Supercharging classical simulation of quantum circuits with the ZX-calculus

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Quantum circuit simulation

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- Tensor-network based methods:
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 - edge cutting
 - ▶ ...
- Stabiliser-decomposition based methods:
 - Stabiliser extent
 - ► Stabiliser rank ← **This talk**

Stabiliser rank

Gottesman-Knill Theorem

A quantum computation consisting of

- stabiliser states,
- Clifford unitaries,
- stabiliser measurements,

can be efficiently classically simulated.

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Observation: Stabiliser states span entire space of states. So:

$$|\psi
angle = \sum_{j=1}^\chi \lambda_j |\psi_j
angle$$
 where $|\psi_j
angle$ is stabiliser.

 χ is called **stabiliser rank** of $|\psi\rangle$.

Simulating using stabiliser rank

- Start with Clifford+T circuit.
- Write each T gate as magic state injection.
- Decompose T states into sum of stabilisers.
- Efficiently simulate resulting Clifford circuits.
- Add results together.
- We're done!

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What's the catch? Stabiliser rank of k T states scales exponentially with k. ... But it's not just 2^k terms. We can do better!

 $\begin{array}{l} \operatorname{Recall} |T\rangle \propto |0\rangle + e^{i\pi/4} |1\rangle. \\ \operatorname{So} \chi(|T\rangle) = 2 \text{ and hence } \chi(|T\rangle^{\otimes k}) \leq 2^k. \end{array}$

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$$\begin{aligned} |T\rangle^{\otimes 2} &\propto |00\rangle + e^{i\pi/4}|01\rangle + e^{i\pi/4}|10\rangle + e^{i\pi/2}|11\rangle \\ &= (|00\rangle + i|11\rangle) + e^{i\pi/4}(|01\rangle + |10\rangle) \end{aligned}$$

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Hence $\chi(|\mathcal{T}
angle^{\otimes 2})=2$ so that

$$\chi(|T\rangle^{\otimes k}) = \chi((|T\rangle^{\otimes 2})^{\otimes k/2}) \le 2^{k/2} = 2^{0.5k}.$$

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Turns out $\chi(|T\rangle^{\otimes 6}) \leq 7$ so $\chi(|T\rangle^{\otimes k}) \leq 2^{\alpha k}$ where $\alpha = \log_2(7)/6 \approx 0.467$. Found by Bravyi, Smith, Smolin (BSS) in 2016.

Our idea

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Described in two papers:

- Kissinger & vdW. Simulating quantum circuits with ZX-calculus reduced stabiliser decompositions. arXiv:2109.01076
- Kissinger, Vilmart & vdW. Classical simulation of quantum circuits with partial and graphical stabiliser decompositions. arXiv:2202.09202

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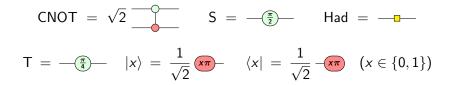
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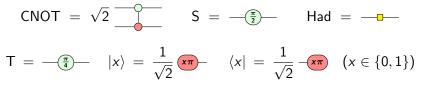
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Let's go through these steps in more detail.

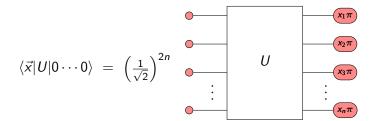
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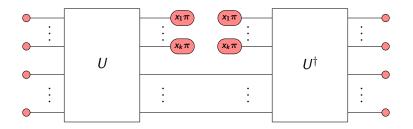


Calculating single amplitude:



Marginal probabilities

To calculate marginal probability, use *doubling* technique:



Strong simulation vs Weak simulation

Weak sim: approx. sample from the same output distribution. Strong sim: approx. calculate any marginal probability.

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We are doing *exact* strong simulation here.

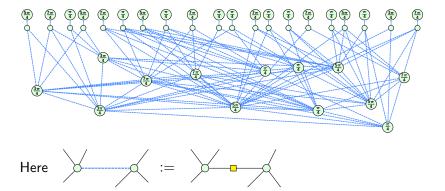
- ► Writing circuit as ZX-diagram √
- Optimising ZX-diagram
- Decomposing magic states

Simplifying ZX-diagrams

We use strategy from our previous paper *Reducing T-count with the ZX-calculus.*

Simplifying ZX-diagrams

We use strategy from our previous paper *Reducing T-count with the ZX-calculus.* This transforms every ZX-diagram into something like this:



Properties of reduced diagram

The important part:

Every spider carries a non-Clifford phase,



Properties of reduced diagram

The important part:

Every spider carries a non-Clifford phase,

• or is part of a *phase gadget*: α ---

Particularly: if original circuit had k T gates, resulting diagram has $\leq 2k$ spiders (regardless of #qubits or #gates).

- ► Writing circuit as ZX-diagram √
- ▶ Optimising ZX-diagram ✓
- Decomposing magic states

Decomposing T-like spiders

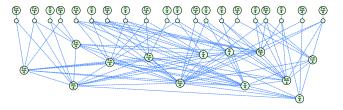
The 6-to-7 magic state decomposition in ZX is:

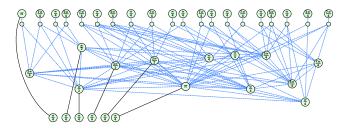
$$e^{i\pi/4} \stackrel{||}{\underset{\overline{a}}{\overline{a}}} \stackrel{||}{\underset{\overline{a}}{\overline{a}}} \stackrel{||}{\underset{\overline{a}}{\overline{a}}} \stackrel{||}{\underset{\overline{a}}{\overline{a}}} \stackrel{||}{\underset{\overline{a}}{\overline{a}}} = +2e^{i\pi/4}$$

$$-\frac{1+\sqrt{2}}{4} \stackrel{||}{\underset{\overline{a}}{\overline{b}}} \stackrel{||}{\underset{\overline{a}}{\overline{a}}} \stackrel{||}{\underset{\overline{a}}{\overline{b}}} \stackrel{||}{\underset{\overline{a}}{\overline{b}}} \stackrel{||}{\underset{\overline{a}}{\overline{b}}} \stackrel{||}{\underset{\overline{a}}{\overline{b}}} \stackrel{||}{\underset{\overline{a}}{\overline{b}}} \stackrel{||}{\underset{\overline{a}}{\overline{a}}} \stackrel{|}{\underset{\overline{a}}{\overline{a}}} \stackrel{||}{\underset{\overline{a}}{\overline{a}}} \stackrel{||}{\underset{\overline{a}}} \stackrel{|}{\underset{\overline{a}}} \stackrel{|}{\underset{\overline{a}$$

Applying the decomposition

We pick some spiders to decompose and unfuse the phases:





And now we can apply the magic state decomposition.

Full algorithm

- After first decomposition we have 7 diagrams.
- Repeat whole procedure for each of these diagrams, end up with 7 * 7 diagrams.
- Repeat until there aren't 6 magic states to decompose.
- Diagrams are then small enough to calculate value of directly.
- Sum values to get the amplitude/prob we were calculating.

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Better decompositions

Improved upper bounds on the stabilizer rank of magic states

Hammam Qassim *^{†‡ §} Hakop Pashayan *^{¶ |} David Gosset *[¶] June 16, 2021

 \Rightarrow improved stabiliser decompositions, including 6-to-6 decomp (giving $\alpha \approx 0.431$ instead of $\alpha \approx 0.467$), and other decomps giving $\alpha < 0.40$.

Cat states

Qassim et al. uses cat states:

$$|\operatorname{cat}_n\rangle := \frac{1}{\sqrt{2}} (\mathbb{I}^{\otimes n} + Z^{\otimes n}) |T\rangle^{\otimes n} = \frac{1}{\sqrt{2}^{n+1}} \left(\overline{\mathbb{I}}_{4}^{\otimes n} - \overline{\mathbb{I}}_{4}^{\otimes n} + \overline{\mathbb{I}}_{4}^{\otimes n} - \overline{\mathbb{I}}_{4}^{\otimes n} \right)$$

Cat states

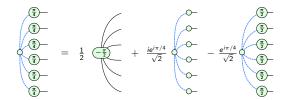
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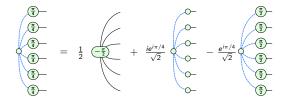
These actually have nice representation in ZX:

$$|\operatorname{cat}_n\rangle = \frac{1}{\sqrt{2}} \phi^{\left(\begin{array}{c} \widehat{\mathfrak{s}} \\ \widehat{\mathfrak{s}} \end{array}\right)}$$

They find good decomp of $|cat_6\rangle$:

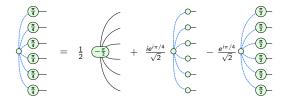


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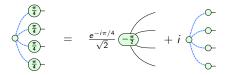


So, if we only have $|cat_6\rangle$ diagrams, we would have $\alpha \approx 0.264$.

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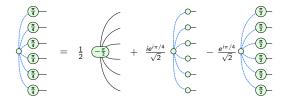


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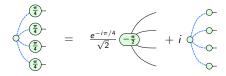


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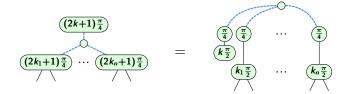
So, if we only have $|cat_6\rangle$ diagrams, we would have $\alpha \approx 0.264$. But we can do even better!



This would give $\alpha = 0.25$. Using these we get good decompositions for $|cat_k\rangle$ with $k \le 6$.

Phase gadgets are cat states

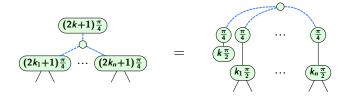
We find cat states as phase gadgets:



So *n*-legged phase gadget is $|cat_{n+1}\rangle$ state.

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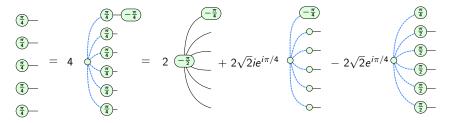
So *n*-legged phase gadget is $|cat_{n+1}\rangle$ state.

So as long as there are phase gadgets with \leq 5 legs, we can use these decompositions.

Partial stabiliser decomp

But what if there are no phase gadgets?

Then we can do the following 'partial' decomp:



This trades 5 magic states for 3 terms with 1 magic per term. So effectively removes 4 magic states. This is then a 4-to-3 decomp: $\alpha \approx 0.396$. We are hence looking for the following things to decompose:

- 1. a phase gadget with 3 legs ($\alpha =$ 0.25),
- 2. a phase gadget with 5 legs ($\alpha \approx$ 0.264),
- 3. a phase gadget with 4 legs (lpha pprox 0.317),
- 4. a phase gadget with 2 legs ($\alpha = 1/3 \approx$ 0.333),
- 5. any 5 T-spiders ($\alpha \approx 0.396$).

So how well does all this work?

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- There are $O(2^{\alpha k})$ diagrams, so total cost is $O(2^{\alpha k}k^2)$.
- (Bravyi et al., 2016) gave $O(2^{\alpha k}k^3)$.
- Benefit comes from preventing 'double work': we 'partially evaluate' the stabilisers by simplifying the diagrams.

Actual benefit

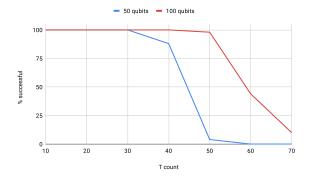
We benchmarked our method on two families of circuits:

- 50- and 100-qubit random Clifford+T circuits built out of Pauli exponentials.
- 50-qubit hidden-shift circuits (type of CCZ circuit).

We are sampling from the output distribution (using strong simulation).

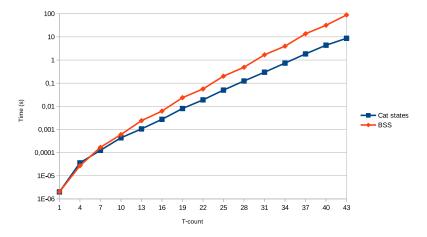
Code is implemented in quizx, a Rust port of PyZX.

Benchmark: Clifford+T



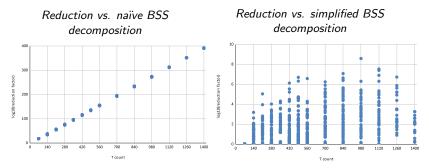
Percentage of random 50- and 100-qubit circuits of a given T-count that were successfully sampled in under 5 minutes. For each T-count 50 random circuits were generated.

Benchmark: Clifford+T cat-decomp comparison



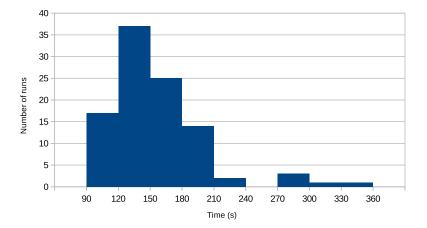
Runtime of random 20-qubit Clifford+T circuit simulations (avg of 10 runs per T-count).

Benchmark: hidden-shift circuit term reduction



Reduction in term count on 50-qubit hidden shift circuits vs. naïve BSS decomposition (left) and BSS decomposition after single ZX-simplification (right).

Benchmark: hidden-shift 50-qubit simulation time



The time distribution of simulating 100 random 50-qubit hidden-shift circuits with T-count 1400 using our new decompositions.

Conclusions

- Using ZX we can greatly speed-up stabiliser rank simulations.
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- It allows us to use better decompositions for substructures of diagrams, and to introduce partial stabiliser decompositions.

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Moral of the story: optimisation and simulation are not separate. They are two sides of the same coin.

Conclusions

- Using ZX we can greatly speed-up stabiliser rank simulations.
- Especially for structured circuits.
- It allows us to use better decompositions for substructures of diagrams, and to introduce partial stabiliser decompositions.
 Moral of the story: optimisation and simulation are not separate.
 They are two sides of the same coin.

Future work:

- Use more diagram optimisations.
- Find heuristics for picking good spiders to decompose.
- Approximate simulation and better weak simulation.
- stabiliser *extent* methods.
- Use quantum measurement w/o computing marginals technique.

Thank you for your attention!

Further reading:

- Kissinger & vdW. Simulating quantum circuits with ZX-calculus reduced stabiliser decompositions. arXiv: 2109.01076
- Kissinger, Vilmart & vdW. Classical simulation of quantum circuits with partial and graphical stabiliser decompositions. arXiv: 2202.09202
- Qassim, Pashayan, Gosset. Improved upper bounds on the stabilizer rank of magic states. arXiv: 2106.07740
- Bravyi, Gosset. Improved classical simulation of quantum circuits dominated by Clifford gates. arXiv: 1601.07601