



Exhaustive Search for Quantum Circuit Optimization using ZX Calculus

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Quantum Computing Applications

- Factorization
 - Shor [21] $\mathcal{O}(\log N^3)$ vs. GNFS [17]
 $\mathcal{O}(\exp \sqrt{\frac{64}{9}} \log N^{\frac{1}{3}} \log \log N^{\frac{2}{3}})$
- Unstructured search
 - Grover [8] $\mathcal{O}(\sqrt{N})$ vs. linear search [12] $\mathcal{O}(N)$
- Simulation of quantum systems
 - Molecular interaction [1]
- Quantum artificial intelligence
 - Perovskite structure prediction [16]
 - Climate modelling [25]

⇒ **Near exponential speedup for certain applications**



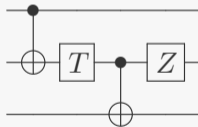
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Quantum Circuits [18]

- Analogous to classical logic gates
- But **reversible**
- Input is reconstructable from output
- Gates are unitary operators
- Not all gates have classical counter part (eg. Hadamard)



Quantum Computing Applications

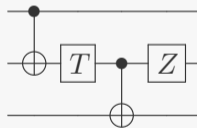
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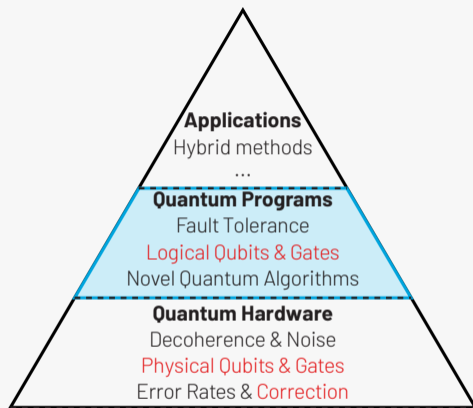
- Analogous to classical logic gates
- But **reversible**
- Input is reconstructable from output
- Composed of unitary operators
- Some gates have classical counterpart (eg. Hadamard)

What is the catch?



Quantum Computing

- **Ressource restrictions:**
 - 127 logical qubits [6]
 - up to ≈ 5000 logical gates
 - short coherence time ($80[\mu\text{s}]$ to $1[\text{ms}]$) [22]
- **Error correction:**
 - noise drives gate error rate [24]
 - limits the number of usable gates
 - overhead varies by an order of magnitude for different gates [20]
- **Quantum computing limited to artificial problems**



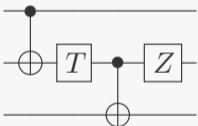
⇒ **Architecture-independent QC optimization**

Quantum Computing

Current Challenges in Quantum Computing

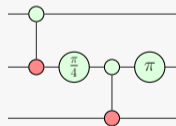
QC Optimization

- Infinite universal gate sets
- Infinite gate commutation rules
- Equivalence verification computational expensive



ZX Calculus [2, 3]

- 8 generators
- 9 well defined rewriting rules
- Rewriting rules preserve semantics



⇒ **State-space for combinatorial optimization is (infinitely) large**

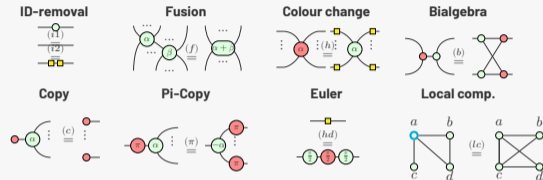
Contribution

- **Formalization** of ZX diagram optimization
- New set of **pruning conditions** for state-space reduction
- Reproducible **framework** that integrates in standard quantum compilation pipelines (≈ 7000 LOC)

ZX calculus

Diagrammatic Reasoning Framework

- Every QC can be expressed as a ZX diagram [26]
- **Semantic preserving rewriting rules**
- Circuit extraction is # P-hard [4]
- Applied for QC optimization and verification

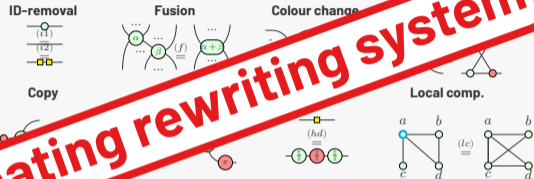


ID	Z	Z-Phase	T	X	X-Phase	H	CNOT

ZX calculus

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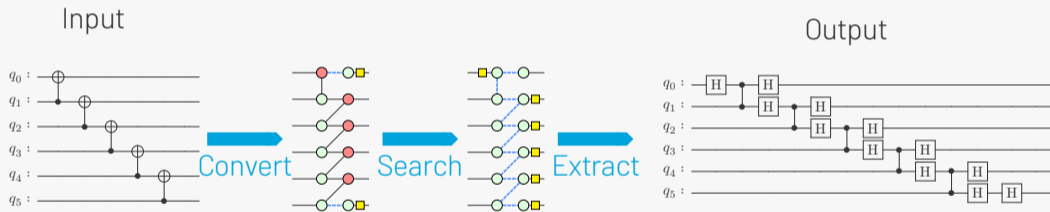
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ZX Calculus is a non-terminating rewriting system

ID	Z	Z-Phase	X	X-Phase	H	CNOT

Optimization Pipeline



Formal ZX diagram optimization

Quantum Circuit Optimization

- Input & output quantum circuit
 - Let **QC** be the set of quantum circuits
- *opt* is a function that maps a quantum circuit to a metric
 - $opt : \mathbf{QC} \rightarrow \mathbb{Z}$

Define quantum circuits

Formal ZX diagram optimization

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Define quantum circuits

- Linear map of qubits
 - Let **LM** be the set of the linear map of qubits
- γ is a function that maps a quantum circuit to its linear map
 - $\gamma : \mathbf{QC} \rightarrow \mathbf{LM}$

Quantum circuits
implement a
linear map of qubits

Formal ZX diagram optimization

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Quantum circuits
implement a
linear map of qubits

- f is a function to transform quantum circuits
 - $f : \mathbf{QC} \rightarrow \mathbf{QC}$
 - f is semantic-preserving $\forall q \in \mathbf{QC}, \gamma(q) = \gamma(f(q))$
 - f minimizes the objective function $\min opt$

Functions that transform
quantum circuits need
to **keep the same**
linear map of qubits

$\Rightarrow f$ is a quantum circuit optimization algorithm

Formal ZX diagram optimization

ZX-based Quantum Circuit Optimization

- ZX diagrams
 - Let **ZX** be the infinite set of all finite ZX diagrams
- α is a function that converts quantum circuits to ZX diagrams
 - $\alpha : \mathbf{QC} \rightarrow \mathbf{ZX}$

Every
quantum circuit
can be converted
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Formal ZX diagram optimization

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 - Let **ZX** be the infinite set of all finite ZX diagrams
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- *extract* is a function that converts ZX diagrams to quantum circuits
 - $extract : \mathbf{ZX} \rightarrow \mathbf{QC} \cup \{\perp\}$
 - A ZX diagram is extractable if $zx \in \mathbf{ZX}, extract(zx) \neq \perp$

Every
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Not every
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Formal ZX diagram optimization

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- Rewriting rules
 - Let $\mathbf{R} = \{i1, i2, f, h, b, c, \pi, hd, lc\}$ be the set of rewriting rules
- $r \in R$ is a function that transforms a ZX diagram
 - $r \in R : \mathbf{ZX} \rightarrow \mathbf{ZX}$

Every

quantum circuit
can be converted
to a ZX diagram

Not every

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Rules are functions
that transform
ZX diagrams

⇒ A ZX based QC algorithm searches for extractable ZX diagrams that improves QC metric(s)

Formal ZX diagram optimization

ZX-based Quantum Circuit Optimization

State-space formed by finite sequence
of rules on converted quantum circuit

- W is the ZX state-space
 - $q \in QC, W \subseteq \mathbf{ZX}$ s.t. $w \in W$ exists for a finite sequence of rewriting rules
 - $w = (r_n \circ \dots \circ r_1 \circ \alpha)(q)$
- w is a function that converts a QC to a ZX diagram and applies a finite sequence of rewriting rules
 - $w = (r_n \circ \dots \circ r_1 \circ \alpha)(q)$

Formal ZX diagram optimization

ZX-based Quantum Circuit Optimization

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- Solutions are all extractable ZX diagrams in the state-space W :

- $S \subseteq W$
- $\forall w \in W, \text{extract}(w) \neq \perp \Leftrightarrow w \in S$

Extractable ZX diagrams of the state-space are solutions

Formal ZX diagram optimization

ZX-based Quantum Circuit Optimization

State-space formed by finite sequence of rules on converted quantum circuit

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Extractable ZX diagrams of the state-space are solutions

- **Optimal solutions** is the largest subset of solutions such that a metric is minimized
 - $O \subseteq S$
 - $\forall o \in O, s \in S, \text{opt}(\text{extract}(o)) \leq \text{opt}(\text{extract}(s))$

All solutions ZX diagrams that minimize a metric are optimal

Experimental Setup

Challenges

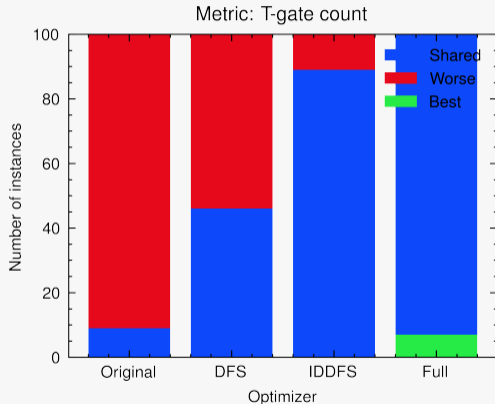
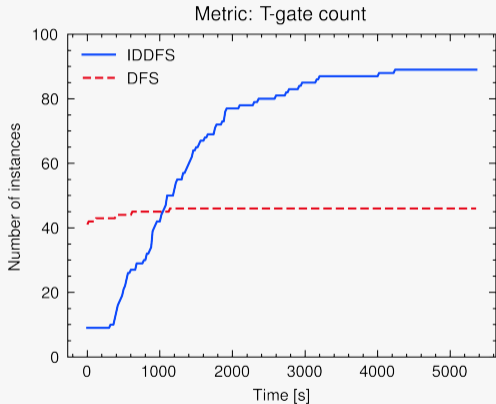
- Non-termination → select efficient pruning conditions
- Failed circuit extraction → compute metrics on ZX diagram; ensure graph-likeness
- High-memory requirement → open question

Benchmark

- 100 standard quantum circuits
- 1.5 hour global timeout
- Pruning conditions:
 - No colour cycle
 - No spider unfusion
 - Rule bundling
- DFS & IDDFS
- Connectivity change takes precedent over spider count

Results

T-gate reduction



Results

T-gate reduction

DFS

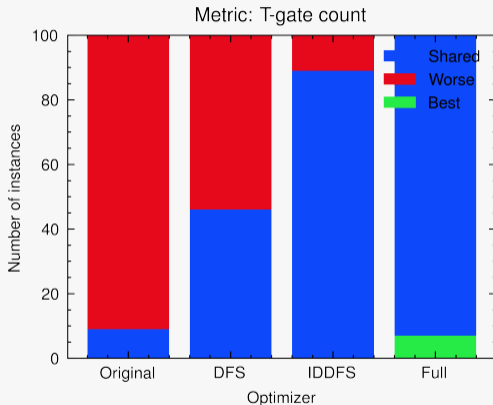
- reduces T-gate count by $\approx 10\%$
- equates Full reduce on 46% of the instances

IDDFS

- reduces T-gate count by $\approx 26\%$
- equates Full reduce on 89% of the instances

Full reduce [11]

- reduces T-gate count by $\approx 27\%$
- always leads to the best result



Results

T-gate reduction

DFS

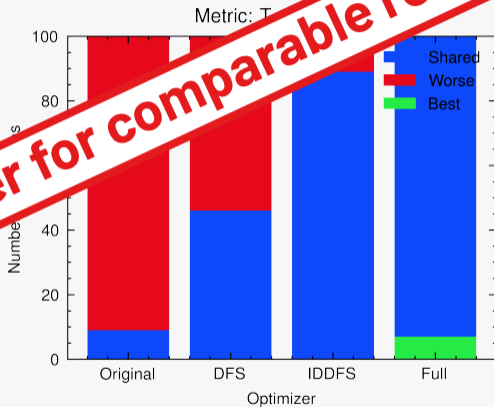
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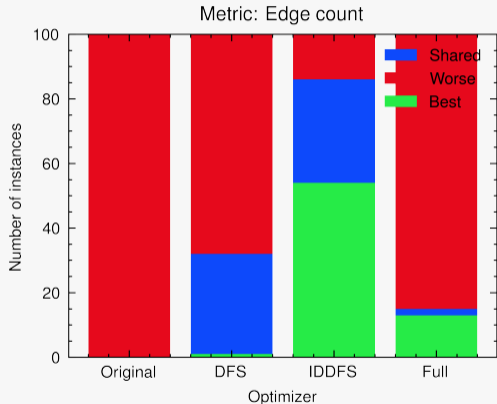
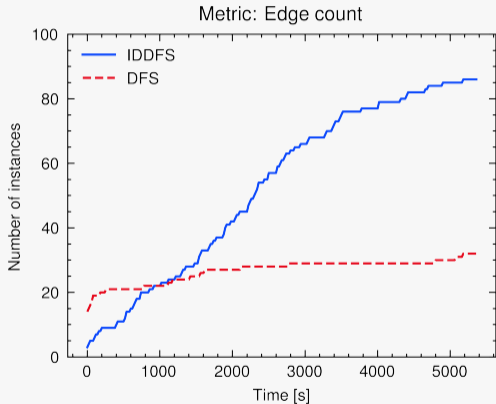
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Orders of magnitudes slower for comparable results

Results

Edge count reduction



Results

Edge count reduction

DFS

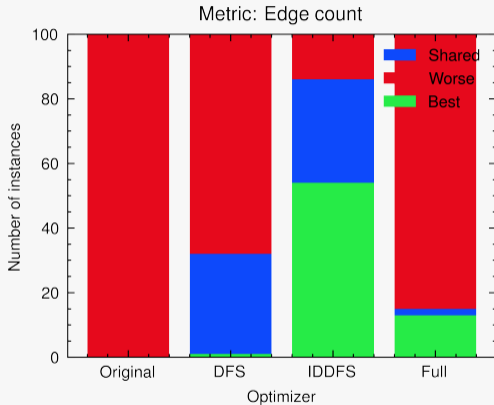
- reduces Edge count by $\approx 11\%$
- best solution on 32% of the instances

IDDFS

- reduces Edge count by $\approx 22\%$
- best solution on 86% of the instances
- comparable to SOTA algorithms that target Edge count (29% reduction) [23]

Full reduce

- not designed for edge count reduction
- reduces Edge count by $\approx 2\%$
- best solution on 15% of the instances



Results

Edge count reduction

DFS

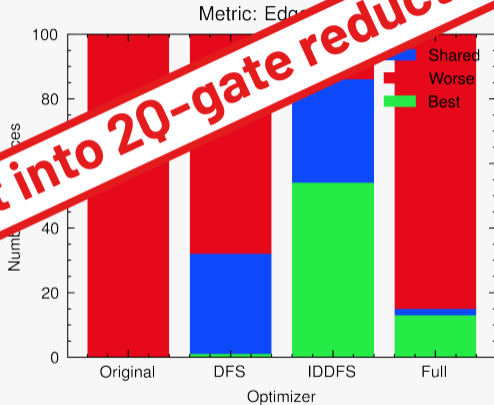
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- comparable to SOTA algorithm [23]
target Edge count (20%)

Full reduce

- not doing edge count reduction
- reduces Edge count by $\approx 2\%$
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How to translate Edge count into 2Q-gate reduction?

Implementation

- \approx 7000 LOC
- Integrates with pre-existing quantum compilation pipelines (Qiskit and PyZX)
- Solver and metric independent
- Currently supported solvers:
 - DFS, IDDFS, Local-Elimination
- New pruning conditions
- Snapshots for time-evolution
- Extensible code base
- Pre-defined benchmark with standard circuits

Code available on Gitlab.



Summary

Conclusion

- **Formalization** of ZX diagram optimization
- New set of **pruning conditions** for state-space reduction
- Reproducible **framework** that integrates in standard quantum compilation pipelines (≈ 7000 LOC)
- **Exhaustive search:**
 - **Equals T-gate count of SOTA** on 89% of the instances
 - **Reduces the edge count** by 22% on average **close to SOTA** (29%)

Outlook

- Improved solvers
 - **Limited-Discrepancy search**
 - Lexicographic search
 - **Better Local-Elimination implementation**
- New pruning conditions
- Target two-qubit gates from edge count

Thank you for your attention.






Check out the Framework.







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