

Minimum and Approximate Minimum k -Cuts in Hypergraphs

Chris Bao, Joshua Wang, William Zhao
Mentor: Yuchong Pan

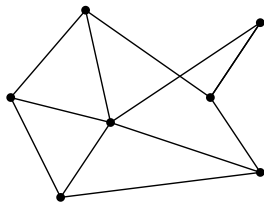
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- 1 Introduction: Minimum Cuts and Karger's Algorithm
- 2 Randomized Contraction Bounds for Approximate Minimum k -Cuts
- 3 The Branching Contraction Algorithm in Unweighted Hypergraphs
- 4 k -Cut Approximation with Hypertree Packing

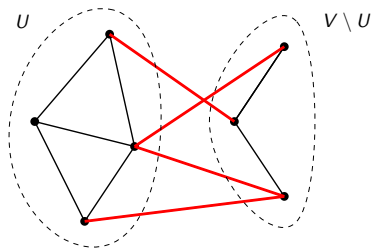
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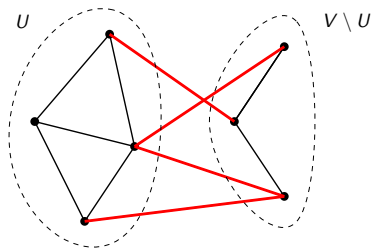


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An edge e **crosses** a cut U if it has one endpoint in U and the other in $V \setminus U$. The **value** of a cut is the number of edges that cross it.

Minimum Cuts in Graphs

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- Example application: Network reliability analysis
- Question: How can we efficiently find minimum and approximate minimum cuts?
- Question: How many minimum cuts can be in a graph with n vertices? How many α -approximate minimum cuts?

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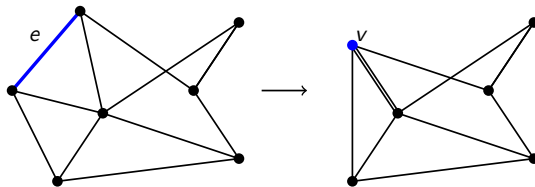
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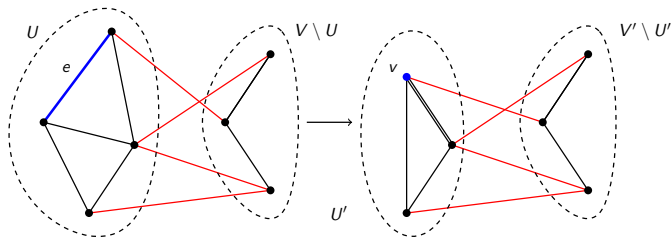


Karger's Algorithm

- For an edge e , a cut $C = (U, V \setminus U)$ will **survive** the contraction of e if e does not cross C

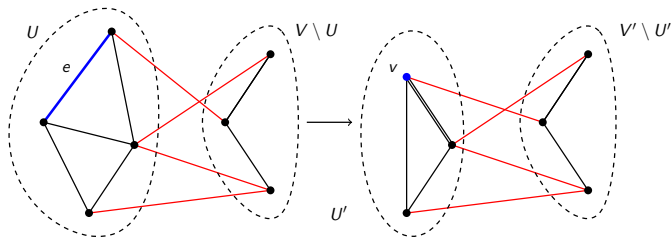
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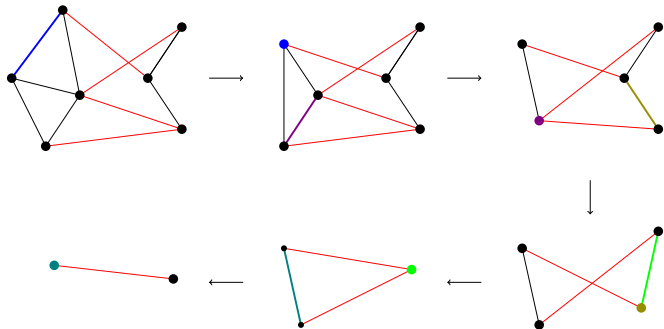
- The smaller a cut C , the more likely C survives the contraction of a randomly selected edge. In particular, a minimum cut is very likely to survive random contraction.

Karger's Algorithm

- **Karger's Algorithm** contracts randomly selected edges until only 2 vertices are left, and then returns the cut between those vertices

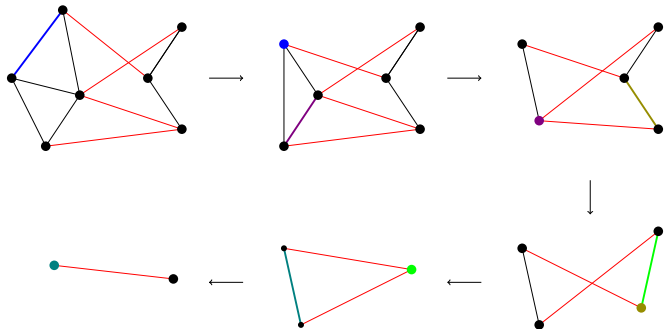
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Theorem (Karger 1993)

There are at most $\binom{n}{2} = O(n^2)$ minimum cuts in an n -vertex graph. Furthermore, there are $O(n^{2\alpha})$ α -approximate minimum cuts.

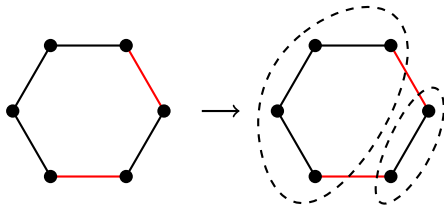
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- The bound is tight: take a cycle graph on n vertices



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Definition

A **k -way cut** (or **k -cut** for short) is a partition of the vertices V into k non-empty subsets U_1, U_2, \dots, U_k . An edge **crosses** a k -cut if its vertices are in different subsets U_i and U_j , and the **size** of a k -cut is the number of edges crossing it.

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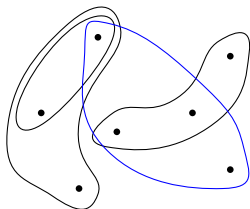
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- Karger's algorithm readily generalizes to the k -cut problem - we stop contraction at k vertices remaining instead of two.

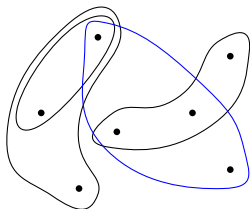
Hypergraphs

- In a graph, edges can connect only two vertices. In a **hypergraph**, we allow **hyperedges** to connect multiple vertices (i.e., each hyperedge $e \in E$ is a subset of V , so $E \subseteq 2^V$.)



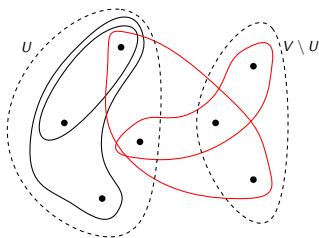
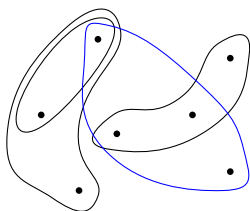
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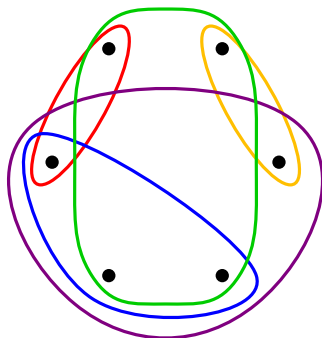
- We generalize the minimum cut and minimum k -cut problems to hypergraphs.

Hypergraphs

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- Example: Communication platforms



Discord

Snapchat

Texting

Email

Carrier Pigeon

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Theorem (Bao, Pan, Wang, and Zhao 2024+)

In a rank- r hypergraph H with n vertices, there are at most

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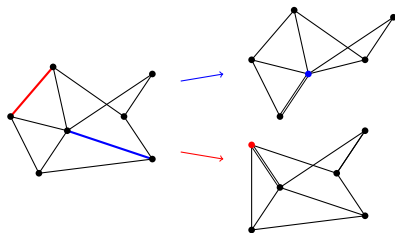
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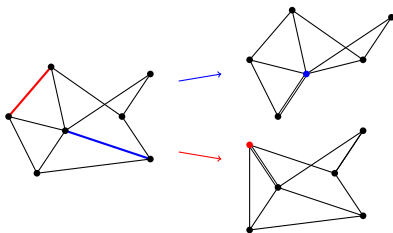
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- Branches occur at **predetermined times**: every time $|V|$ decreases by a factor of $\sqrt{2}$.

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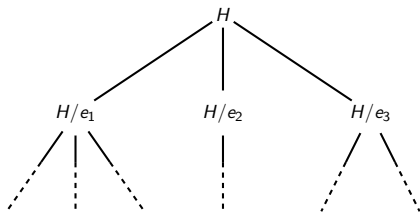
$$P_{\text{Branch}}(e) = 1 - \frac{\binom{n-|e|}{k-1}}{\binom{n}{k-1}}$$

Contraction Tree

- The contracted hypergraphs can be organized into a tree. Although the tree itself is random, we can still analyze its expected properties.

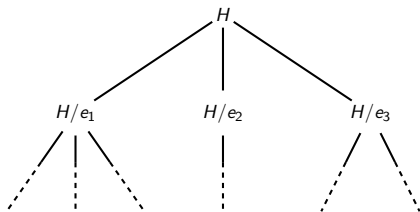
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Theorem (Fox et al. 2018)

The branching contraction algorithm finds a minimum k -cut with high probability in $\tilde{O}(mn^{2k-2})$ expected time

Branching Contraction Time Complexity

- These bounds are tight for graphs. However, $m = O(n^2)$ in a graph, while $m = \Omega(2^n)$ is possible in a hypergraph. A finer examination of the time complexity is needed.

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Theorem (Bao, Pan, Wang, and Zhao 2024+)

For an unweighted hypergraph without parallel edges, the branching contraction algorithm works in $\tilde{O}(mn^k + n^{2k})$ expected time.

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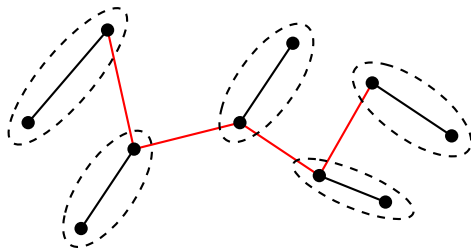
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$$k = 5$$

$$|F| = 9$$

$$n = 10$$

$$|F| + k - n = 4$$

Fractional k -Cut

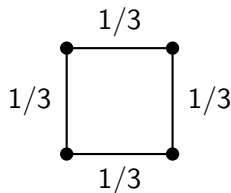
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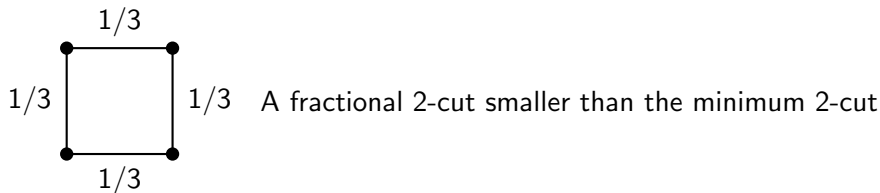
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A fractional 2-cut smaller than the minimum 2-cut

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In graphs, the minimum k -cut has value at most $2(1 - 1/n)$ times the minimum fractional k -cut. Furthermore, the minimum fractional k -cut is easier to compute.

Approximating Minimum k -cut Using Tree Packing

- Quanrud (2019) begins by computing a $(1 + \varepsilon)$ -approximate minimum fractional k -cut, using linear programming and a technique called Multiplicative Weight Update.

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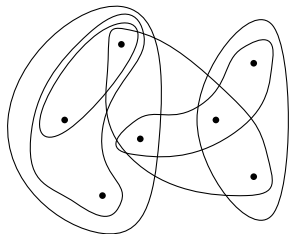
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- The fractional k -cut can be “rounded” to a $2(1 + \varepsilon)$ -approximate minimum k -cut.
- Take every edge with weight at least $1/2$. If this does not form a k -cut, then greedily complete with cuts from the minimum spanning tree (or forest).

Definition

A **hyperforest** is a set of hyperedges F such that any subset $X \subseteq F$ contains at least $|X| + 1$ vertices.

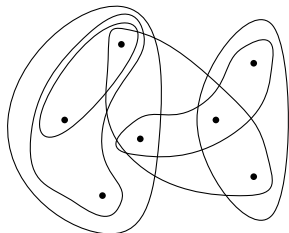
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Theorem (Bao, Pan, Wang, and Zhao 2024+)

A minimum hypertree can be computed in $O(mn^2)$ time.

Hypertree Packing

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Theorem (Bao, Pan, Wang, and Zhao 2024+)

There exists a $r(1 + \varepsilon)$ -approximation to minimum k -cut in

$$O(mn \log^2 n / \varepsilon^2 + mn^2)$$

time for hypergraphs, where r is the rank of the hypergraph.

Acknowledgements

- We thank our mentor, Yuchong Pan, for his guidance and support.
- We thank PRIMES-USA and the organizers for this opportunity.

- D. Karger and C. Stein. *A new approach to the minimum cut problem*. J. ACM **43** (1996).
- D. Kogan and R. Krauthgamer. *Sketching cuts in graphs and hypergraphs*. ITCS '15 **6** (2015).
- M. Baiou and F. Barahona. *On some algorithmic aspects of hypergraphic matroids*. Discrete Math. **346** (2023)
- K. Fox, D. Panigrahi, and F. Zhang. *Minimum cut and minimum k -cut in hypergraphs via branching contractions*. SIAM (2019)
- K. Quanrud. *Fast and Deterministic Approximations for k -Cut*. LIPIcs **145** (2019).