Quantum complexity of minimum cut

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(CWI, ULB)

with Ronald de Wolf (FOCS '20, arXiv:1911.07306), Troy Lee (CCC '21, arXiv:2011.09823) and Paweł Gawrychowski

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Graph

graph
$$G = (V, E, w)$$



|V| = n nodes, $|E| = m \le {n \choose 2}$ edges, weights w

cut in G= set of edges $C \subseteq E$ such that G - C is disconnected



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minimum cut (value) in *G* = weight of cut with minimum total weight

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minimum cut (value) in G = weight of cut with minimum total weight = $\min_{C \subseteq E} \{ w(C) \mid G - C \text{ disconnected} \}$

cut in G

= set of edges $C \subseteq E$ such that G - C is disconnected



 $\begin{array}{l} \mbox{minimum cut (value) in } G \\ \mbox{= weight of cut with minimum total weight} \\ \mbox{= } \min_{C \subseteq E} \{ w(C) \mid G - C \mbox{ disconnected} \} \\ \mbox{= } \min_{S \subset V} \{ w(E(S, S^c)) \} \end{array}$

fundamental problem

in graph theory and combinatorial optimization

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relevant for graph robustness, graph partitioning (divide-and-conquer), VLSI design, ...



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related:

maximum flow, minimum *k*-cut, minimum *s*-*t* cut, sparsest cut, ...

long line of research

from 1970s (first polynomial algorithm by Podderyugin)

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through 1990s (first near-linear time $\widetilde{O}(m)$ algorithm by Karger)

to today (recent breakthrough by Li, arXiv:2106.05513))

Deterministic Mincut in Almost-Linear Time

Jason Li^{*} Carnegie Mellon University

December 16, 2020

this work: quantum complexity of minimum cut

i.e., what is the complexity of solving minimum cut on a quantum computer?

Quantum query complexity

quantum query complexity of minimum cut

= total number of queries to input

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e.g., # queries to string $x \in \{0, 1\}^N$ to determine OR(x)?

classically = $\Omega(N)$, quantum = $O(\sqrt{N})$ (Grover search)

adjacency matrix of G = (V, E, w):

$$A = \begin{bmatrix} w_{1,1} & w_{1,2} & \dots \\ w_{1,2} & w_{2,2} & \dots \\ \vdots & \vdots & \ddots \end{bmatrix} \in \mathbb{R}^{n \times n}$$

input size $= n^2$

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query:

 $|i,j,0\rangle \mapsto |i,j,w_{i,j}\rangle$

adjacency list of G = (V, E, w): (1): (2, $w_{1,2}$), (6, $w_{1,6}$), ... (2): (3, $w_{2,3}$), ... : (n): ... input size = $m \le \binom{n}{2}$

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Quantum time complexity

quantum time complexity of minimum cut

= total number of elementary gates, queries, QRAM operations*



* quantum-read/classical-write RAM

Why study quantum complexity of these problems?

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see e.g. talk "Quantum algorithms for optimization" by Ronald de Wolf at Simons' institute ('21) Quantum algorithms and graph problems

common caveats:

Quantum algorithms and graph problems

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(subquadratic) polynomial speedup

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QRAM requirements





 \bullet Grover search: $O(n^{1.5}),\, \widetilde{O}(n^{1.428\ldots})$ (103)



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• Grover search + quantum walks, learning graphs, ...: $\widetilde{O}(n^{1.3})$ (15), $O(n^{1.296...})$ (12), $O(n^{1.285...})$ (13), $\widetilde{O}(n^{1.25})$ (14)

* results by [Szegedy, '03], [Magniez-Santha-Szegedy, SICOMP'07], [Belovs, STOC'12],

[Lee-Magniez-Santha, SODA'13], [Le Gall, FOCS '14]



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$\leftrightarrow \Omega(n)$ lower bound

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related to element distinctness often paired with new tools in quantum query complexity

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Example: connectivity

Is graph G connected?



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Theorem (Dürr-Heiligman-Høyer-Mhalla, SICOMP '06) The quantum complexity of connectivity is

 $\widetilde{\Theta}(n^{3/2})$ [M] and $\widetilde{\Theta}(n)$ [L].
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 $\rightarrow \Omega(n^{3/2})$ [M] and $\Omega(n)$ [L] for minimum cut!

Quantum query bounds for minimum cut

quantum query complexity problem:

given quantum query access to



determine Hamming weight $|x| = \sum x_{ij}$

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[Beals-Buhrman-Cleve-Mosca-de Wolf, FOCS '98]: "counting" requires $n^2/2$ quantum queries to $x \rightarrow$ no quantum speedup for counting

minimum cut instance [M]:



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value minimum cut w(C) = |x|, so counting reduces to minimum cut

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no quantum speedup for minimum cut!

"approximate counting" loophole:

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quantum speedup for approximate minimum cut?

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 \downarrow

quantum speedup for minimum cut with few edges?

Quantum algorithm for *approximate* minimum cut

based on [A - de Wolf, FOCS '20]

key idea (exploits "approximate counting" loophole):

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2. solve minimum cut in sparsifier

cut sparsifier *H* is subgraph of *G* such that



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• *H* is sparse



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- *H* is sparse
- cuts in *H* approximate cuts in *G*:

$$w_H(E_H(S, S^c)) = (1 \pm \epsilon) w_G(E_G(S, S^c)), \quad \forall S \subset V$$



cut sparsifier H is subgraph of G such that

- *H* is sparse
- cuts in *H* approximate cuts in *G*:

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! in particular: minimum cut of $H \epsilon$ -approximates minimum cut of G

? how sparse can H be ?

Theorem (Karger '94, Benczúr-Karger '96)

Every graph *G* has an ϵ -cut sparsifier *H* with $\widetilde{O}(n/\epsilon^2)$ edges, which can be found in time $\widetilde{O}(m)$.

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- building block of many classical approximation algorithms for graph problems
- crucial component in efficient classical algorithms for *exact* minimum cut (Karger, J. ACM '00)

Quantum speedup for sparsification

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e.g., if $m \sim n^2$ then speedup from $\sim n^2$ (classical) to $\sim n^{3/2}$ (quantum)

Iterative sparsification: [Koutis-Xu '16, Fung-Hariharan-Harvey-Panigrahi '19]



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• construct $\widetilde{O}(1/\epsilon^2)$ minimum spanning trees and keep these edges

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construct *O*(1/ε²) minimum spanning trees and keep these edges
 keep any remaining edge with probability 1/2

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construct \$\tilde{O}(1/\epsilon^2)\$ minimum spanning trees and keep these edges
keep any remaining edge with probability 1/2

 \rightarrow *repeat* $O(\log n)$ *times:* ϵ -cut sparsifier with $\widetilde{O}(n/\epsilon^2)$ edges





• use quantum algorithm to construct $\tilde{O}(1/\epsilon^2)$ minimum spanning trees and keep these edges



- **1** use quantum algorithm to construct $\tilde{O}(1/\epsilon^2)$ minimum spanning trees and keep these edges
- implicitly sample any remaining edge with probability 1/2



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- **1** use quantum algorithm to construct $\tilde{O}(1/\epsilon^2)$ minimum spanning trees and keep these edges
- implicitly sample any remaining edge with probability 1/2

 \rightarrow repeat $O(\log n)$ times: ϵ -cut sparsifier with $\widetilde{O}(n/\epsilon^2)$ edges

total complexity (after more work): $\tilde{O}(\sqrt{mn}/\epsilon)$




1: Construct ϵ -sparsifier H

 $\triangleright \widetilde{O}(\sqrt{mn}/\epsilon)$ queries/time



- 1: Construct ϵ -sparsifier H
- 2: Solve minimum cut in H

 $\triangleright \widetilde{O}(\sqrt{mn}/\epsilon)$ queries/time \triangleright no queries, $\widetilde{O}(n)$ time

Theorem

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Similarly, quantum speedup for

- approximating other cut problems (max cut, sparsest cut, ...)
- approximately solving Laplacian systems

Ο...

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Ο...

Open question: is $\widetilde{O}(\sqrt{mn}/\epsilon)$ optimal?

Quantum algorithm for exact minimum cut in simple graphs

based on [A - Lee, CCC '21]

1. simple graphs have small number of edges in minimum cuts

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- 2. find these edges using graph sparsifier and Grover search

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! first focus on query complexity

"simple graphs have *small* number of edges involved in minimum cuts" [Kawarabayashi-Thorup, J. ACM '18]



Kawarabayashi-Thorup partition

Theorem (Kawarabayashi-Thorup)

For simple graphs, there exists partition $V = P_1 \cup \cdots \cup P_k$ such that

- there are O(n) edges between partitions,
- 2 partition "respects" all (nontrivial) near-minimum cuts.



Algorithm 1 Quantum query algorithm for MIN CUT

1:



Algorithm 2 Quantum query algorithm for MIN CUT

1: construct KT partition $V = P_1 \cup \cdots \cup P_k$

▷ ?? queries



Algorithm 3 Quantum query algorithm for MIN CUT

1: construct KT partition $V = P_1 \cup \cdots \cup P_k$

▷ ?? queries

2: find O(n) edges between partitions

 $\triangleright O(n^{3/2})_M / O(\sqrt{mn})_L$ queries



Algorithm 4 Quantum query algorithm for MIN CUT

- 1: construct KT partition $V = P_1 \cup \cdots \cup P_k$
- 2: find O(n) edges between partitions
- 3: find minimum cut of contracted graph



.

 $\triangleright O(n^{3/2})_M / O(\sqrt{mn})_L$ queries

▷ no queries



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⊳ ?? queries

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greedy algorithm:

repeatedly refine partition by iterating over near-minimum cuts



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too costly!

observation:

cut values in $G \approx$ cut values in sparsifier H



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hence,

KT partition of $H \approx$ KT partition of G

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KT partition of $H \approx$ KT partition of G

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1: Construct sparsifier H of G

 $\triangleright \widetilde{O}(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$ queries



- 1: Construct sparsifier H of G
- 2: construct KT partition of H

 $\triangleright \widetilde{O}(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$ queries

⊳ no queries



- 1: Construct sparsifier H of G
- 2: construct KT partition of H
- 3: find O(n) edges of *G* between partitions

 $\triangleright \widetilde{O}(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$ queries

▷ no queries

 $\triangleright O(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$ queries



1: Construct sparsifier *H* of *G* 2: construct KT partition of *H* 3: find O(n) edges of *G* between partitions 4: solve minimum cut on contracted graph in time $\widetilde{O}(n)$ $\sim no$ queries $\sim O(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$ queries $\sim O(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$ queries



Adjacency matrix model

Adjacency matrix model

algorithm for minimum cut with quantum query complexity $\widetilde{O}(n^{3/2})$

+

 $\Omega(n^{3/2})$ lower bound on quantum query complexity of connectivity (Dürr-Heiligman-Høyer-Mhalla '06)

Adjacency matrix model

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Theorem

The quantum query complexity of minimum cut in simple graphs is

 $\widetilde{\Theta}(n^{3/2})$ [M].

algorithm for minimum cut with quantum query complexity $\tilde{O}(\sqrt{mn})$

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Theorem (adjacency list)

The quantum query complexity of minimum cut in simple graphs is

 $\widetilde{O}(\sqrt{mn})$ [L] and $\Omega(n)$ [L].

algorithm for minimum cut with quantum query complexity $\tilde{O}(\sqrt{mn})$

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Theorem (adjacency list)

The quantum query complexity of minimum cut in simple graphs is

 $\widetilde{O}(\sqrt{mn})$ [L] and $\Omega(n)$ [L].

! open question: $\widetilde{O}(n)$ quantum algorithm ?

Time efficient quantum algorithm for minimum cut

based on [A - Lee - Gawrychowski]
Quantum algorithm for minimum cut (adj.list)

- 1: Construct sparsifier H of G
- 2: construct KT partition of H
- 3: find O(n) edges of G between partitions
- 4: solve MIN CUT on contracted graph in time $\widetilde{O}(n)$

 \triangleright time $\widetilde{O}(n^{3/2})_M/\widetilde{O}(\sqrt{mn})_L$ [A-de Wolf '20]

⊳ time ???

 \triangleright time $O(n^{3/2})_M/O(\sqrt{mn})_L$

 \triangleright time $\widetilde{O}(n)$ [Karger '00]



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Kawarabayashi-Thorup (STOC '15): classical algorithm for simple graphs in time $\widetilde{O}(m)$



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Kawarabayashi-Thorup (STOC '15): classical algorithm for simple graphs in time $\widetilde{O}(m)$

•

A-Lee (CCC '21): quantum algorithm for weighted graphs in time $\widetilde{O}(n^{3/2})$ [M,L]



⊳ time ???

•

Kawarabayashi-Thorup (STOC '15): classical algorithm for simple graphs in time $\widetilde{O}(m)$

•

A-Lee (CCC '21): quantum algorithm for weighted graphs in time $\widetilde{O}(n^{3/2})$ [M,L]

•

A-Gawrychowski-Lee (forthcoming): classical algorithm for weighted graphs in time $\widetilde{O}(m)$ quantum algorithm for weighted graphs in time $\widetilde{O}(\sqrt{mn})$ [L] **Classical algorithm for KT partition**

key ideas:

- 1. "tree-respecting cuts"
- 2. generating set for KT partition

introduced by Karger (J. ACM '00)

graph G, spanning tree T



introduced by Karger (J. ACM '00)

graph G, spanning tree T



cut "2-respects" T if cuts at most 2 edges of T

introduced by Karger (J. ACM '00)

graph G, spanning tree T



cut "2-respects" *T* if cuts at most 2 edges of *T* ! at most $\binom{n-1}{2}$ 2-respecting cuts

Lemma (Karger)

Any near-minimum cut 2-respects "random" spanning tree T with large probability.

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"better" greedy algorithm for KT partition: iterate over 2-respecting cuts

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Any near-minimum cut 2-respects "random" spanning tree T with large probability.

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! still too slow



Lemma (A-Lee '20, A-Gawrychowski-Lee '21)



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• Characterize "generating" set of *O*(*n*) 2-respecting cuts that induces KT partition



Lemma (A-Lee '20, A-Gawrychowski-Lee '21)

- Characterize "generating" set of *O*(*n*) 2-respecting cuts that induces KT partition
- Find generating set in time $\widetilde{O}(m)$ (classically) or $\widetilde{O}(\sqrt{mn})$ (quantum)



Theorem

Can construct the KT partition of a weighted graph in time $\widetilde{O}(m)$ (classically) or $\widetilde{O}(\sqrt{mn})$ (quantumly).

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Can construct the KT partition of a weighted graph in time $\widetilde{O}(m)$ (classically) or $\widetilde{O}(\sqrt{mn})$ (quantumly).

Corollary

There is a quantum algorithm for finding a minimum cut in a simple graph with time complexity $\tilde{O}(n^{3/2})_M/\tilde{O}(\sqrt{mn})_L$.

Quantum time/query complexity of minimum cut:

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weighted graphs, exact: no quantum speedup

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Quantum time/query complexity of minimum cut:

weighted graphs, exact: no quantum speedup weighted graphs, approximate: $\widetilde{O}(n^{3/2}/\epsilon)_M / \widetilde{O}(\sqrt{mn}/\epsilon)_L$ simple graphs, exact: $\widetilde{\Theta}(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$

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weighted graphs, exact: no quantum speedup weighted graphs, approximate: $\widetilde{O}(n^{3/2}/\epsilon)_M / \widetilde{O}(\sqrt{mn}/\epsilon)_L$ simple graphs, exact: $\widetilde{\Theta}(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$

Key tools:

graph sparsification and Kawarabayashi-Thorup partition

Quantum time/query complexity of minimum cut:

weighted graphs, exact: no quantum speedup weighted graphs, approximate: $\widetilde{O}(n^{3/2}/\epsilon)_M / \widetilde{O}(\sqrt{mn}/\epsilon)_L$ simple graphs, exact: $\widetilde{\Theta}(n^{3/2})_M / \widetilde{O}(\sqrt{mn})_L$

Key tools:

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Open questions:

$\widetilde{O}(n)_L$ quantum algorithms for minimum cut and sparsification in simple graphs?

 $\Omega(\sqrt{mn}/\epsilon)_L$ lower bound for approximate minimum cut?