorithm on a laptop
 - ▶ 40 qubits, 50 T gates
- ▶ Python+C implementation by Iskren Vankov, me
- ▶ Upcoming: CUDA implementation?

New concept: Stabilizer Rank

- ▶ How to decompose arbitrary $|\Psi\rangle$ into stabilizer states?
- ▶ Improve naïve runtime $O(m2^n)$ to $O(m2^{\alpha n})$ for $\alpha < 1$?

Appendix

- ▶ Alexei Kitaev's Stabilizer Toolkit
- ▶ Sampling larger bitstrings x
- ▶ Stabilizer decomposition of $|A^{\otimes t}\rangle$
- ▶ Computing inner products in $O(\chi)$ rather than $O(\chi^2)$


Alexei Kitaev's Stabilizer Toolkit

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- ▶ Traditional representation: $G \subset \mathcal{P}_n$
- ▶ Efficient representation:³
 - ▶ Affine space \mathcal{K} : Subspace of \mathbb{F}^2 such that $\mathcal{L}(\mathcal{K}) = h \oplus \mathcal{K}$
 - ▶ Quadratic form q : Function $q : \mathcal{K} \rightarrow \mathbb{Z}_8$ with properties

$$|\mathcal{K}, q\rangle = 2^{-k/2} \sum_{x \in \mathcal{K}} e^{\frac{i\pi}{4} q(x)} |x\rangle$$

- ▶ Algorithms:
 - ▶ Inner product of two states in $O(n^3)$
 - ▶ Measure a Pauli operator in $O(n^2)$
 - ▶ Sample random stabilizer states in $O(n^2)$ on average ($O(n^3)$ worst case)

³H. J. Garca and I. L. Markov, "Hybrid Techniques for Simulating Quantum Circuits using the Heisenberg Representation" 

Sampling larger bitstrings x

- ▶ Projector Π_x has 2^w generators. Contributes to $|\Pi|\Psi\rangle|^2$.
- ▶ Achieve polynomial time: Sample first bit of x , x_1 , then evaluate conditional probability for next bit, etc:

$$x = x_1 x_2 \dots x_w$$

$$P(x_2|x_1) = \frac{P(x_1, x_2)}{P(x_1)} \rightarrow P(x_3|x_1 x_2) = \frac{P(x_1, x_2, x_3)}{P(x_1, x_2)} \dots$$

- ▶ w^4 : not particularly fast, but at least not exponential

Stabilizer decomposition of $|A^{\otimes t}\rangle$

$$|A^{\otimes t}\rangle \approx \sum_a^\chi z_a |\phi_a\rangle = |\Psi\rangle$$

$$|\langle A^{\otimes t} | \Psi \rangle|^2 \geq 1 - \delta$$

- ▶ Low stabilizer rank decomposition $\chi = O(2^{0.23t} \delta^{-1})$

$$|A\rangle = e^{i\pi/8} HS^\dagger \left(\cos \frac{\pi}{8} |0\rangle + \sin \frac{\pi}{8} |1\rangle \right) = e^{i\pi/8} HS^\dagger |H\rangle$$

- ▶ Find $|\mathcal{L}\rangle$ such that $|\langle H^{\otimes t} | \mathcal{L} \rangle|^2 \geq 1 - \delta$
- ▶ Choose a dimension k such that $4 \geq 2^k v^{2t} \delta \geq 2$,
 $v = \cos \frac{\pi}{8}$, $|H\rangle = \frac{1}{2v} (|0\rangle + |+\rangle)$
- ▶ Sample a random subspace $\mathcal{L} \subset \mathbb{F}_2^n$ with dimension k .
From Markov's inequality:

$$\Pr [|\langle H^{\otimes t} | \mathcal{L} \rangle|^2 \geq 1 - \delta] \geq \Omega(\delta)$$

- ▶ Keep sampling \mathcal{L} until this is true. Should take $O(\delta^{-1})$.

Computing inner products in $O(\chi)$ rather than $O(\chi^2)$

- ▶ Given $\Pi|\Psi\rangle = \sum_a^\chi z_a \Pi|\phi_a\rangle$, compute $|\Pi|\Psi\rangle|^2$
- ▶ Naïve method: $O(\chi^2)$ calculations of $z_a z_b \langle\phi_a|\Pi|\phi_b\rangle$
- ▶ With clever trick: $O(\chi)$ calculations of $z_a \langle\theta_i|\Pi|\phi_a\rangle$ with $L = 1/p_f \epsilon$ random states $|\theta_i\rangle$

$$\text{Random variable } \alpha = \frac{2^t}{L} \sum_{i=1}^L |\langle\theta_i|\Pi|\Psi\rangle|^2$$

$$\Pr [(1 - \epsilon)|\Pi|\Psi\rangle|^2 \leq \alpha \leq (1 + \epsilon)|\Pi|\Psi\rangle|^2] \geq 1 - p_f$$

- ▶ Derivation uses that stabilizer states \mathcal{S} are a 2-design

$$\sum_{\theta \in \mathcal{S}} (|\theta\rangle\langle\theta| \otimes |\theta\rangle\langle\theta|) = \int_{\text{Haar}} (|\phi\rangle\langle\phi| \otimes |\phi\rangle\langle\phi|) d\phi$$