

Extended Formulations for Radial Cones

Matthias Walter (RWTH Aachen)

Joint work with

Stefan Weltge (TU Munich)

IMO Oberseminar, Magdeburg, 02.11.2018



Combinatorial optimization problem:

- ▶ Ground set E (finite)
- ▶ Feasible solutions $\mathcal{F} \subseteq 2^E$
- ▶ Objective vector $c \in \mathbb{Q}^E$
- ▶ Goal: minimize cost $c(F) := \sum_{e \in F} c_e$ over all $F \in \mathcal{F}$.

Combinatorial optimization problem:

- ▶ Ground set E (finite)
- ▶ Feasible solutions $\mathcal{F} \subseteq 2^E$
- ▶ Objective vector $c \in \mathbb{Q}^E$
- ▶ Goal: minimize cost $c(F) := \sum_{e \in F} c_e$ over all $F \in \mathcal{F}$.

Augmentation problem:

- ▶ Given $F \in \mathcal{F}$, determine optimality or find $F' \in \mathcal{F}$ with $c(F') < c(F)$.

Combinatorial optimization problem:

- ▶ Ground set E (finite)
- ▶ Feasible solutions $\mathcal{F} \subseteq 2^E$
- ▶ Objective vector $c \in \mathbb{Q}^E$
- ▶ Goal: minimize cost $c(F) := \sum_{e \in F} c_e$ over all $F \in \mathcal{F}$.

Augmentation problem:

- ▶ Given $F \in \mathcal{F}$, determine optimality or find $F' \in \mathcal{F}$ with $c(F') < c(F)$.

Theorem (Schulz, Weismantel & Ziegler, 1995; Grötschel & Lovász, 1995)

*We can solve the augmentation problem (for arbitrary objective vectors) in polynomial time **if and only if** we can solve the optimization problem (for arbitrary objective vectors) in polynomial time.*

Combinatorial optimization problem:

- ▶ Ground set E (finite)
- ▶ Feasible solutions $\mathcal{F} \subseteq 2^E$
- ▶ Objective vector $c \in \mathbb{Q}^E$
- ▶ Goal: minimize cost $c(F) := \sum_{e \in F} c_e$ over all $F \in \mathcal{F}$.

Augmentation problem:

- ▶ Given $F \in \mathcal{F}$, determine optimality or find $F' \in \mathcal{F}$ with $c(F') < c(F)$.

Theorem (Schulz, Weismantel & Ziegler, 1995; Grötschel & Lovász, 1995)

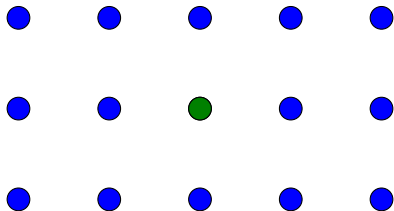
*We can solve the augmentation problem (for arbitrary objective vectors) in polynomial time **if and only if** we can solve the optimization problem (for arbitrary objective vectors) in polynomial time.*

Idea:

- ▶ Suppose $c \in \{0, 1\}^E$, how many augmentation steps will you need?
- ▶ Apply bit scaling.

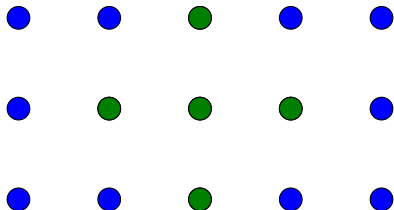
Polyhedral method:

- Identify $F \in \mathcal{F}$ with $\chi(F) \in \{0, 1\}^E$ s.t. $\chi(F)_e = 1 \iff e \in F$.



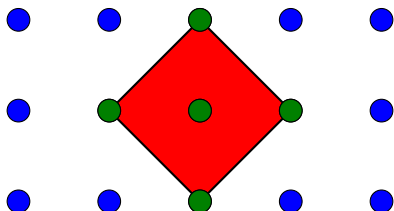
Polyhedral method:

- ▶ Identify $F \in \mathcal{F}$ with $\chi(F) \in \{0, 1\}^E$ s.t. $\chi(F)_e = 1 \iff e \in F$.
- ▶ Let $X := \{\chi(F) : F \in \mathcal{F}\} \subseteq \{0, 1\}^E$.
- ▶ Optimization problem is then to minimize $\langle c, x \rangle$ over $x \in X$.



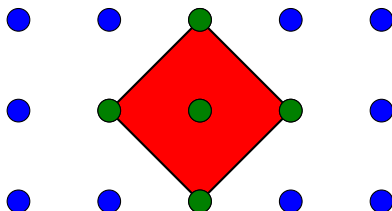
Polyhedral method:

- ▶ Identify $F \in \mathcal{F}$ with $\chi(F) \in \{0, 1\}^E$ s.t. $\chi(F)_e = 1 \iff e \in F$.
- ▶ Let $X := \{\chi(F) : F \in \mathcal{F}\} \subseteq \{0, 1\}^E$.
- ▶ Optimization problem is then to minimize $\langle c, x \rangle$ over $x \in \text{conv}(X)$.



Polyhedral method:

- ▶ Identify $F \in \mathcal{F}$ with $\chi(F) \in \{0, 1\}^E$ s.t. $\chi(F)_e = 1 \iff e \in F$.
- ▶ Let $X := \{\chi(F) : F \in \mathcal{F}\} \subseteq \{0, 1\}^E$.
- ▶ Optimization problem is then to minimize $\langle c, x \rangle$ over $x \in \text{conv}(X)$.
- ▶ Find an outer description of $\text{conv}(X)$, i.e., $\text{conv}(X) = \{x \in \mathbb{R}^E : Ax \leq b\}$.
- ▶ Optimization problem is now an LP and we can use black-box solvers.¹



¹... or devise primal-dual algorithms.

One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.

One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

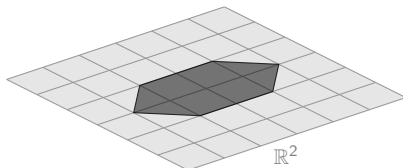
One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi: \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?



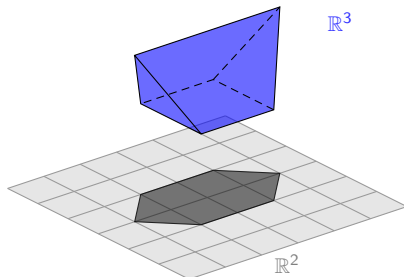
One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi: \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?



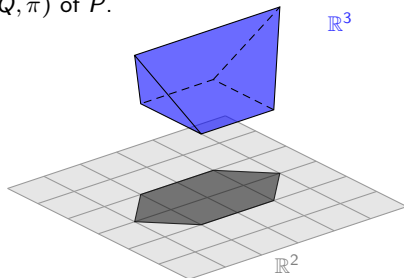
One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi : \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .



One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi : \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .
- ▶ Alternative viewpoint: model using additional variables

One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

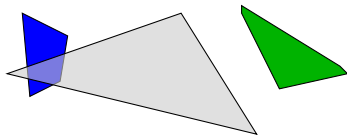
Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi : \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .
- ▶ Alternative viewpoint: model using additional variables

Theorem (Balas, 1979)

Let $P_1, \dots, P_k \subseteq \mathbb{R}^n$ be polytopes. Then $\text{xc}(P_1 \cup \dots \cup P_k) \leq \sum_{i=1}^k (\text{xc}(P_i) + 1)$.

Disjunctive programming:



One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

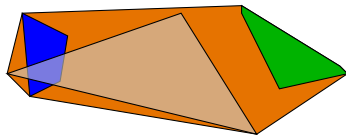
Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi : \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .
- ▶ Alternative viewpoint: model using additional variables

Theorem (Balas, 1979)

Let $P_1, \dots, P_k \subseteq \mathbb{R}^n$ be polytopes. Then $\text{xc}(P_1 \cup \dots \cup P_k) \leq \sum_{i=1}^k (\text{xc}(P_i) + 1)$.

Disjunctive programming:



One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi: \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .
- ▶ Alternative viewpoint: model using additional variables

Theorem (Balas, 1979)

Let $P_1, \dots, P_k \subseteq \mathbb{R}^n$ be polytopes. Then $\text{xc}(P_1 \cup \dots \cup P_k) \leq \sum_{i=1}^k (\text{xc}(P_i) + 1)$.

For parity polytope:

- ▶ $X = \bigcup_{k \text{ even}} \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i = k\}$

One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi: \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .
- ▶ Alternative viewpoint: model using additional variables

Theorem (Balas, 1979)

Let $P_1, \dots, P_k \subseteq \mathbb{R}^n$ be polytopes. Then $\text{xc}(P_1 \cup \dots \cup P_k) \leq \sum_{i=1}^k (\text{xc}(P_i) + 1)$.

For parity polytope:

- ▶ $\text{conv}(X) = \text{conv}\left(\bigcup_{k \text{ even}} \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i = k\}\right)$

One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi: \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .
- ▶ Alternative viewpoint: model using additional variables

Theorem (Balas, 1979)

Let $P_1, \dots, P_k \subseteq \mathbb{R}^n$ be polytopes. Then $\text{xc}(P_1 \cup \dots \cup P_k) \leq \sum_{i=1}^k (\text{xc}(P_i) + 1)$.

For parity polytope:

- ▶ $\text{conv}(X) = \text{conv}\left(\bigcup_{k \text{ even}} \text{conv}\left(\{x \in \{0, 1\}^n : \sum_{i=1}^n x_i = k\}\right)\right)$

One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi: \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .
- ▶ Alternative viewpoint: model using additional variables

Theorem (Balas, 1979)

Let $P_1, \dots, P_k \subseteq \mathbb{R}^n$ be polytopes. Then $\text{xc}(P_1 \cup \dots \cup P_k) \leq \sum_{i=1}^k (\text{xc}(P_i) + 1)$.

For parity polytope:

- ▶ $\text{conv}(X) = \text{conv}\left(\bigcup_{k \text{ even}} \{x \in [0, 1]^n : \sum_{i=1}^n x_i = k\}\right)$

One drawback of the polyhedral method:

- ▶ Consider $X := \{x \in \{0, 1\}^n : \sum_{i=1}^n x_i \text{ even}\}$.
- ▶ Optimization is easy: first over $\{0, 1\}^n$, potentially flip 1 coordinate.
- ▶ Inequality description (Jeroslow, 1975) requires 2^{n-1} inequalities:

$$\sum_{i \in I} (1 - x_i) + \sum_{i \notin I} x_i \geq 1 \text{ for all } I \subseteq [n] \text{ with } |I| \text{ odd}$$

Potential cure: extended formulations

- ▶ $P = \text{conv}(X)$ has many facets, but maybe there exists an **extension** (Q, π) ($Q \subseteq \mathbb{R}^d$, $\pi: \mathbb{R}^d \rightarrow \mathbb{R}^n$ linear with $P = \pi(Q)$) with **few facets**?
- ▶ The **extension complexity** $\text{xc}(P)$ of P is the minimum number of facets of an extension (Q, π) of P .
- ▶ Alternative viewpoint: model using additional variables

Theorem (Balas, 1979)

Let $P_1, \dots, P_k \subseteq \mathbb{R}^n$ be polytopes. Then $\text{xc}(P_1 \cup \dots \cup P_k) \leq \sum_{i=1}^k (\text{xc}(P_i) + 1)$.

For parity polytope:

- ▶ $\text{conv}(X) = \text{conv}\left(\bigcup_{k \text{ even}} \{x \in [0, 1]^n : \sum_{i=1}^n x_i = k\}\right)$
- ▶ Applying the theorem: $\text{xc}(\text{conv}(X)) \leq \mathcal{O}(n^2)$

Hard problems:

- ▶ Max-Cut problem: cut polytope for K_n (complete graph with n nodes) has extension complexity $2^{\Omega(n)}$ (Fiorini, Massar, Pokutta, Tiwary & de Wolf, 2012), best bound is 1.5^n (Kaibel & Weltge, 2013).

Hard problems:

- ▶ Max-Cut problem: cut polytope for K_n (complete graph with n nodes) has extension complexity $2^{\Omega(n)}$ (Fiorini, Massar, Pokutta, Tiwary & de Wolf, 2012), best bound is 1.5^n (Kaibel & Weltge, 2013).
- ▶ Lots of other hard problems inherit lower bound:
 - ▶ If F is face of P , then $\text{xc}(F) \leq \text{xc}(P)$.
 - ▶ For linear maps π we have $\text{xc}(\pi(P)) \leq \text{xc}(P)$.

Hard problems:

- ▶ Max-Cut problem: cut polytope for K_n (complete graph with n nodes) has extension complexity $2^{\Omega(n)}$ (Fiorini, Massar, Pokutta, Tiwary & de Wolf, 2012), best bound is 1.5^n (Kaibel & Weltge, 2013).
- ▶ Lots of other hard problems inherit lower bound:
 - ▶ If F is face of P , then $\text{xc}(F) \leq \text{xc}(P)$.
 - ▶ For linear maps π we have $\text{xc}(\pi(P)) \leq \text{xc}(P)$.
- ▶ Based on Karp reductions, write cut polytope as projection of a face of your favorite polytope (TSP, Stable set, 3d matching, etc.).

Hard problems:

- ▶ Max-Cut problem: cut polytope for K_n (complete graph with n nodes) has extension complexity $2^{\Omega(n)}$ (Fiorini, Massar, Pokutta, Tiwary & de Wolf, 2012), best bound is 1.5^n (Kaibel & Weltge, 2013).
- ▶ Lots of other hard problems inherit lower bound:
 - ▶ If F is face of P , then $\text{xc}(F) \leq \text{xc}(P)$.
 - ▶ For linear maps π we have $\text{xc}(\pi(P)) \leq \text{xc}(P)$.
- ▶ Based on Karp reductions, write cut polytope as projection of a face of your favorite polytope (TSP, Stable set, 3d matching, etc.).

Matching:

- ▶ A perfect matching in a graph $G = (V, E)$ is a set $M \subseteq E$ with $|M \cap \delta(v)| = 1$.
- ▶ The weighted perfect matching problem can be solved in polynomial time (Edmonds, 1965).

Hard problems:

- ▶ Max-Cut problem: cut polytope for K_n (complete graph with n nodes) has extension complexity $2^{\Omega(n)}$ (Fiorini, Massar, Pokutta, Tiwary & de Wolf, 2012), best bound is 1.5^n (Kaibel & Weltge, 2013).
- ▶ Lots of other hard problems inherit lower bound:
 - ▶ If F is face of P , then $\text{xc}(F) \leq \text{xc}(P)$.
 - ▶ For linear maps π we have $\text{xc}(\pi(P)) \leq \text{xc}(P)$.
- ▶ Based on Karp reductions, write cut polytope as projection of a face of your favorite polytope (TSP, Stable set, 3d matching, etc.).

Matching:

- ▶ A perfect matching in a graph $G = (V, E)$ is a set $M \subseteq E$ with $|M \cap \delta(v)| = 1$.
- ▶ The weighted perfect matching problem can be solved in polynomial time (Edmonds, 1965).

Theorem (Rothvoss, 2013)

For every even n , $\text{xc}(P_{\text{pmatch}}(n)) \geq 2^{\Omega(n)}$. Here, $P_{\text{pmatch}}(n)$ denotes the perfect matching polytope for K_n .

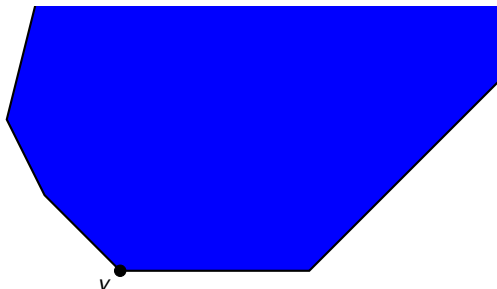
Polyhedral version of the augmentation problem:

- ▶ Consider a polyhedron $P = \{x \in \mathbb{R}^n : Ax \leq b\}$ and an objective vector $c \in \mathbb{R}^n$.
- ▶ Given a point $v \in P$, determine optimality or find improving direction $d \in \mathbb{R}^n$, i.e., $\langle c, d \rangle < 0$ and $v + d \in P$.

Polyhedral version of the augmentation problem:

- ▶ Consider a polyhedron $P = \{x \in \mathbb{R}^n : Ax \leq b\}$ and an objective vector $c \in \mathbb{R}^n$.
- ▶ Given a point $v \in P$, determine optimality or find improving direction $d \in \mathbb{R}^n$, i.e., $\langle c, d \rangle < 0$ and $v + d \in P$.
- ▶ The polyhedron for this task is the **radial cone**:

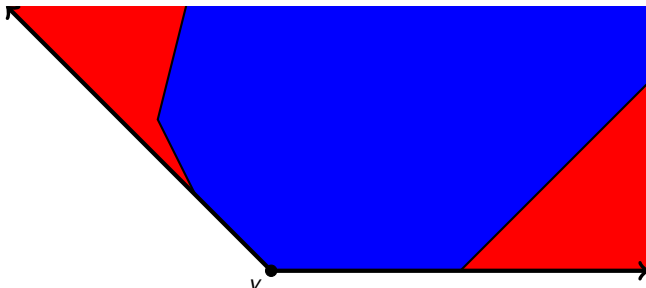
$$\begin{aligned}
 K_P(v) &:= \text{cone}(P - v) + v \\
 &= \{x \in \mathbb{R}^n : A_{i,*}x \leq b_i \text{ for all } i \text{ with } A_{*,i}v = b_i\}
 \end{aligned}$$



Polyhedral version of the augmentation problem:

- ▶ Consider a polyhedron $P = \{x \in \mathbb{R}^n : Ax \leq b\}$ and an objective vector $c \in \mathbb{R}^n$.
- ▶ Given a point $v \in P$, determine optimality or find improving direction $d \in \mathbb{R}^n$, i.e., $\langle c, d \rangle < 0$ and $v + d \in P$.
- ▶ The polyhedron for this task is the **radial cone**:

$$\begin{aligned}
 K_P(v) &:= \text{cone}(P - v) + v \\
 &= \{x \in \mathbb{R}^n : A_{i,*}x \leq b_i \text{ for all } i \text{ with } A_{*,i}v = b_i\}
 \end{aligned}$$



Nice problems:

- ▶ For $v \in P$ we have $xc(K_P(v)) \leq xc(P)$.
- ▶ Consequence: nice polyhedra have nice radial cones.

Nice problems:

- ▶ For $v \in P$ we have $xc(K_P(v)) \leq xc(P)$.
- ▶ Consequence: nice polyhedra have nice radial cones.

Hard problems:

- ▶ Braun, Fiorini, Pokutta & Steurer showed that also the cut cone (radial cone of the cut polytope at vertex $\textcircled{1}$) has exponential extension complexity.
- ▶ Extension complexity of radial cones is inherited to projections and faces.

Nice problems:

- ▶ For $v \in P$ we have $xc(K_P(v)) \leq xc(P)$.
- ▶ Consequence: nice polyhedra have nice radial cones.

Hard problems:

- ▶ Braun, Fiorini, Pokutta & Steurer showed that also the cut cone (radial cone of the cut polytope at vertex $\textcircled{1}$) has exponential extension complexity.
- ▶ Extension complexity of radial cones is inherited to projections and faces.
- ▶ Consequence: exponential lower bounds for your favorite polytopes (TSP, Stable set, 3d matching, etc.) that correspond to hard problems.

Nice problems:

- ▶ For $v \in P$ we have $xc(K_P(v)) \leq xc(P)$.
- ▶ Consequence: nice polyhedra have nice radial cones.

Hard problems:

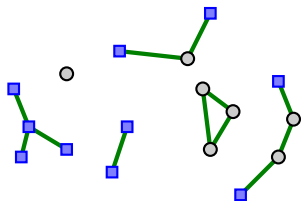
- ▶ Braun, Fiorini, Pokutta & Steurer showed that also the cut cone (radial cone of the cut polytope at vertex $\textcircled{1}$) has exponential extension complexity.
- ▶ Extension complexity of radial cones is inherited to projections and faces.
- ▶ Consequence: exponential lower bounds for your favorite polytopes (TSP, Stable set, 3d matching, etc.) that correspond to hard problems.

What remains?

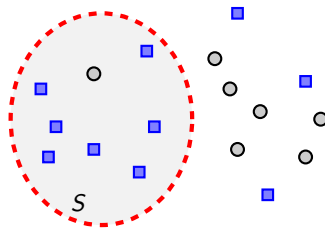
- ▶ Matching polytopes & friends ([this talk](#)).
- ▶ Stable-set polytopes of claw-free or perfect graphs.

Definitions ($K_n = (V_n, E_n)$ complete graph on n nodes; $T \subseteq V$, $|T|$ even):

- ▶ $J \subseteq E$ is a **T-join** if
 $|J \cap \delta(v)|$ is odd $\iff v \in T$

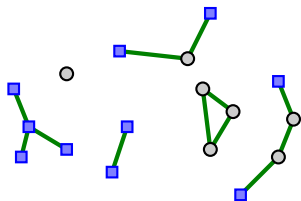


- ▶ $C = \delta(S) \subseteq E$ is a **T-cut** if
 $|S \cap T|$ is odd.

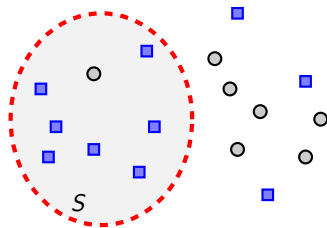


Definitions ($K_n = (V_n, E_n)$ complete graph on n nodes; $T \subseteq V$, $|T|$ even):

- ▶ $J \subseteq E$ is a **T-join** if $|J \cap \delta(v)|$ is odd $\iff v \in T$



- ▶ $C = \delta(S) \subseteq E$ is a **T-cut** if $|S \cap T|$ is odd.



Facts:

- ▶ Both minimization problems can be solved in polynomial time for $c \geq 0$.
- ▶ Each **T-join** J intersects each **T-cut** C in at least one edge:

$$|J \cap C| = \langle \chi(J), \chi(C) \rangle \geq 1$$

Polyhedra (Edmonds & Johnson, 1973):

▶ T -join Polyhedron $P_{T\text{-join}}(n)^\dagger$:

$$\begin{aligned} \langle \chi(C), x \rangle &\geq 1 && \text{for each } T\text{-cut } C \\ x_e &\geq 0 && \text{for each } e \in E \end{aligned}$$

▶ T -cut Polyhedron $P_{T\text{-cut}}(n)^\dagger$:

$$\begin{aligned} \langle \chi(J), x \rangle &\geq 1 && \text{for each } T\text{-join } J \\ x_e &\geq 0 && \text{for each } e \in E \end{aligned}$$

Polyhedra (Edmonds & Johnson, 1973):

▶ T -join Polyhedron $P_{T\text{-join}}(n)^\dagger$:

$$\langle \chi(C), x \rangle \geq 1 \quad \text{for each } T\text{-cut } C$$

$$x_e \geq 0 \quad \text{for each } e \in E$$

▶ T -cut Polyhedron $P_{T\text{-cut}}(n)^\dagger$:

$$\langle \chi(J), x \rangle \geq 1 \quad \text{for each } T\text{-join } J$$

$$x_e \geq 0 \quad \text{for each } e \in E$$

Relation to perfect matchings:

- ▶ A T -join $J \subseteq E$ is a perfect matching on nodes T if and only if $x = \chi(J)$ satisfies the valid inequalities $x_e \geq 0$ for all $e \in E \setminus E[T]$ and $\sum_{e \in \delta(v)} x_e \geq 1$ for all $v \in T$ with equality.

Polyhedra (Edmonds & Johnson, 1973):

▶ **T-join Polyhedron** $P_{T\text{-join}}(n)^\uparrow$:

$$\langle \chi(C), x \rangle \geq 1 \quad \text{for each } T\text{-cut } C$$

$$x_e \geq 0 \quad \text{for each } e \in E$$

▶ **T-cut Polyhedron** $P_{T\text{-cut}}(n)^\uparrow$:

$$\langle \chi(J), x \rangle \geq 1 \quad \text{for each } T\text{-join } J$$

$$x_e \geq 0 \quad \text{for each } e \in E$$

Relation to perfect matchings:

- ▶ A T -join $J \subseteq E$ is a perfect matching on nodes T if and only if $x = \chi(J)$ satisfies the valid inequalities $x_e \geq 0$ for all $e \in E \setminus E[T]$ and $\sum_{e \in \delta(v)} x_e \geq 1$ for all $v \in T$ with equality.
- ▶ Thus, $P_{T\text{-join}}(n)^\uparrow$ contains $P_{\text{pmatch}}(|T|)$ as a face.

Polyhedra (Edmonds & Johnson, 1973):

▶ **T-join Polyhedron** $P_{T\text{-join}}(n)^\uparrow$:

$$\langle \chi(C), x \rangle \geq 1 \quad \text{for each } T\text{-cut } C$$

$$x_e \geq 0 \quad \text{for each } e \in E$$

▶ **T-cut Polyhedron** $P_{T\text{-cut}}(n)^\uparrow$:

$$\langle \chi(J), x \rangle \geq 1 \quad \text{for each } T\text{-join } J$$

$$x_e \geq 0 \quad \text{for each } e \in E$$

Relation to perfect matchings:

- ▶ A T -join $J \subseteq E$ is a perfect matching on nodes T if and only if $x = \chi(J)$ satisfies the valid inequalities $x_e \geq 0$ for all $e \in E \setminus E[T]$ and $\sum_{e \in \delta(v)} x_e \geq 1$ for all $v \in T$ with equality.
- ▶ Thus, $P_{T\text{-join}}(n)^\uparrow$ contains $P_{\text{pmatch}}(|T|)$ as a face.
- ▶ Consequence: $\text{xc}(P_{T\text{-join}}(n)^\uparrow) \geq 2^{\Omega(|T|)}$

Polyhedra (Edmonds & Johnson, 1973):

▶ **T-join Polyhedron** $P_{T\text{-join}}(n)^\uparrow$:

$$\langle \chi(C), x \rangle \geq 1 \quad \text{for each } T\text{-cut } C$$

$$x_e \geq 0 \quad \text{for each } e \in E$$

▶ **T-cut Polyhedron** $P_{T\text{-cut}}(n)^\uparrow$:

$$\langle \chi(J), x \rangle \geq 1 \quad \text{for each } T\text{-join } J$$

$$x_e \geq 0 \quad \text{for each } e \in E$$

Relation to perfect matchings:

- ▶ A T -join $J \subseteq E$ is a perfect matching on nodes T if and only if $x = \chi(J)$ satisfies the valid inequalities $x_e \geq 0$ for all $e \in E \setminus E[T]$ and $\sum_{e \in \delta(v)} x_e \geq 1$ for all $v \in T$ with equality.
- ▶ Thus, $P_{T\text{-join}}(n)^\uparrow$ contains $P_{\text{pmatch}}(|T|)$ as a face.
- ▶ Consequence: $\text{xc}(P_{T\text{-join}}(n)^\uparrow) \geq 2^{\Omega(|T|)}$

Proposition (Walter & Weltge, 2018)

For every n and every set $T \subseteq V_n$, $\text{xc}(P_{T\text{-join}}(n)^\uparrow) \leq \mathcal{O}(n^2 \cdot 2^{|T|})$.

Idea:

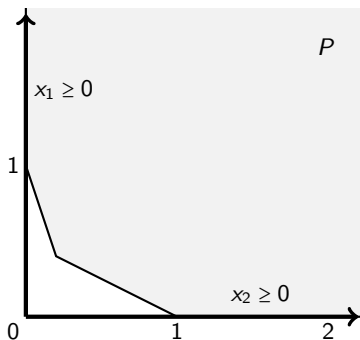
- ▶ For each $S \subseteq T$ with $|S| = \frac{1}{2}|T|$, consider the b -flow polyhedron for $b_v = -1$ for all $v \in S$, $b_v = 1$ for all $v \in T \setminus S$ and $b_v = 0$ otherwise.
- ▶ Apply disjunctive programming over all such polyhedra.

Definitions:

- ▶ A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- ▶ Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$



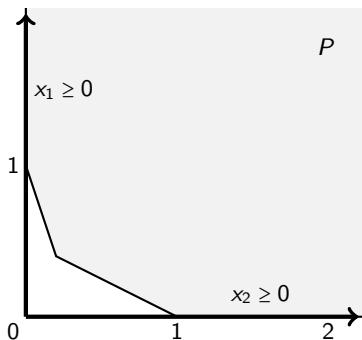
Definitions:

- ▶ A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- ▶ Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$

- ▶ The **blocker** of P is defined via $B(P) := \{y \in \mathbb{R}_+^d : \langle x, y \rangle \geq 1 \text{ for all } x \in P\}$.



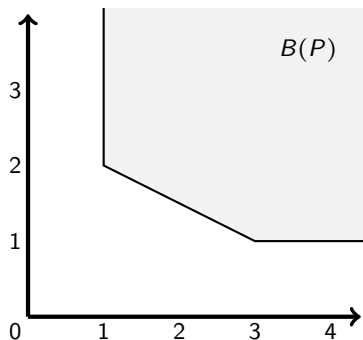
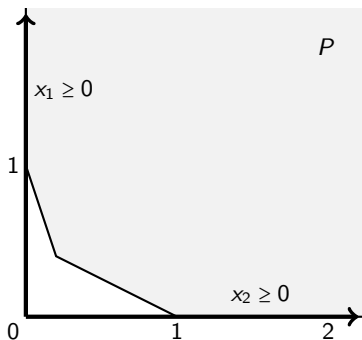
Definitions:

- ▶ A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- ▶ Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$

- ▶ The **blocker** of P is defined via $B(P) := \{y \in \mathbb{R}_+^d : \langle x, y \rangle \geq 1 \text{ for all } x \in P\}$.



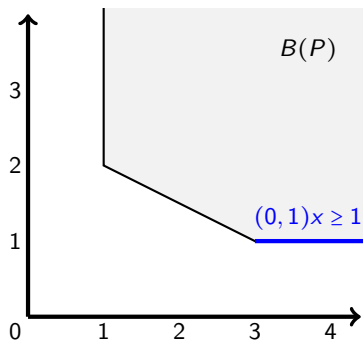
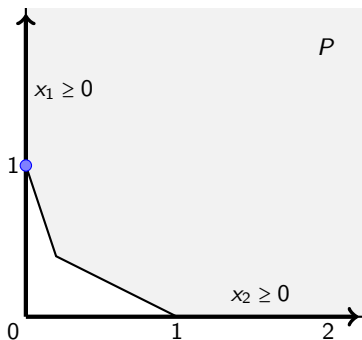
Definitions:

- ▶ A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- ▶ Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$

- ▶ The **blocker** of P is defined via $B(P) := \{y \in \mathbb{R}_+^d : \langle x, y \rangle \geq 1 \text{ for all } x \in P\}$.



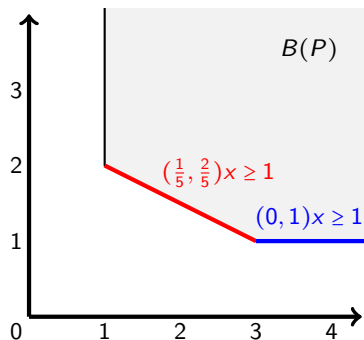
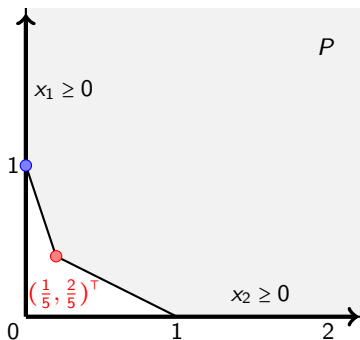
Definitions:

- ▶ A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- ▶ Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$

- ▶ The **blocker** of P is defined via $B(P) := \{y \in \mathbb{R}_+^d : \langle x, y \rangle \geq 1 \text{ for all } x \in P\}$.



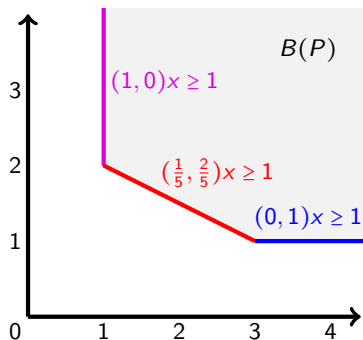
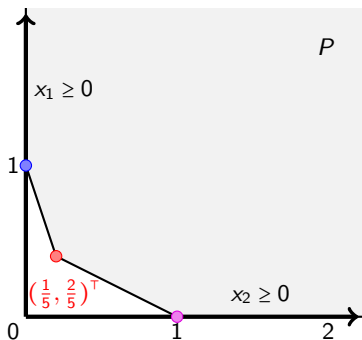
Definitions:

- ▶ A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- ▶ Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$

- ▶ The **blocker** of P is defined via $B(P) := \{y \in \mathbb{R}_+^d : \langle x, y \rangle \geq 1 \text{ for all } x \in P\}$.



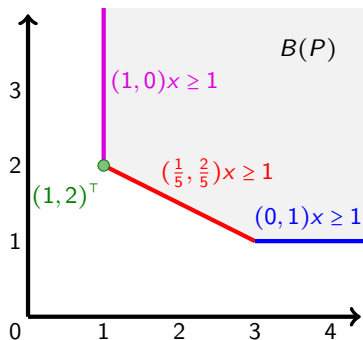
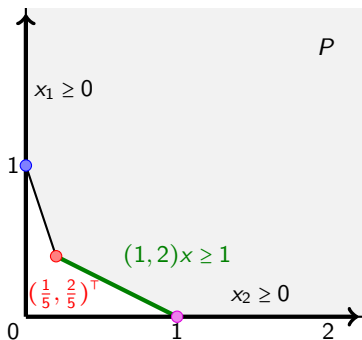
Definitions:

- ▶ A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- ▶ Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$

- ▶ The **blocker** of P is defined via $B(P) := \{y \in \mathbb{R}_+^d : \langle x, y \rangle \geq 1 \text{ for all } x \in P\}$.



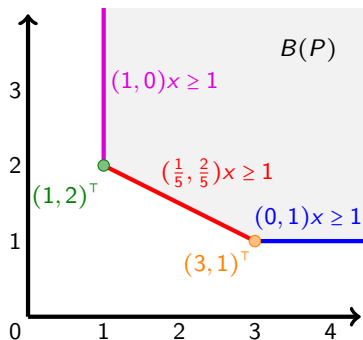
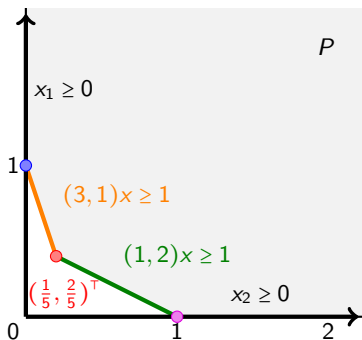
Definitions:

- ▶ A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- ▶ Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$

- ▶ The **blocker** of P is defined via $B(P) := \{y \in \mathbb{R}_+^d : \langle x, y \rangle \geq 1 \text{ for all } x \in P\}$.



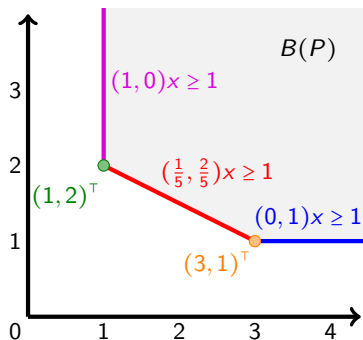
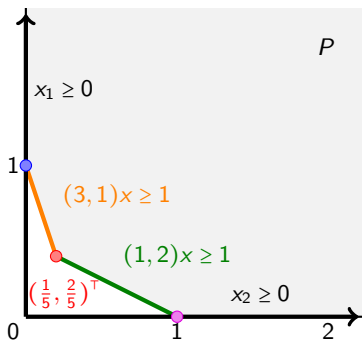
Definitions:

- A polyhedron $P \subseteq \mathbb{R}_+^d$ is **blocking** if $x' \geq x$ implies $x' \in P$ for all $x \in P$.
- Possible descriptions are:

$$P = \{x \in \mathbb{R}_+^d : \langle y^{(i)}, x \rangle \geq 1 \text{ for } i = 1, \dots, m\} \quad (y^{(1)}, \dots, y^{(m)} \in \mathbb{R}_+^d)$$

$$P = \text{conv}\{x^{(1)}, \dots, x^{(k)}\} + \mathbb{R}_+^d \quad (x^{(1)}, \dots, x^{(k)} \in \mathbb{R}_+^d)$$

- The **blocker** of P is defined via $B(P) := \{y \in \mathbb{R}_+^d : \langle x, y \rangle \geq 1 \text{ for all } x \in P\}$.
- If P is blocking, then $B(B(P)) = P$.



Proposition (Martin, 1991; Conforti, Kaibel, Walter & Weltge, 2015)

Given a non-empty polyhedron Q and $\gamma \in \mathbb{R}$, let

$$P := \{x : \langle y, x \rangle \geq \gamma \text{ for all } y \in Q\}.$$

Then $\text{xc}(P) \leq \text{xc}(Q) + 1$.

Proof:

- ▶ Let $Q = \{Tz : Az \leq b\}$, where A has $m = \text{xc}(Q)$ rows.

Proposition (Martin, 1991; Conforti, Kaibel, Walter & Weltge, 2015)

Given a non-empty polyhedron Q and $\gamma \in \mathbb{R}$, let

$$P := \{x : \langle y, x \rangle \geq \gamma \text{ for all } y \in Q\}.$$

Then $\text{xc}(P) \leq \text{xc}(Q) + 1$.

Proof:

- ▶ Let $Q = \{Tz : Az \leq b\}$, where A has $m = \text{xc}(Q)$ rows.

$$\hat{x} \in P \iff \min \{\langle \hat{x}, y \rangle : y \in Q\} \geq \gamma$$

Proposition (Martin, 1991; Conforti, Kaibel, Walter & Weltge, 2015)

Given a non-empty polyhedron Q and $\gamma \in \mathbb{R}$, let

$$P := \{x : \langle y, x \rangle \geq \gamma \text{ for all } y \in Q\}.$$

Then $\text{xc}(P) \leq \text{xc}(Q) + 1$.

Proof:

- ▶ Let $Q = \{Tz : Az \leq b\}$, where A has $m = \text{xc}(Q)$ rows.

$$\hat{x} \in P \iff \min \{\langle \hat{x}, y \rangle : y \in Q\} \geq \gamma$$

$$\iff \min \{\langle \hat{x}, Tz \rangle : Az \leq b\} \geq \gamma$$

Proposition (Martin, 1991; Conforti, Kaibel, Walter & Weltge, 2015)

Given a non-empty polyhedron Q and $\gamma \in \mathbb{R}$, let

$$P := \{x : \langle y, x \rangle \geq \gamma \text{ for all } y \in Q\}.$$

Then $\text{xc}(P) \leq \text{xc}(Q) + 1$.

Proof:

- ▶ Let $Q = \{Tz : Az \leq b\}$, where A has $m = \text{xc}(Q)$ rows.

$$\hat{x} \in P \iff \min \{\langle \hat{x}, y \rangle : y \in Q\} \geq \gamma$$

$$\iff \min \{\langle \hat{x}, Tz \rangle : Az \leq b\} \geq \gamma$$

$$\iff \max \{\langle b, \lambda \rangle : A^T \lambda = T^T \hat{x}, \lambda \leq \mathbb{0}\} \geq \gamma$$

Proposition (Martin, 1991; Conforti, Kaibel, Walter & Weltge, 2015)

Given a non-empty polyhedron Q and $\gamma \in \mathbb{R}$, let

$$P := \{x : \langle y, x \rangle \geq \gamma \text{ for all } y \in Q\}.$$

Then $\text{xc}(P) \leq \text{xc}(Q) + 1$.

Proof:

- ▶ Let $Q = \{Tz : Az \leq b\}$, where A has $m = \text{xc}(Q)$ rows.

$$\hat{x} \in P \iff \min \{\langle \hat{x}, y \rangle : y \in Q\} \geq \gamma$$

$$\iff \min \{\langle \hat{x}, Tz \rangle : Az \leq b\} \geq \gamma$$

$$\iff \max \{\langle b, \lambda \rangle : A^T \lambda = T^T \hat{x}, \lambda \leq \mathbb{0}\} \geq \gamma$$

$$\iff \exists \lambda \leq \mathbb{0} : A^T \lambda = T^T \hat{x}, \langle b, \lambda \rangle \geq \gamma$$

Proposition (Martin, 1991; Conforti, Kaibel, Walter & Weltge, 2015)

Given a non-empty polyhedron Q and $\gamma \in \mathbb{R}$, let

$$P := \{x : \langle y, x \rangle \geq \gamma \text{ for all } y \in Q\}.$$

Then $\text{xc}(P) \leq \text{xc}(Q) + 1$.

Proof:

- ▶ Let $Q = \{Tz : Az \leq b\}$, where A has $m = \text{xc}(Q)$ rows.

$$\hat{x} \in P \iff \min \{\langle \hat{x}, y \rangle : y \in Q\} \geq \gamma$$

$$\iff \min \{\langle \hat{x}, Tz \rangle : Az \leq b\} \geq \gamma$$

$$\iff \max \{\langle b, \lambda \rangle : A^T \lambda = T^T \hat{x}, \lambda \leq \mathbb{0}\} \geq \gamma$$

$$\iff \exists \lambda \leq \mathbb{0} : A^T \lambda = T^T \hat{x}, \langle b, \lambda \rangle \geq \gamma$$

- ▶ Thus, $P = \{x : \exists \lambda \leq \mathbb{0} : A^T \lambda = T^T x, \langle b, \lambda \rangle \geq \gamma\}$ is an extension with $m + 1$ inequalities.

Proposition (Martin, 1991; Conforti, Kaibel, Walter & Weltge, 2015)

Given a non-empty polyhedron Q and $\gamma \in \mathbb{R}$, let

$$P := \{x : \langle y, x \rangle \geq \gamma \text{ for all } y \in Q\}.$$

Then $\text{xc}(P) \leq \text{xc}(Q) + 1$.

Proof:

- Let $Q = \{Tz : Az \leq b\}$, where A has $m = \text{xc}(Q)$ rows.

$$\hat{x} \in P \iff \min \{\langle \hat{x}, y \rangle : y \in Q\} \geq \gamma$$

$$\iff \min \{\langle \hat{x}, Tz \rangle : Az \leq b\} \geq \gamma$$

$$\iff \max \{\langle b, \lambda \rangle : A^T \lambda = T^T \hat{x}, \lambda \leq \mathbb{0}\} \geq \gamma$$

$$\iff \exists \lambda \leq \mathbb{0} : A^T \lambda = T^T \hat{x}, \langle b, \lambda \rangle \geq \gamma$$

- Thus, $P = \{x : \exists \lambda \leq \mathbb{0} : A^T \lambda = T^T x, \langle b, \lambda \rangle \geq \gamma\}$ is an extension with $m + 1$ inequalities.

Consequences:

- $\text{xc}(B(P))$ and $\text{xc}(P)$ differ by at most d .

Proposition (Martin, 1991; Conforti, Kaibel, Walter & Weltge, 2015)

Given a non-empty polyhedron Q and $\gamma \in \mathbb{R}$, let

$$P := \{x : \langle y, x \rangle \geq \gamma \text{ for all } y \in Q\}.$$

Then $\text{xc}(P) \leq \text{xc}(Q) + 1$.

Proof:

- Let $Q = \{Tz : Az \leq b\}$, where A has $m = \text{xc}(Q)$ rows.

$$\hat{x} \in P \iff \min \{\langle \hat{x}, y \rangle : y \in Q\} \geq \gamma$$

$$\iff \min \{\langle \hat{x}, Tz \rangle : Az \leq b\} \geq \gamma$$

$$\iff \max \{\langle b, \lambda \rangle : A^T \lambda = T^T \hat{x}, \lambda \leq \mathbb{0}\} \geq \gamma$$

$$\iff \exists \lambda \leq \mathbb{0} : A^T \lambda = T^T \hat{x}, \langle b, \lambda \rangle \geq \gamma$$

- Thus, $P = \{x : \exists \lambda \leq \mathbb{0} : A^T \lambda = T^T x, \langle b, \lambda \rangle \geq \gamma\}$ is an extension with $m + 1$ inequalities.

Consequences:

- $\text{xc}(B(P))$ and $\text{xc}(P)$ differ by at most d .
- $2^{\Omega(|T|)} \leq \text{xc}(P_{T\text{-cut}}(n)^\dagger) \leq \mathcal{O}(n^2 \cdot 2^{|T|})$.

Polar object of radial cone:

- ▶ Any $v \in P$ defines a face $F_{B(P)}(v) := \{y \in B(P) : \langle v, y \rangle = 1\}$ of $B(P)$.

Polar object of radial cone:

- Any $v \in P$ defines a face $F_{B(P)}(v) := \{y \in B(P) : \langle v, y \rangle = 1\}$ of $B(P)$.

Lemma

Let $P \subseteq \mathbb{R}_+^d$ be a blocking polyhedron and let $v \in P$.

- (i) $F_{B(P)}(v) = \{y \in \mathbb{R}^d : \langle v, y \rangle = 1, \langle x, y \rangle \geq 1 \ \forall x \in K_P(v)\}$.
- (ii) $K_P(v) = \{x \in \mathbb{R}^d : \langle y, x \rangle \geq 1 \ \forall y \in F_{B(P)}(v)\}$.

Proof:

- Elementary convex geometry

Polar object of radial cone:

- Any $v \in P$ defines a face $F_{B(P)}(v) := \{y \in B(P) : \langle v, y \rangle = 1\}$ of $B(P)$.

Lemma

Let $P \subseteq \mathbb{R}_+^d$ be a blocking polyhedron and let $v \in P$.

- (i) $F_{B(P)}(v) = \{y \in \mathbb{R}^d : \langle v, y \rangle = 1, \langle x, y \rangle \geq 1 \ \forall x \in K_P(v)\}$.
- (ii) $K_P(v) = \{x \in \mathbb{R}^d : \langle y, x \rangle \geq 1 \ \forall y \in F_{B(P)}(v)\}$.

Proof:

- Elementary convex geometry

Consequence:

- $\text{xc}(K_P(v))$ and $\text{xc}(F_{B(P)}(v))$ differ by at most 1.
- To prove lower **or** upper bounds on $\text{xc}(K_P(v))$, it suffices to do analyze the face $F_{B(P)}(v)$ of $B(P)$.

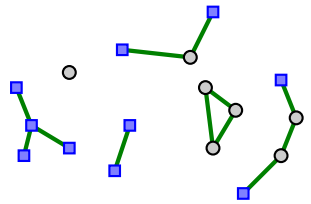
Theorem (Ventura & Eisenbrand, 2003)

For every set $T \subseteq V_n$ with $|T|$ even and every vertex v of $P_{T\text{-join}}(n)^\uparrow$, corresponding to a T -join $J \subseteq E_n$ in K_n , the extension complexity of the radial cone of $P_{T\text{-join}}(n)$ at v is most $\mathcal{O}(|J| \cdot n^2)$.

Theorem (Ventura & Eisenbrand, 2003)

For every set $T \subseteq V_n$ with $|T|$ even and every vertex v of $P_{T\text{-join}}(n)^\dagger$, corresponding to a T -join $J \subseteq E_n$ in K_n , the extension complexity of the radial cone of $P_{T\text{-join}}(n)$ at v is most $\mathcal{O}(|J| \cdot n^2)$.

Our new proof:



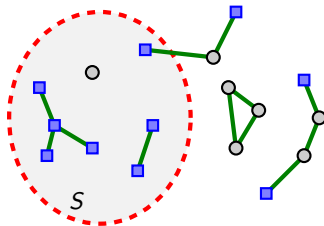
Theorem (Ventura & Eisenbrand, 2003)

For every set $T \subseteq V_n$ with $|T|$ even and every vertex v of $P_{T\text{-join}}(n)^\dagger$, corresponding to a T -join $J \subseteq E_n$ in K_n , the extension complexity of the radial cone of $P_{T\text{-join}}(n)$ at v is most $\mathcal{O}(|J| \cdot n^2)$.

Our new proof:

- By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-cut}}(n)^\dagger : \sum_{e \in J} x_e = 1 \right\}.$$



Theorem (Ventura & Eisenbrand, 2003)

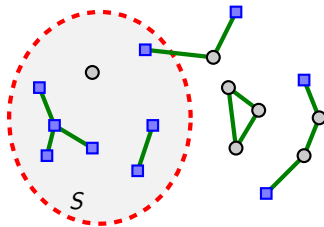
For every set $T \subseteq V_n$ with $|T|$ even and every vertex v of $P_{T\text{-join}}(n)^\dagger$, corresponding to a T -join $J \subseteq E_n$ in K_n , the extension complexity of the radial cone of $P_{T\text{-join}}(n)$ at v is most $\mathcal{O}(|J| \cdot n^2)$.

Our new proof:

- By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-cut}}(n)^\dagger : \sum_{e \in J} x_e = 1 \right\}.$$

- For each $m \in J$, let F_m be the face of P with $x_m = 1$ (and $x_e = 0 \ \forall e \in J \setminus \{m\}$).



Theorem (Ventura & Eisenbrand, 2003)

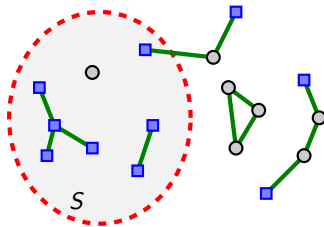
For every set $T \subseteq V_n$ with $|T|$ even and every vertex v of $P_{T\text{-join}}(n)^\dagger$, corresponding to a T -join $J \subseteq E_n$ in K_n , the extension complexity of the radial cone of $P_{T\text{-join}}(n)$ at v is most $\mathcal{O}(|J| \cdot n^2)$.

Our new proof:

- By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-cut}}(n)^\dagger : \sum_{e \in J} x_e = 1 \right\}.$$

- For each $m \in J$, let F_m be the face of P with $x_m = 1$ (and $x_e = 0 \ \forall e \in J \setminus \{m\}$).
- But F_m is also a face of $P_{T'\text{-cut}}(n)^\dagger$ for $T' = m$ (set containing the nodes).



Theorem (Ventura & Eisenbrand, 2003)

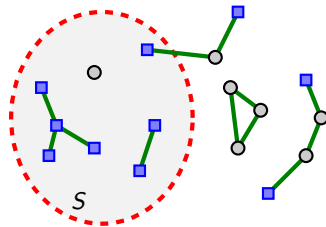
For every set $T \subseteq V_n$ with $|T|$ even and every vertex v of $P_{T\text{-join}}(n)^\dagger$, corresponding to a T -join $J \subseteq E_n$ in K_n , the extension complexity of the radial cone of $P_{T\text{-join}}(n)$ at v is most $\mathcal{O}(|J| \cdot n^2)$.

Our new proof:

- By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-cut}}(n)^\dagger : \sum_{e \in J} x_e = 1 \right\}.$$

- For each $m \in J$, let F_m be the face of P with $x_m = 1$ (and $x_e = 0 \ \forall e \in J \setminus \{m\}$).
- But F_m is also a face of $P_{T'\text{-cut}}(n)^\dagger$ for $T' = m$ (set containing the nodes).
- We obtain $\text{xc}(F_m) \leq \mathcal{O}(n^2 \cdot 2^{|T'|}) = \mathcal{O}(n^2)$.



Theorem (Ventura & Eisenbrand, 2003)

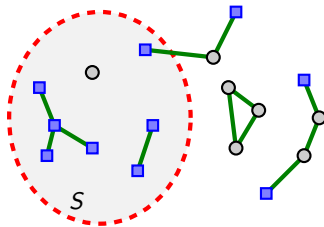
For every set $T \subseteq V_n$ with $|T|$ even and every vertex v of $P_{T\text{-join}}(n)^\dagger$, corresponding to a T -join $J \subseteq E_n$ in K_n , the extension complexity of the radial cone of $P_{T\text{-join}}(n)$ at v is most $\mathcal{O}(|J| \cdot n^2)$.

Our new proof:

- By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-cut}}(n)^\dagger : \sum_{e \in J} x_e = 1 \right\}.$$

- For each $m \in J$, let F_m be the face of P with $x_m = 1$ (and $x_e = 0 \ \forall e \in J \setminus \{m\}$).
- But F_m is also a face of $P_{T'\text{-cut}}(n)^\dagger$ for $T' = m$ (set containing the nodes).
- We obtain $\text{xc}(F_m) \leq \mathcal{O}(n^2 \cdot 2^{|T'|}) = \mathcal{O}(n^2)$.
- P is convex hull of union of all F_m .



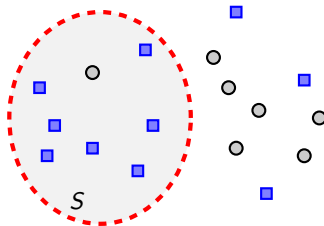
Theorem (Walter & Weltge, 2018)

For sets $T \subseteq V_n$ with $|T|$ even and vertices v of $P_{T\text{-cut}}(n)^\dagger$, the extension complexity of the radial cone of $P_{T\text{-cut}}(n)$ at v is least $2^{\Omega(|T|)}$.

Theorem (Walter & Weltge, 2018)

For sets $T \subseteq V_n$ with $|T|$ even and vertices v of $P_{T\text{-cut}}(n)^\dagger$, the extension complexity of the radial cone of $P_{T\text{-cut}}(n)$ at v is least $2^{\Omega(|T|)}$.

Proof:

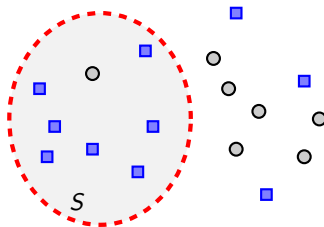


Theorem (Walter & Weltge, 2018)

For sets $T \subseteq V_n$ with $|T|$ even and vertices v of $P_{T\text{-cut}}(n)^\uparrow$, the extension complexity of the radial cone of $P_{T\text{-cut}}(n)$ at v is least $2^{\Omega(|T|)}$.

Proof:

- ▶ Let $v = \chi(\delta(S))$.



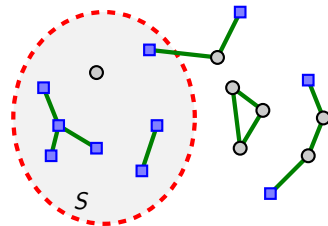
Theorem (Walter & Weltge, 2018)

For sets $T \subseteq V_n$ with $|T|$ even and vertices v of $P_{T\text{-cut}}(n)^\uparrow$, the extension complexity of the radial cone of $P_{T\text{-cut}}(n)$ at v is least $2^{\Omega(|T|)}$.

Proof:

- ▶ Let $v = \chi(\delta(S))$.
- ▶ By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-join}}(n)^\uparrow : \sum_{e \in \delta(S)} x_e = 1 \right\}.$$



Theorem (Walter & Weltge, 2018)

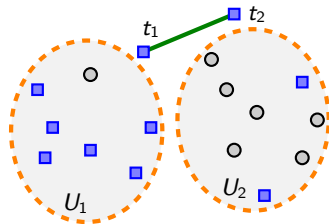
For sets $T \subseteq V_n$ with $|T|$ even and vertices v of $P_{T\text{-cut}}(n)^\dagger$, the extension complexity of the radial cone of $P_{T\text{-cut}}(n)$ at v is least $2^{\Omega(|T|)}$.

Proof:

- ▶ Let $v = \chi(\delta(S))$.
- ▶ By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-join}}(n)^\dagger : \sum_{e \in \delta(S)} x_e = 1 \right\}.$$

- ▶ Let $t_1 \in S$, $t_2 \in V_n \setminus S$ as well as
 $U_1 := S \setminus \{t_1\}$, $U_2 := (V_n \setminus (S \cup \{t_2\}))$.



Theorem (Walter & Weltge, 2018)

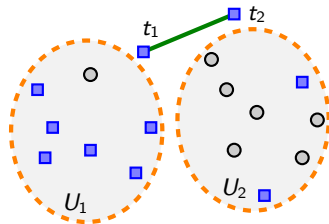
For sets $T \subseteq V_n$ with $|T|$ even and vertices v of $P_{T\text{-cut}}(n)^\dagger$, the extension complexity of the radial cone of $P_{T\text{-cut}}(n)$ at v is least $2^{\Omega(|T|)}$.

Proof:

- ▶ Let $v = \chi(\delta(S))$.
- ▶ By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-join}}(n)^\dagger : \sum_{e \in \delta(S)} x_e = 1 \right\}.$$

- ▶ Let $t_1 \in S$, $t_2 \in V_n \setminus S$ as well as $U_1 := S \setminus \{t_1\}$, $U_2 := (V_n \setminus (S \cup \{t_2\}))$.
- ▶ Let F be the face of P with $x_{\{t_1, t_2\}} = 1$ and $x_e = 0$ for all edges between U_1 , U_2 and $\{t_1, t_2\}$.



Theorem (Walter & Weltge, 2018)

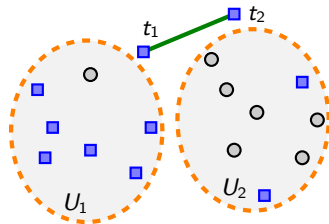
For sets $T \subseteq V_n$ with $|T|$ even and vertices v of $P_{T\text{-cut}}(n)^\dagger$, the extension complexity of the radial cone of $P_{T\text{-cut}}(n)$ at v is least $2^{\Omega(|T|)}$.

Proof:

- ▶ Let $v = \chi(\delta(S))$.
- ▶ By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-join}}(n)^\dagger : \sum_{e \in \delta(S)} x_e = 1 \right\}.$$

- ▶ Let $t_1 \in S$, $t_2 \in V_n \setminus S$ as well as $U_1 := S \setminus \{t_1\}$, $U_2 := (V_n \setminus (S \cup \{t_2\}))$.
- ▶ Let F be the face of P with $x_{\{t_1, t_2\}} = 1$ and $x_e = 0$ for all edges between U_1 , U_2 and $\{t_1, t_2\}$.
- ▶ F is a Cartesian product of a vector and two $(T \cap U_i)$ -join polyhedra on U_i for $i = 1, 2$, where $|T_1| + |T_2| = |T| - 2$.



Theorem (Walter & Weltge, 2018)

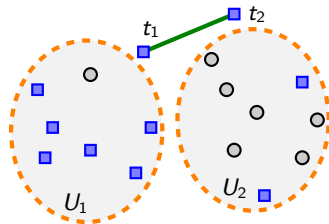
For sets $T \subseteq V_n$ with $|T|$ even and vertices v of $P_{T\text{-cut}}(n)^\dagger$, the extension complexity of the radial cone of $P_{T\text{-cut}}(n)$ at v is least $2^{\Omega(|T|)}$.

Proof:

- ▶ Let $v = \chi(\delta(S))$.
- ▶ By Lemma, theorem reduces to $\text{xc}(P)$ for

$$P := \left\{ x \in P_{T\text{-join}}(n)^\dagger : \sum_{e \in \delta(S)} x_e = 1 \right\}.$$

- ▶ Let $t_1 \in S$, $t_2 \in V_n \setminus S$ as well as $U_1 := S \setminus \{t_1\}$, $U_2 := (V_n \setminus (S \cup \{t_2\}))$.
- ▶ Let F be the face of P with $x_{\{t_1, t_2\}} = 1$ and $x_e = 0$ for all edges between U_1 , U_2 and $\{t_1, t_2\}$.
- ▶ F is a Cartesian product of a vector and two $(T \cap U_i)$ -join polyhedra on U_i for $i = 1, 2$, where $|T_1| + |T_2| = |T| - 2$.
- ▶ We obtain $\text{xc}(P) \geq \text{xc}(F) \geq 2^{\Omega(|T_i|)}$ for $i = 1, 2$.



Thanks!

Conclusion:

- ▶ Extended formulations can help, but only **sometimes**.
- ▶ Although polynomially solvable, there is no obvious way to solve the minimum-weight T -cut problem with LP techniques.

Thanks!

Conclusion:

- ▶ Extended formulations can help, but only **sometimes**.
- ▶ Although polynomially solvable, there is no obvious way to solve the minimum-weight T -cut problem with LP techniques.

Other candidates for investigation:

- ▶ Stable-set polytopes of claw-free graphs (current work with Gianpaolo Oriolo and Gautier Stauffer).
- ▶ Stable-set polytopes of perfect graphs (polyhedral description is known, but best extended formulation has $\mathcal{O}(n^{\log n})$ facets).