

Connected cuts

Master thesis

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Date 2025

Table of Contents

Preliminaries

Notation

MINIMUM CONNECTED k -CUT

MINIMUM CONNECTED k -CUT WITH SOURCE

NP-completeness

NP-completeness of MINIMUM CONNECTED k -CUT WITH SOURCE

NP-completeness of MINIMUM CONNECTED k -CUT

Bicriteria approximation algorithm

Absorptive flow

Integer linear program

Approximation

Algorithm

Notation

- graph $G = (V, E)$, V - vertices, $E \subseteq \binom{V}{2}$ - edges
- $\{u, v\}$ - single edge
- $n := |V|$
- $[l] := \{1, 2, \dots, l\}$ and $[k, l] := \{k, k + 1, \dots, l\}$
- $\langle \ell, r \rangle := \{x \in \mathbb{R} \mid \ell \leq x \leq r\}$

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Definition

The **partition** of vertices V into $\ell \in \mathbb{N}$ parts is a set $\mathcal{V} = \{V_1, V_2, \dots, V_\ell\}$ such that $\bigcup_{i=1}^{\ell} V_i = V$ and $\forall i \neq j \in [\ell] : V_i \cap V_j = \emptyset$.

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Definition A graph is **connected** if and only if every pair of vertices has a path between them.

Definition

An **induced subgraph** $G[S]$ on vertices $S \subseteq V$ is a graph $G' = (S, E')$, s.t.

$\{u, v\} \in E' \iff u \in S \wedge v \in S \wedge \{u, v\} \in E$. We say that $S \subseteq V$ is connected if the induced subgraph $G[S]$ is connected.

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$\{u, v\} \in E' \iff u \in S \wedge v \in S \wedge \{u, v\} \in E$. We say that $S \subseteq V$ is **connected** if the induced subgraph $G[S]$ is connected.

- $E(S, V \setminus S)$ - edges between S and $V \setminus S$
- $e(S, V \setminus S) = |E(S, V \setminus S)|$
- partition $S \subset V$ and $V \setminus S$ is a **cut**
- we also call $E(S, V \setminus S)$ a cut (the terms are interchangeable)

Definition

For a connected graph $G = (V, E)$ and $k \in \mathbb{N}$, a **connected k -cut** is $S \subset V$ such that all following properties hold:

- 1 $G[S]$ is connected.
- 2 $|S| = k$.

MINIMUM CONNECTED k -CUT

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Problem

Minimum connected k -cut is to find connected k -cut minimizing $e(S, V \setminus S)$.

MINIMUM CONNECTED k -CUT WITH SOURCE

Definition For a connected graph $G = (V, E)$, a vertex $s \in V$ and $k \in \mathbb{N}$ a **connected k -cut with source** is $S \subset V$ such that all following properties hold:

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Problem *Minimum connected k -cut with source is to find connected k -cut with source minimizing $e(S, V \setminus S)$.*

MINIMUM CONNECTED k -CUT

Example

See a MINIMUM CONNECTED k -CUT example. For this graph we consider $k = 3$.

- 1 $G[S]$ is connected.
- 2 $|S| = k$.
- 3 $e(S, V \setminus S)$ is minimized.

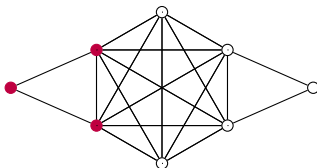


Figure: Possible solution with purple vertices chosen into S .

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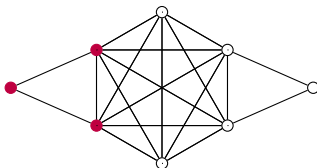


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Question

What if we omit one of them.

Connected k -cut

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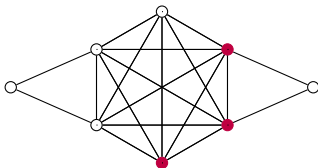


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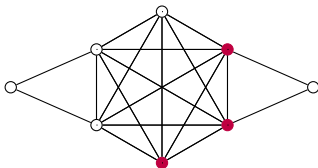


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Solution

We may use only search algorithms.

MINIMUM CONNECTED CUT

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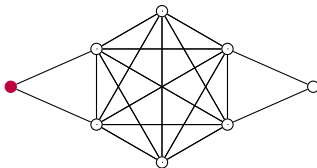


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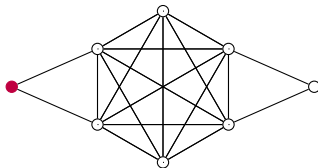


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Solution

Either use flows and cuts or polyhedron by Garg [4].

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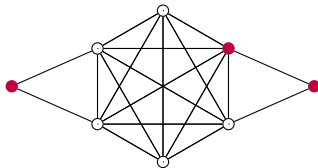


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MINIMUM k -CUT

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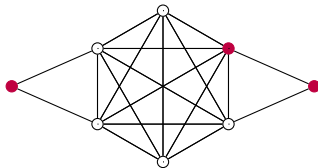


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Solution

Use the results from minimum bisection by Feige and Krauthgamer [1].

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- We use polynomial time reduction from MINIMUM BISECTION.
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- The MINIMUM BISECTION problem is NP-complete [2, 3, 5].

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- We may assume that $|V|$ is even.
- The MINIMUM BISECTION problem is NP-complete [2, 3, 5].

Problem

Given graph $G = (V, E)$ the goal is to find a partition $\mathcal{V} = \{V_1, V_2\}$ where $|V_1| = |V_2| = \frac{1}{2}|V|$ and $e(V_1, V_2)$ is minimized.

NP-completeness of MINIMUM CONNECTED k -CUT WITH SOURCE

Lemma

The MINIMUM CONNECTED k -CUT WITH SOURCE problem is NP-complete.

NP-completeness of MINIMUM CONNECTED k -CUT WITH SOURCE

Lemma *The MINIMUM CONNECTED k -CUT WITH SOURCE problem is NP-complete.*

Proof.

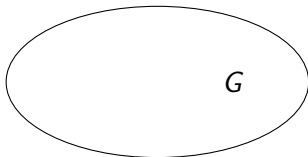
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- set $k = \frac{n}{2} + 1$

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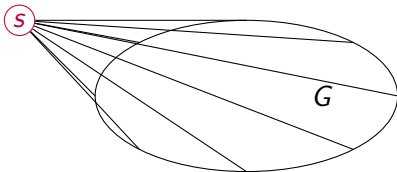


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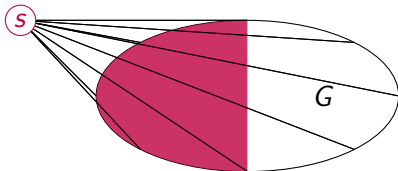


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NP-completeness of MINIMUM CONNECTED k -CUT

Theorem

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- create graph $H := K_{n^2}$ and choose n vertices into B
- connect B to G via matching and set $k = n^2 + \frac{n}{2}$

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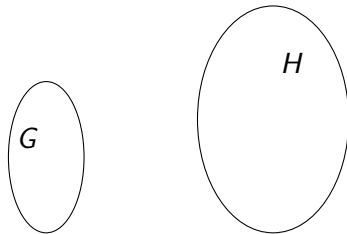


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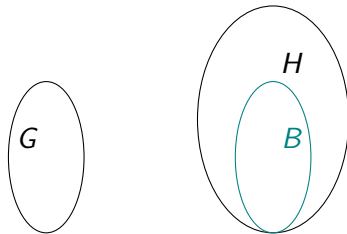


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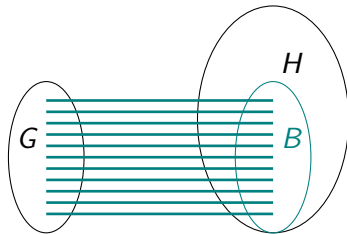


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Given a graph $G = (V, E)$, $s \in V$ and an integer $k \in \mathbb{N}$, $k \leq n$, an **absorptive flow** is a pair of functions (f_V, f_E) , $f_V : V \rightarrow \mathbb{R}$, $f_E : E \rightarrow \mathbb{R}$, satisfying the following properties:

- 1 $\sum_{v \in V} f_V(v) = k$
- 2 $f_V(s) = 1$
- 3 $\forall v \in V : 0 \leq f_V(v) \leq 1$
- 4 $\sum_{v \in V, \{s, v\} \in E} f_E(\{s, v\}) = k - 1$
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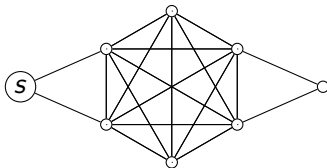
Definition

Given an absorptive flow (f_V, f_E) on a graph $G = (V, E)$ we set $S = \{v \in V \mid f_V(v) > 0\}$. The **boundary of absorptive flow** is $E(S, V \setminus S)$, where its size is $e(S, V \setminus S)$.

Examples of absorptive flows

Example

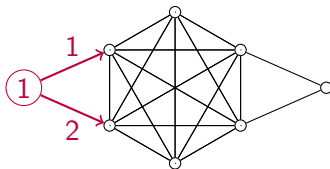
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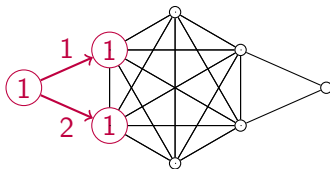
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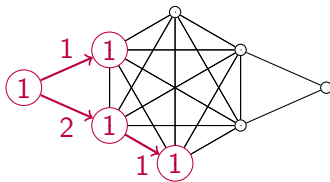
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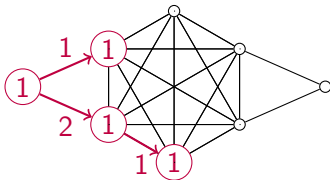
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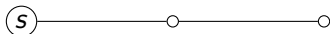
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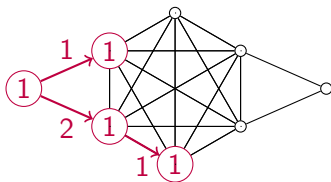
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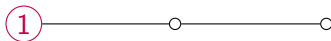
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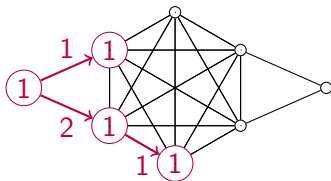
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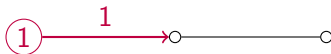
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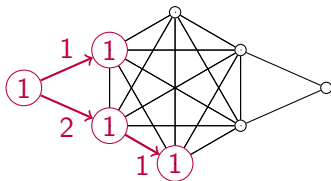
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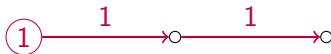
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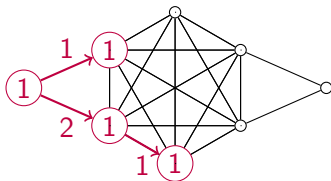
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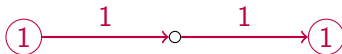
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Integer linear program (ILP) for an absorptive flow with min boundary

Variables

$$\forall v \in V \quad f_v = \begin{cases} 1 & \text{if the vertex } v \text{ absorbs.} \\ 0 & \text{otherwise.} \end{cases}$$

$$\forall \{u, v\} \in E \quad f_{uv}, f_{vu} \geq 0 \text{ the amount of flow} \\ \text{on edge } \{u, v\} \\ \text{going in the respective direction.}$$

$$\forall e \in E \quad x_e = \begin{cases} 1 & \text{if } e \in E(S, V \setminus S). \\ 0 & \text{otherwise.} \end{cases}$$

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Objective

$$\min \sum_{e \in E} x_e$$

Integer linear program (ILP) for an absorptive flow with min boundary

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$$x_{\{u,v\}} \geq f_u - f_v \quad \forall \{u,v\} \in E$$

$$x_{\{u,v\}} \geq f_v - f_u \quad \forall \{u,v\} \in E$$

$$\sum_{v \in V, \{s,v\} \in E} f_{sv} = k - 1$$

$$f_s = 1$$

$$\sum_{u \in V, \{u,v\} \in E} f_{uv} = \sum_{u \in V, \{v,u\} \in E} f_{vu} + f_v \quad \forall v \in V \setminus \{s\}$$

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$$(k - 1) \cdot f_v \geq \sum_{u \in V, \{uv\} \in E} f_{uv} \quad \forall v \in V \setminus \{s\}$$

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$$\sum_{u \in V} f_u = k$$

$$(k - d(s, v)) \cdot f_v \geq \sum_{u \in V, \{uv\} \in E} f_{uv} \quad \forall v \in V \setminus \{s\}$$

Lemma *The optimal solution (x^*, f^*) of ILP correspond to a MINIMUM CONNECTED k -CUT and vice versa.*

ILP properties

Lemma *The optimal solution (x^*, f^*) of ILP correspond to a MINIMUM CONNECTED k -CUT and vice versa.*

Lemma *The integrality gap between the integer linear program and its linear relaxation is at least k .*

Bound a relaxation of the ILP

- d_x is a pseudo-metric on $V \times V \rightarrow \mathbb{R}_0^+$; derived from x_e as a shortest path
- OPT_{LP} denotes the optimum of linear program relaxation

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Lemma

For any $0 \leq r < 1$ these bounds $1 \leq |\mathcal{B}_{d_x}(s, r)| \leq 1 + \frac{k-1}{1-r}$ holds.

Bounds

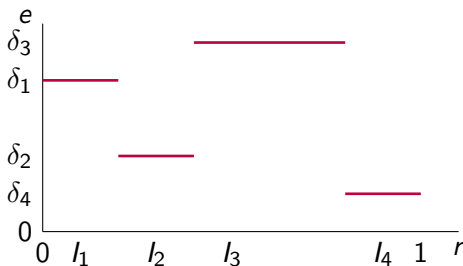
Lemma

Given a constant $0 < c < 1$, let $\mathcal{I}_c = \{r \in \langle 0, 1 \rangle \mid e(\mathcal{B}_{d_x}(s, r), V \setminus \mathcal{B}_{d_x}(s, r)) > \frac{1}{c} \text{OPT}_{\text{LP}}\}$. Then \mathcal{I}_c corresponds to a set of subintervals of $\langle 0, 1 \rangle$ and the sum of their length is at most c .

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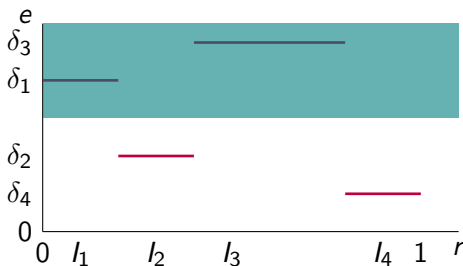
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Algorithm

- we propose a bicriteria approximation algorithm for the MINIMUM CONNECTED k -CUT WITH SOURCE problem, and consequently also for MINIMUM CONNECTED k -CUT problem
- the technique we use is known as a *Ball growing technique* [4]

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For a given graph $G = (V, E)$, $k \in \mathbb{N}$ and $0 < c < 1$ there exists a polynomial time algorithm, which produces a connected cut S , such that $1 \leq |S| \leq \mathcal{O}(k)$ and $e(S, V \setminus S) \leq \frac{1}{c} \text{OPT}_{\text{LP}}$.

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- proof follows from previous Lemmata and the fact, that the function is piece-wise constant
- the stricter c we choose the worse $|S|$ we may get, but there is no guarantee
- the choice of c does not change the time complexity

Summary

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- defining MINIMUM CONNECTED k -CUT and MINIMUM CONNECTED k -CUT WITH SOURCE problems
- proving NP-completeness
- bicriteria approximation of LP relaxation with the usage of a growing ball technique

**Thank you for your
attention.**

Bibliography



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