

Volker Kaibel

Constructing Extended Formulations

Nov 6, 2017

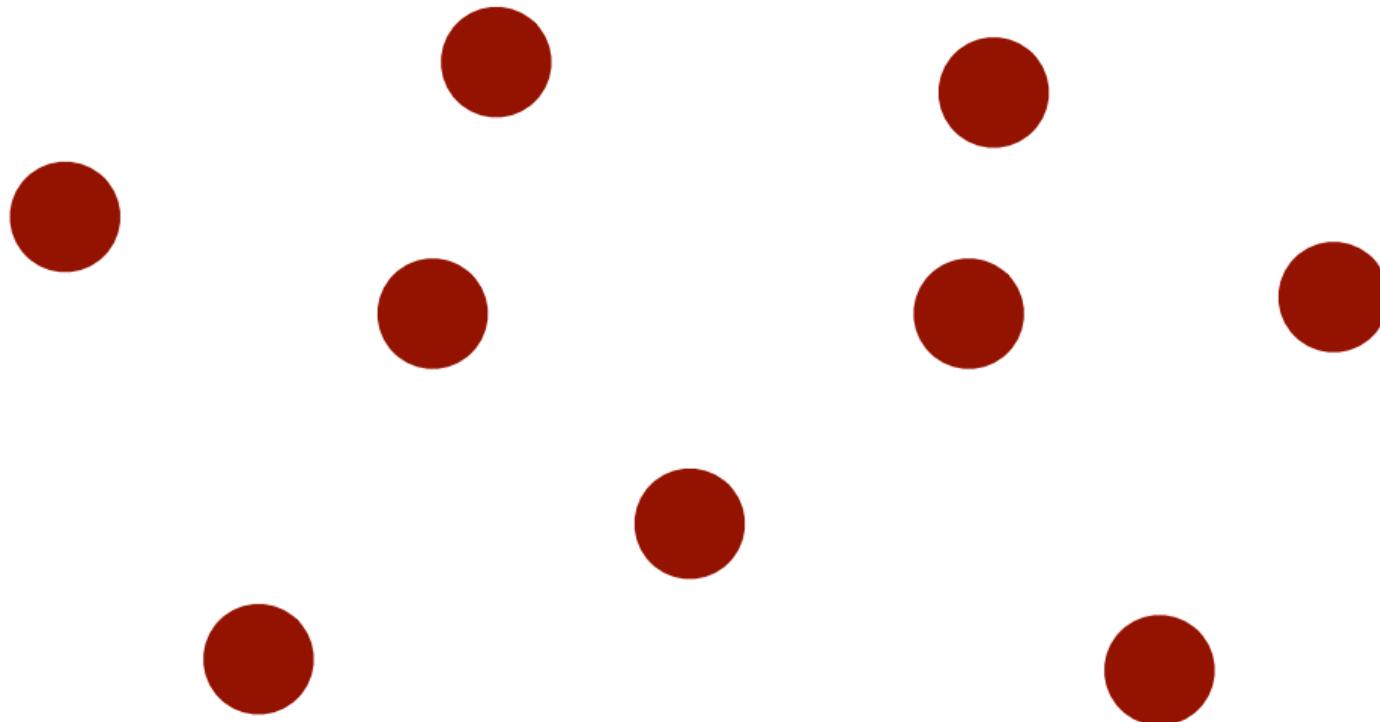
Simons Institute, Berkeley



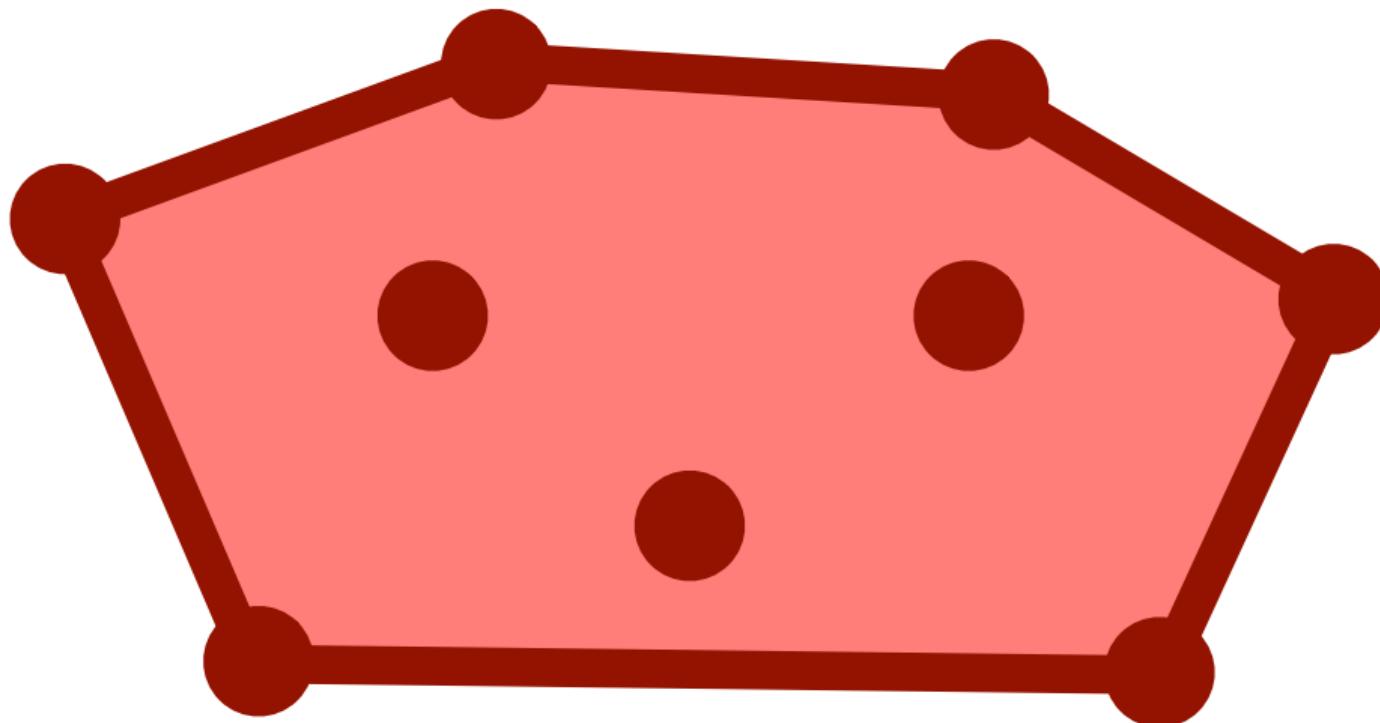
Outline

- ① The Concept
- ② Disjunctive Programming
- ③ Dynamic Programming
- ④ Branched Polyhedral Systems
- ⑤ Dualization
- ⑥ Redundant Information
- ⑦ Reflections

Convex Hulls and Linear Programming



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From points to polytopes

$$\max\{\langle c, x \rangle : x \in X\} = \max\{\langle c, x \rangle : x \in \text{conv}(X)\}$$

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For $|X| < \infty$ there is $A, b : \text{conv}(X) = \{x \in \mathbb{R}^n : Ax \leq b\}$

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$$\max\{\langle c, x \rangle : Ax \leq b, x \in \mathbb{R}^n\} = \min\{\langle b, y \rangle : A^t y = c, y \in \mathbb{R}_+^m\}$$

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LP-algorithms

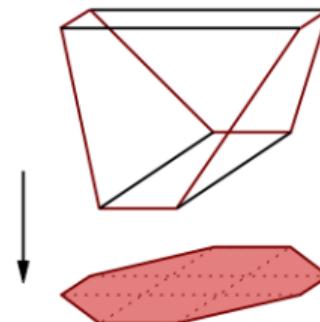
Efficient both in theory and praxis.

Representations as Projections

Extension of a Polytope $P \subseteq \mathbb{R}^n$:

A polytope $Q \subseteq \mathbb{R}^d$ and a linear projection $p : \mathbb{R}^d \rightarrow \mathbb{R}^n$ with
$$P = p(Q).$$

Size: Number of facets of Q

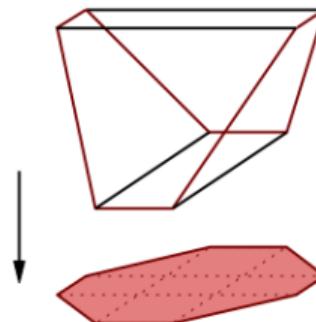


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Extended Formulation of P :

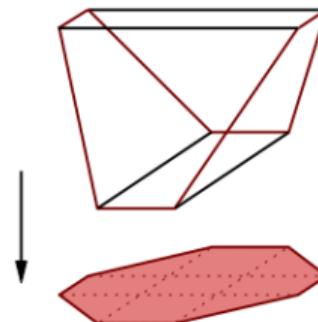
Linear description of some extension of P

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Extension Complexity of P :

$\text{xc}(P)$ = smallest size of any extension of P

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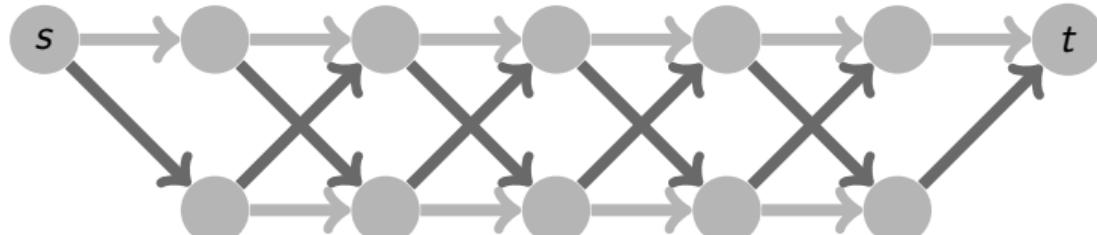
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$$\text{xc}(\text{conv}\{v \in \{0, 1\}^n : v \text{ has even } \# \text{ of 1's}\}) \leq 4n - 4$$



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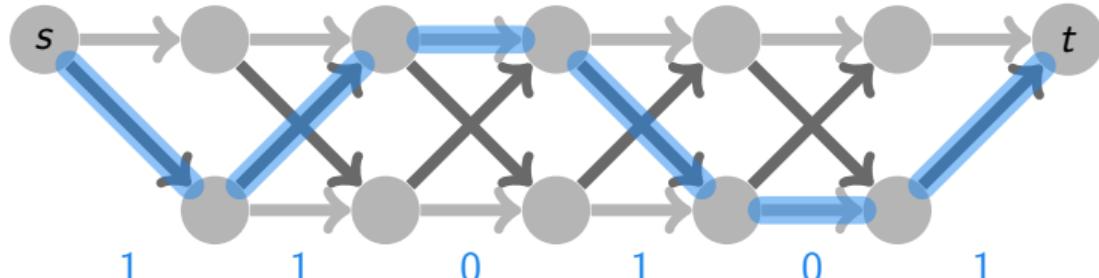
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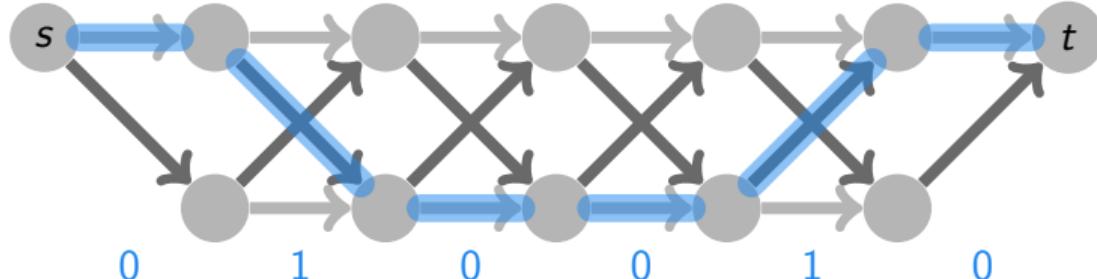
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Jobs with processing times p_1, \dots, p_n



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QUEYRANNE 1993

For $0 < p_1 \leq \dots \leq p_n$: Description by one equation and

$$\sum_{i \in I} p_i x_i \geq \sum_{i=1}^{|I|} p_i \sum_{j=1}^i p_j \quad \text{for all } \emptyset \neq I \subseteq [n]$$

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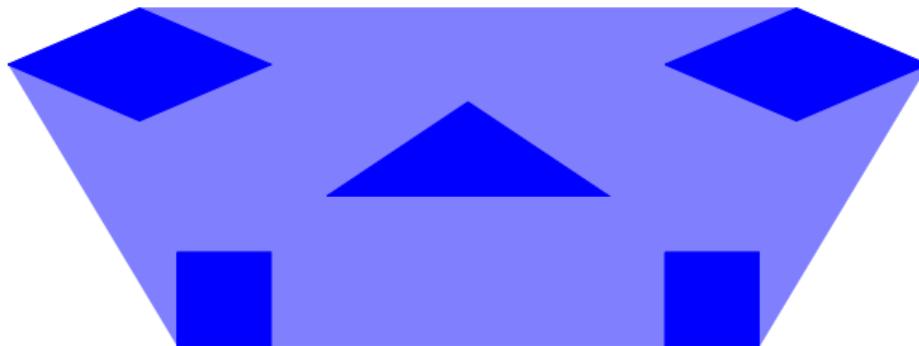
The cube $Q = [0, 1]^{\binom{[n]}{2}}$ projects to $\mathsf{P}_{\text{ct}}(p_1, \dots, p_n)$ via

$$x_i = \sum_{j=1}^{i-1} p_j y_{\{i,j\}} + \sum_{j=i+1}^n p_j (1 - y_{\{i,j\}}) \quad \text{for all } i \in [n].$$

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Unions of Polytopes



BALAS 1975

For polytopes $P_1, \dots, P_q \subseteq \mathbb{R}^m$ (with $\dim(P_i) > 0$)

$$\text{xc}(\text{conv}(\bigcup_{i=1}^q P_i)) \leq \sum_{i=1}^q \text{xc}(P_i)$$

holds.

Matching Polytopes

Matchings with ℓ edges

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EDMONDS 1965

$P_{\text{match}}^\ell(n)$ is described by $x \geq \mathbf{0}$, $x(E) = \ell$, and:

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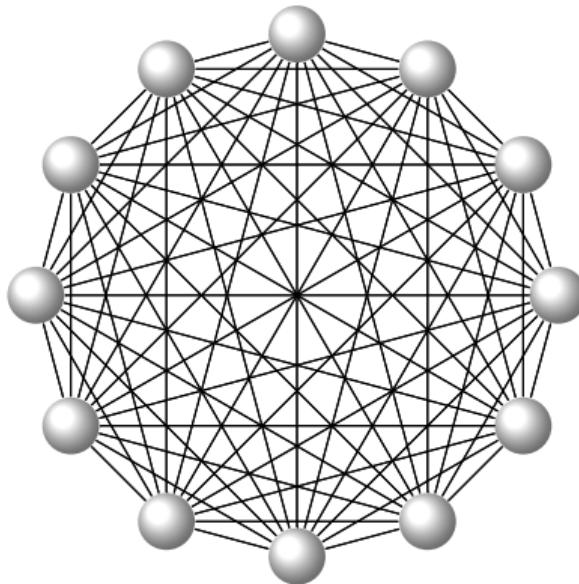
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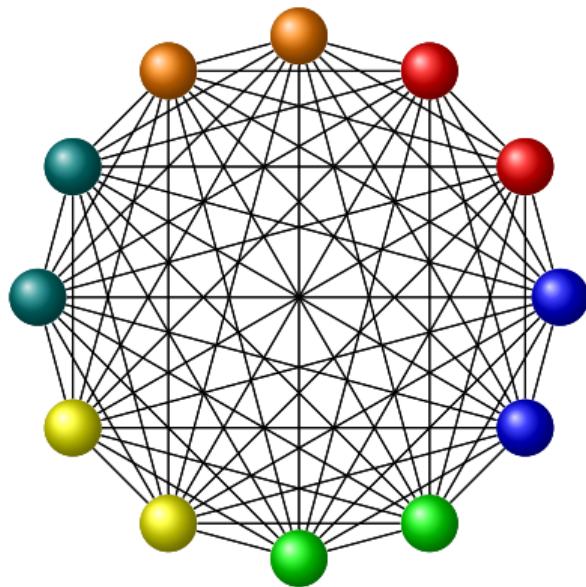
The Strategy

- ① Cover by few subproblems.
- ② Find small (extended) formulations for subproblems.
- ③ Take (convex hull of) union.

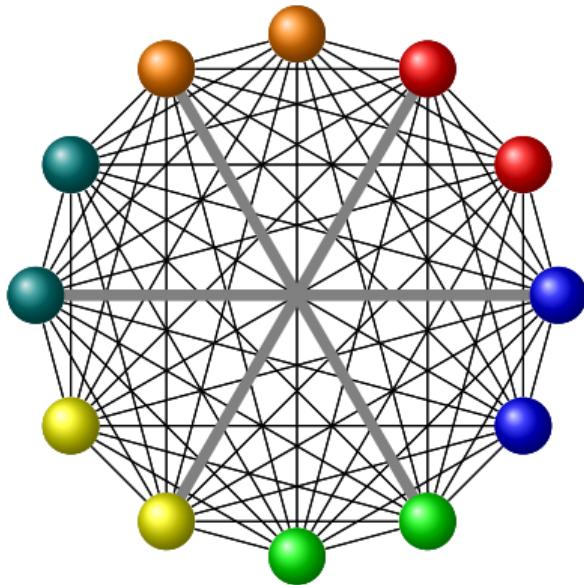
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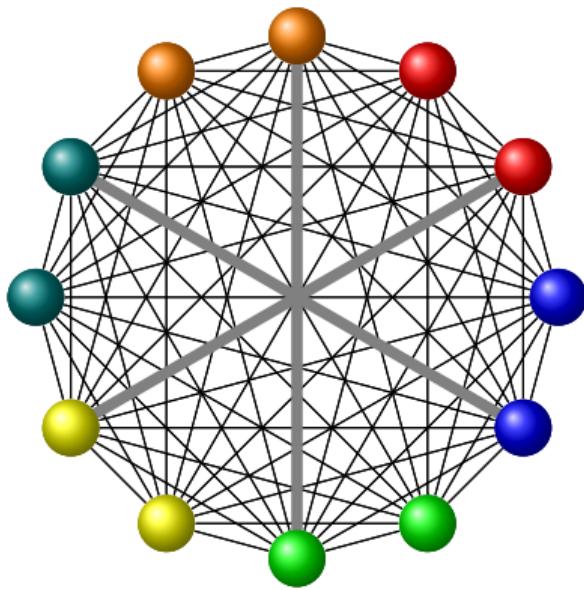
Special Matchings



Colorful Matchings

For $V = W_1 \uplus \dots \uplus W_{2\ell}$, a matching $M \subseteq E$ is **colorful** if it matches exactly one node from each set W_i .

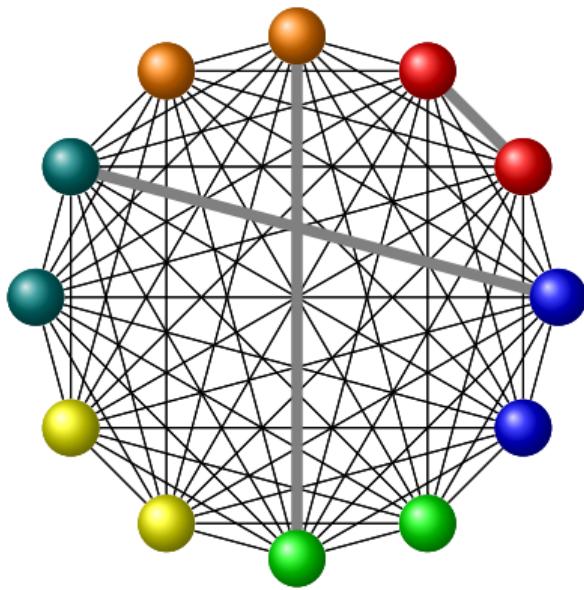
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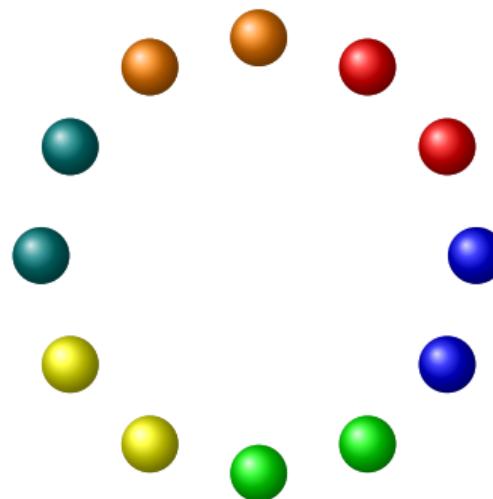
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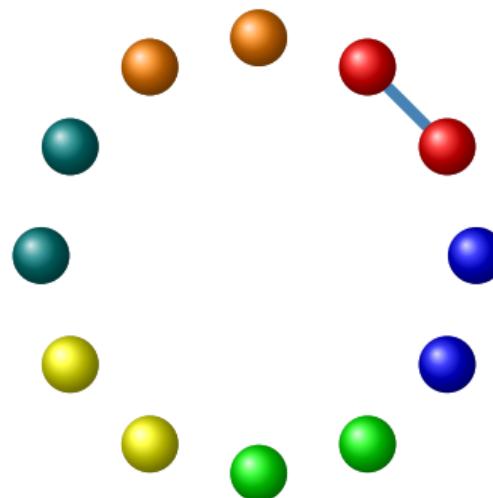
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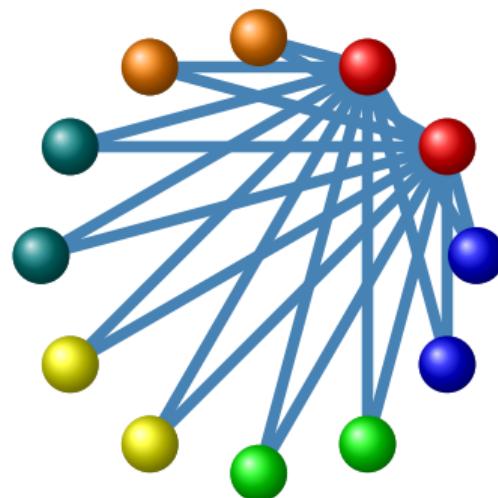
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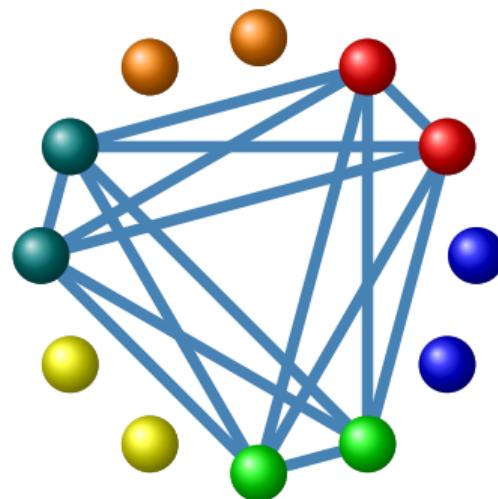
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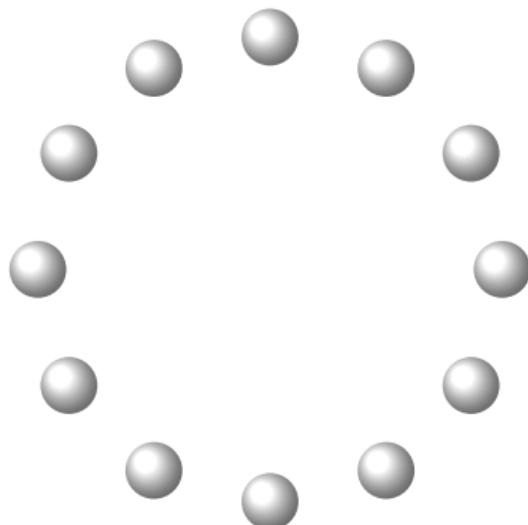
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(n, k) -perfect hash function family of size q

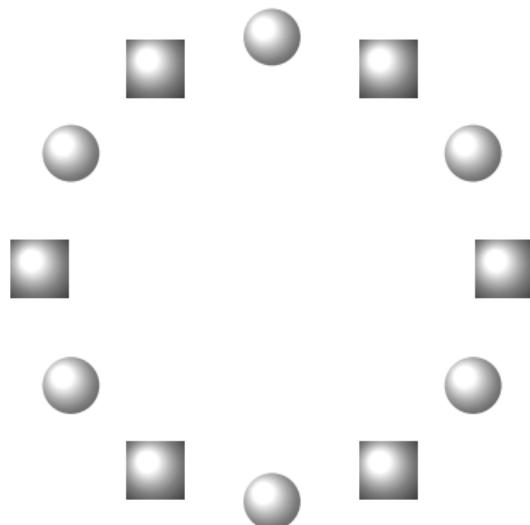
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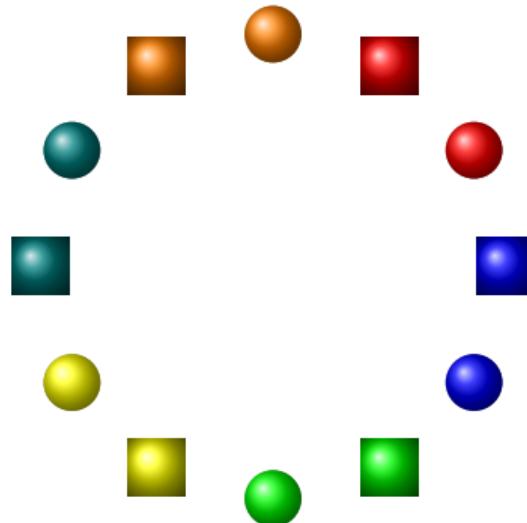
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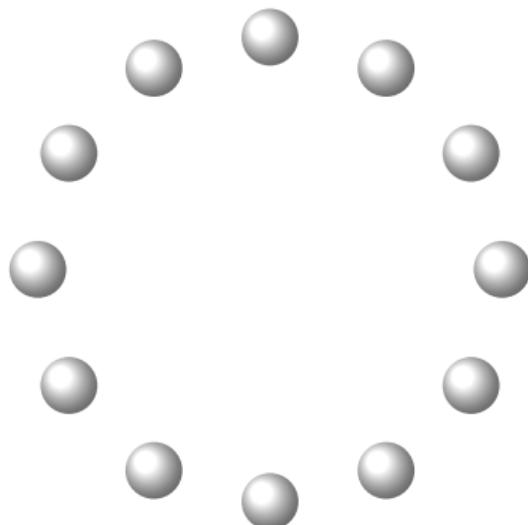
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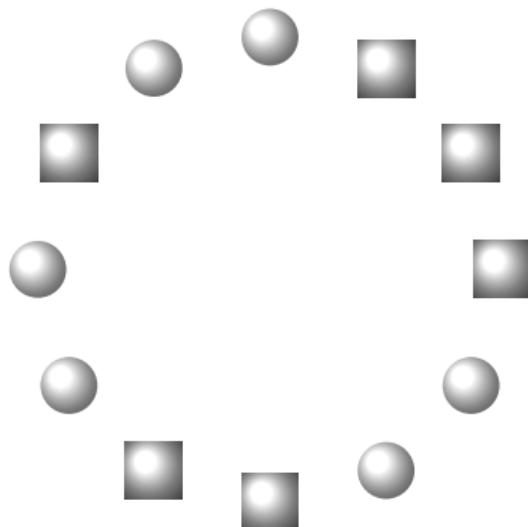
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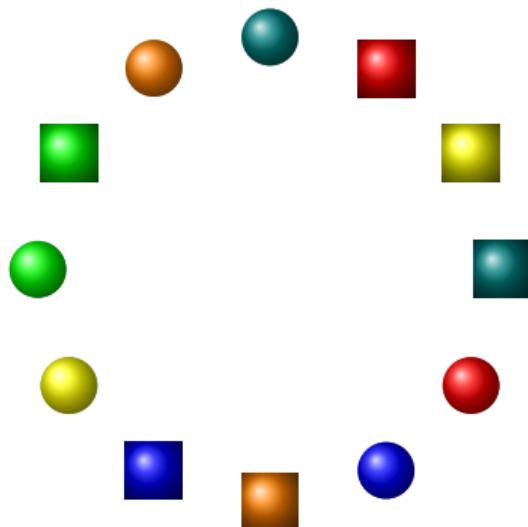
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ALON, YUSTER, ZWICK 1995

For n and k there are (n, k) -perfect hash function families of size

$$2^{O(k)} \log(n).$$

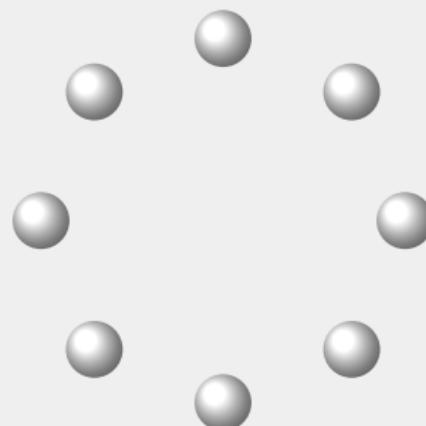
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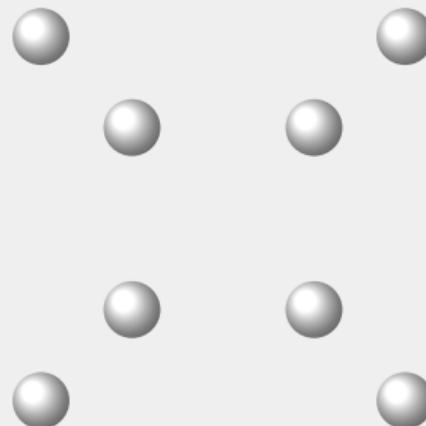
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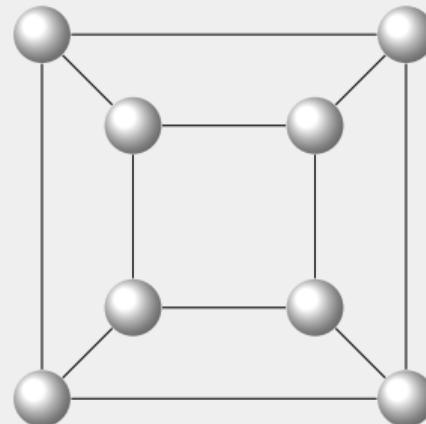
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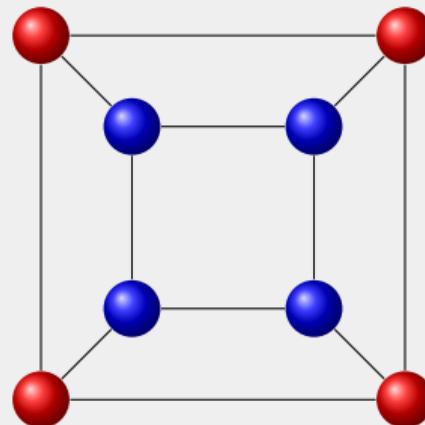
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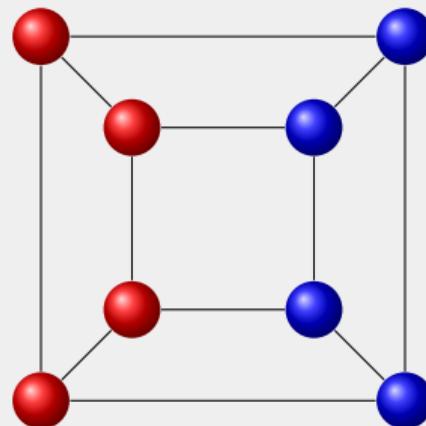
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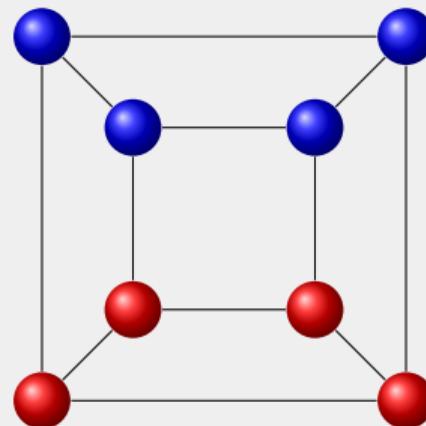
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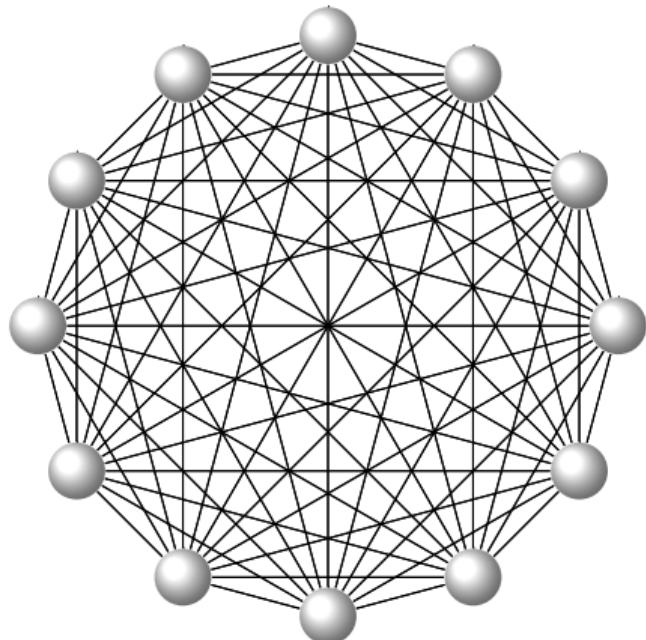
Rothvoss 2014

$$xc(P_{\text{match}}(n)) \geq 2^{\Omega(n)}$$

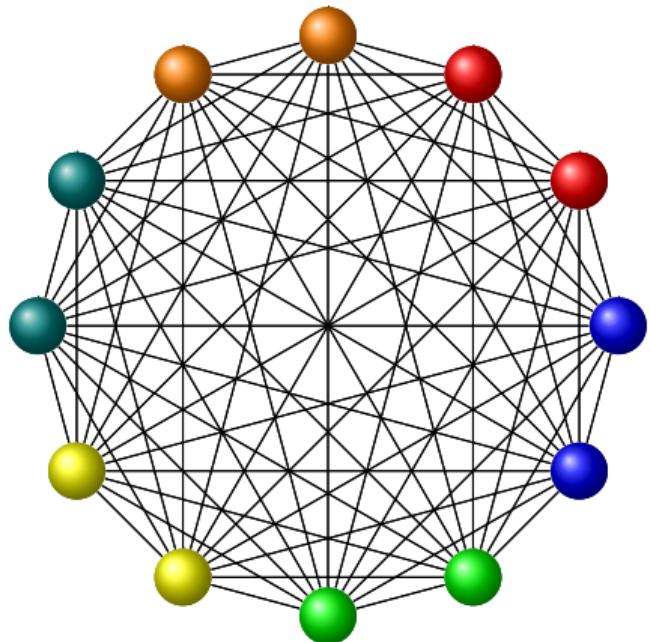
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- ① The Concept
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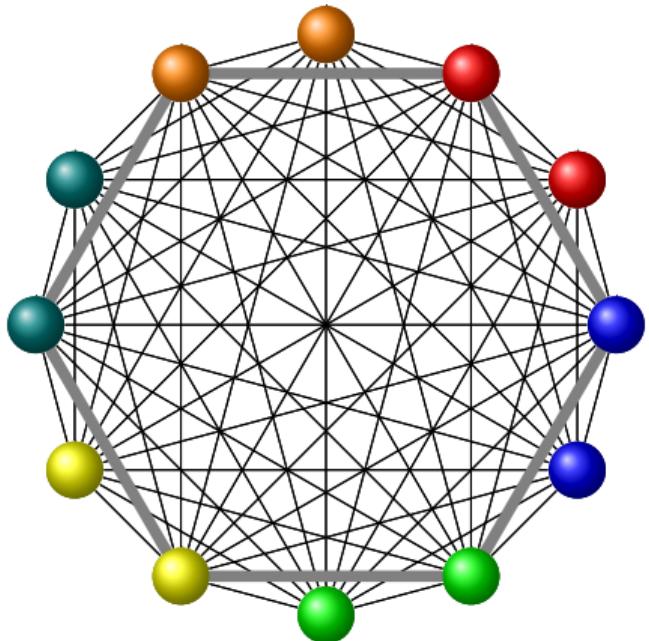
Special Cycles



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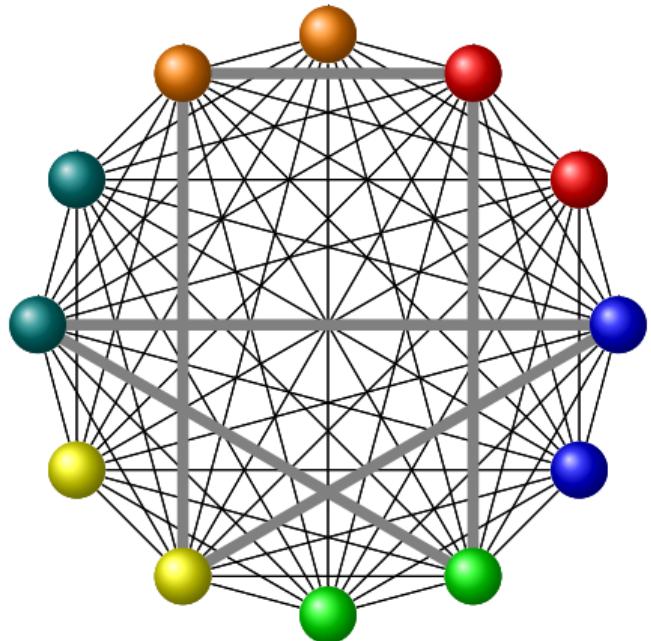
Special Cycles



Colorful cycles

For $V = W_1 \uplus \dots \uplus W_\ell$, a cycle $C \subseteq E$ is **colorful** if it visits each set W_i exactly once.

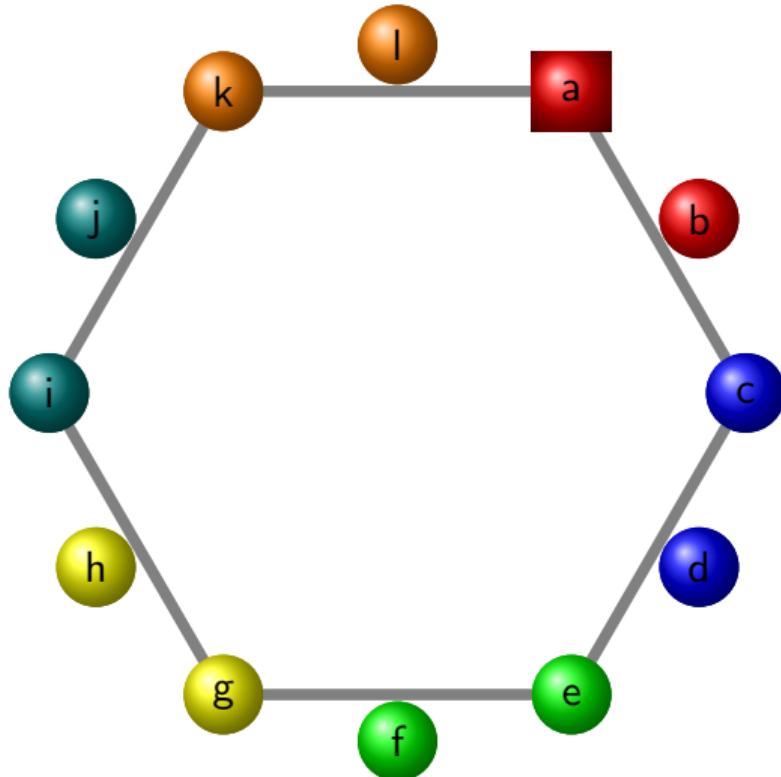
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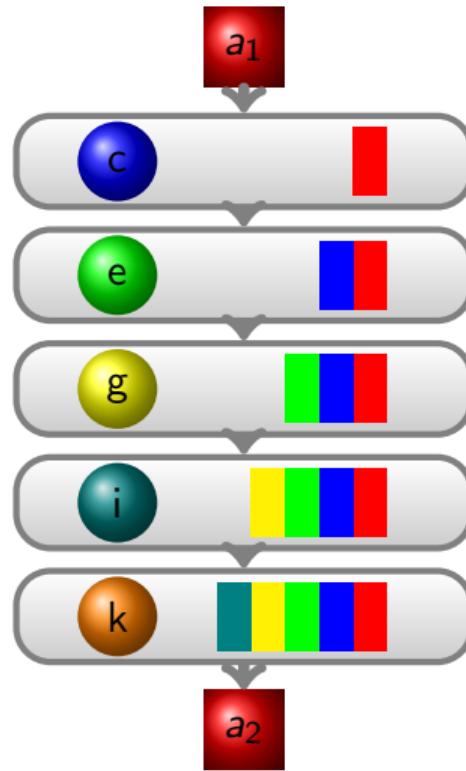
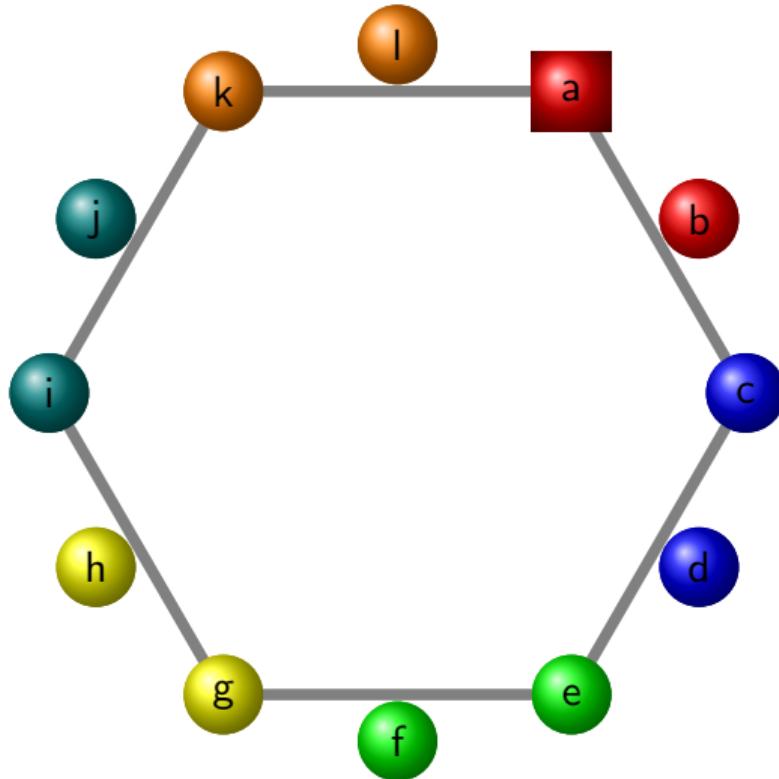
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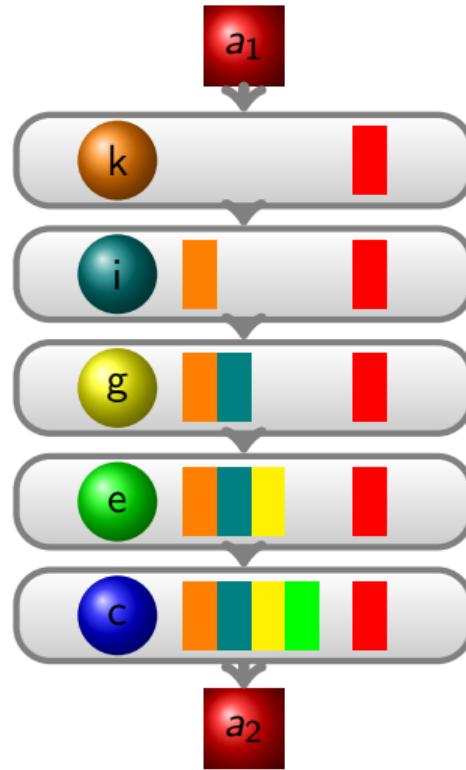
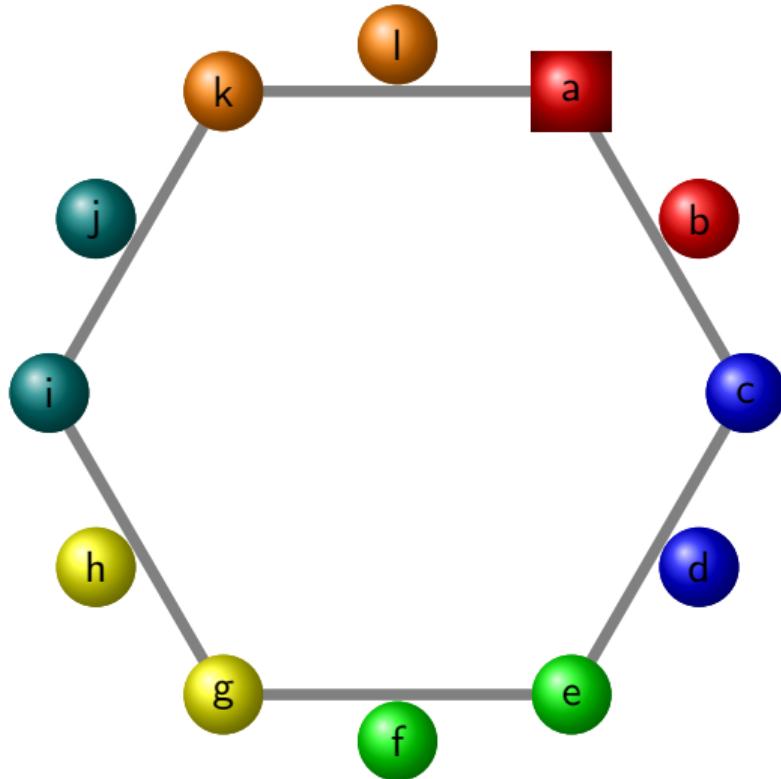
Colorful Cycles with Prescribed Node a



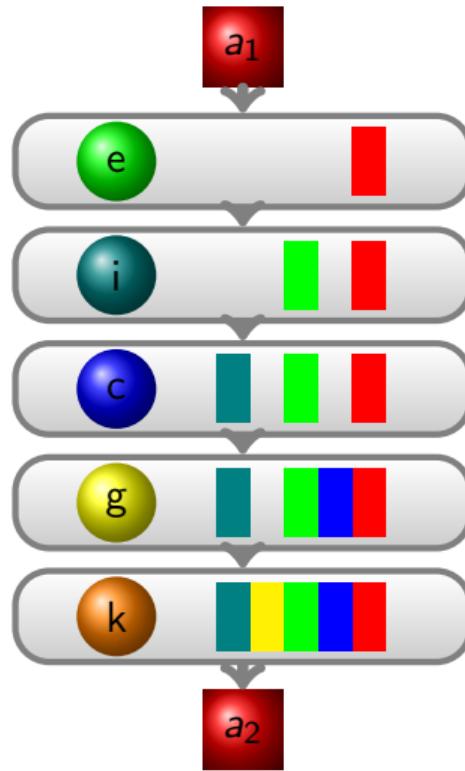
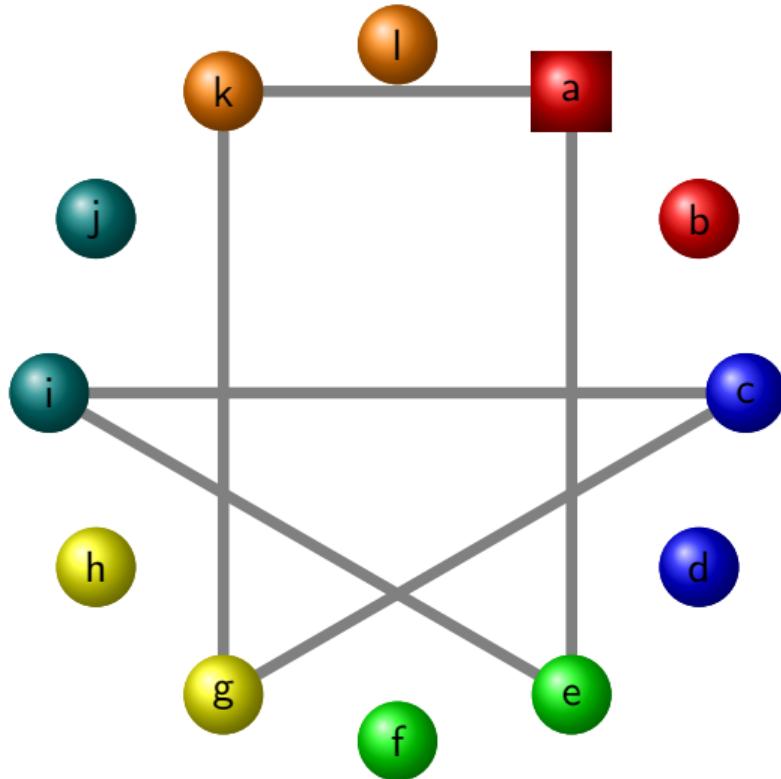
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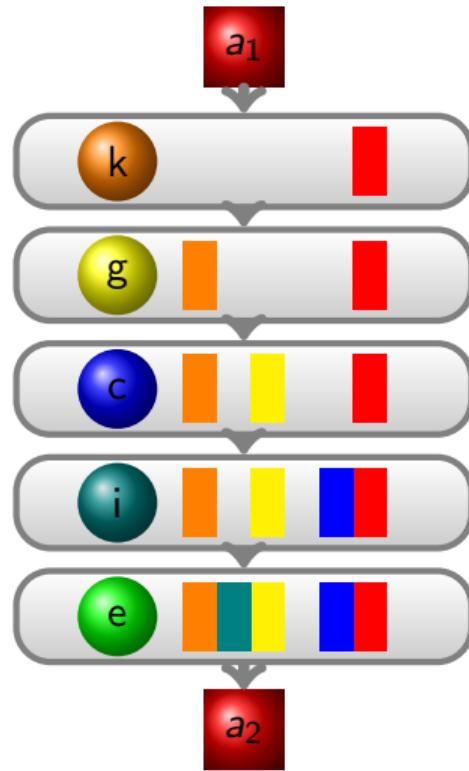
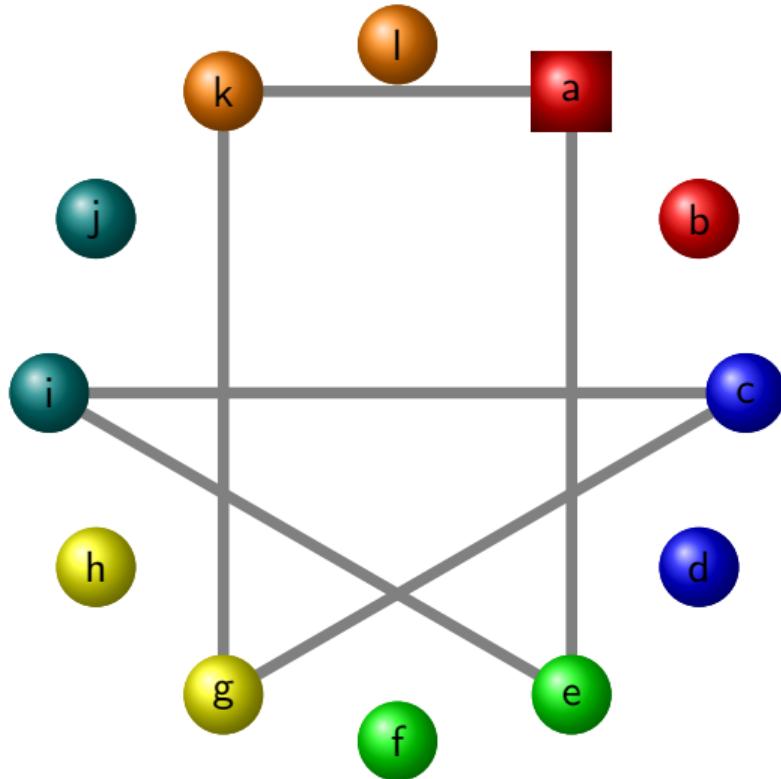
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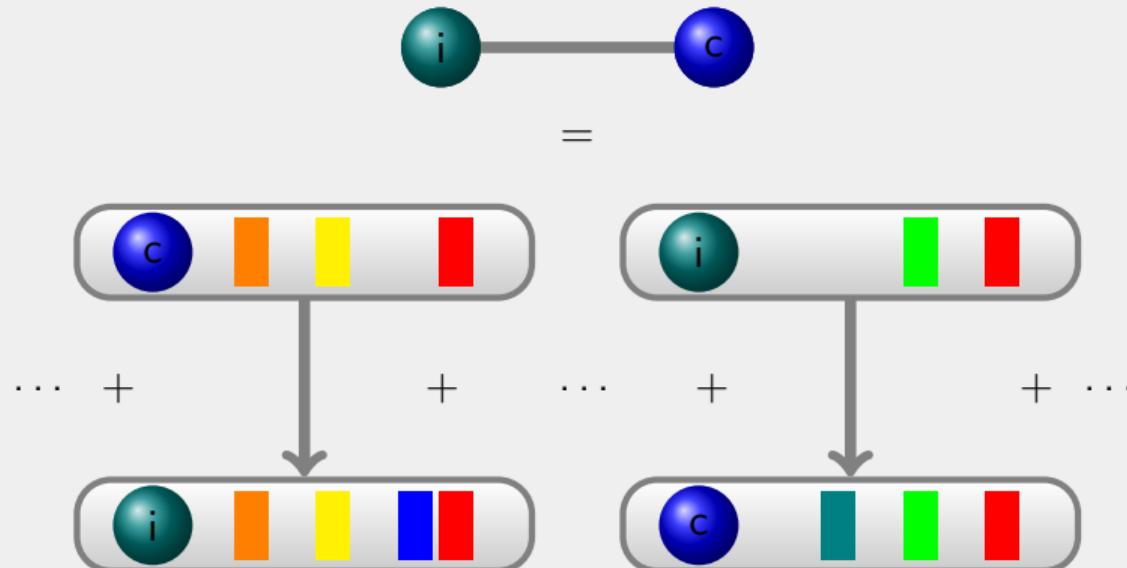
Colorful Cycles with Prescribed Node a



Colorful Cycle Polytopes (Prescribed Node)

Extended Formulation via

- a_1-a_2 flows of value one
- and projection



Combining Things for Cycle Polytopes

We have seen:

- $P_{\text{cycl}}^\ell(n)$ is the convex hull of $2^{O(\ell)} n \log(n)$ colorful cycle polytopes (prescribed)...

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Fiorini, Massar, de Wolff, Tiwary, Pokutta 2012

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Hyperpath Polytopes

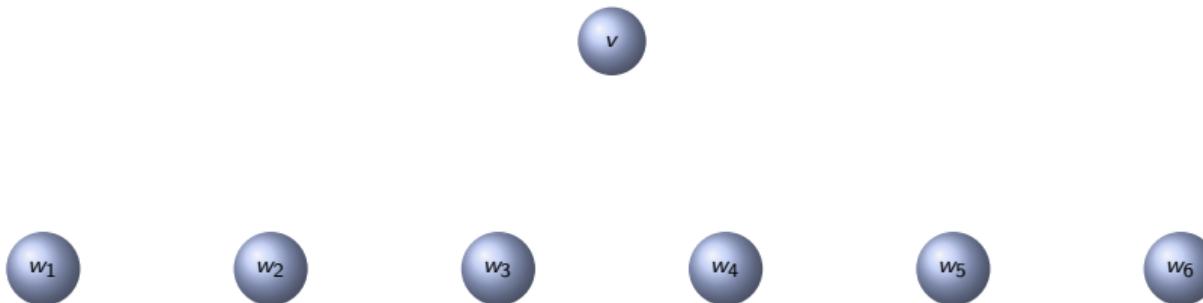
MARTIN, RARDIN, CAMPBELL 1990

The source-to-terminals path polytope of a directed single-tail acyclic hypergraph with a unique source and **terminal-disjoint head sets** is described by the conservation equations and nonnegativity constraints.

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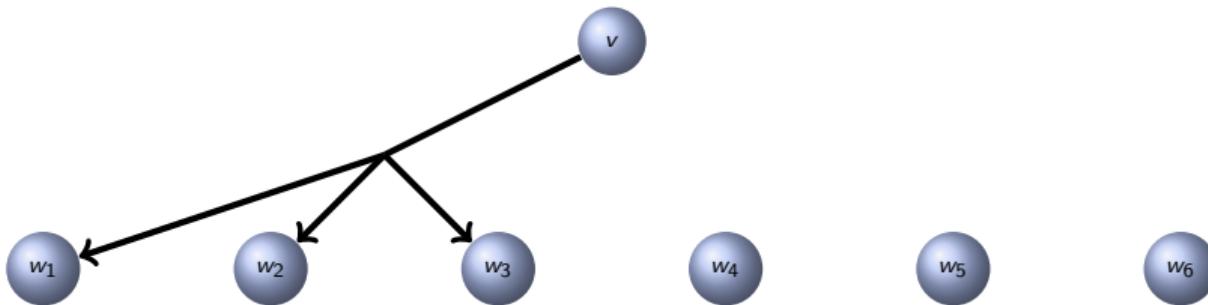
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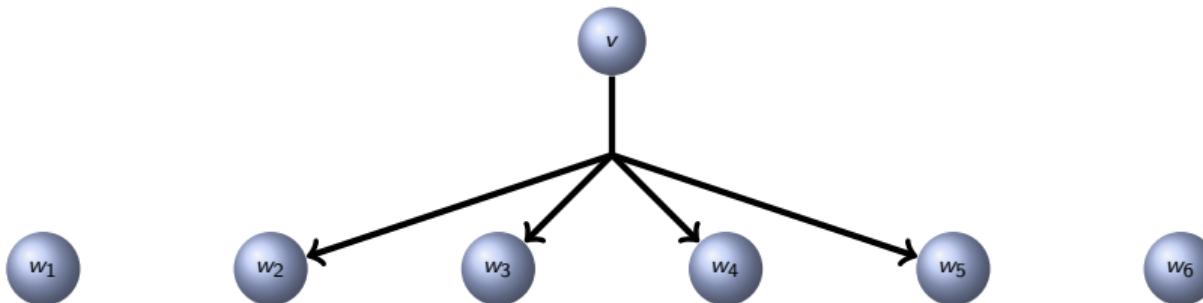
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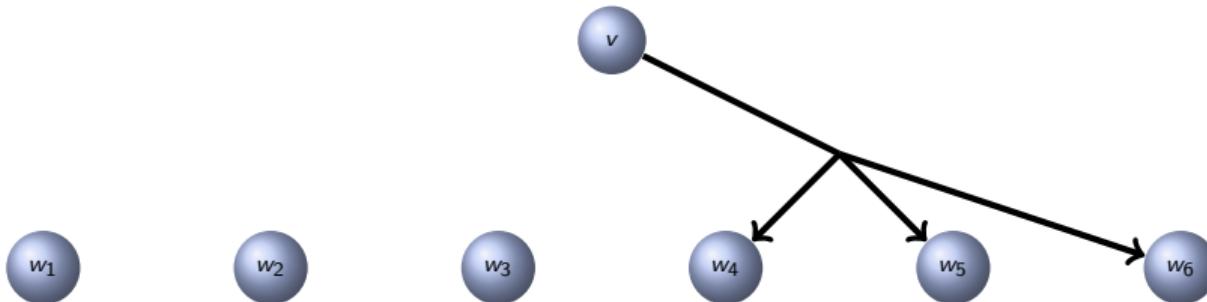
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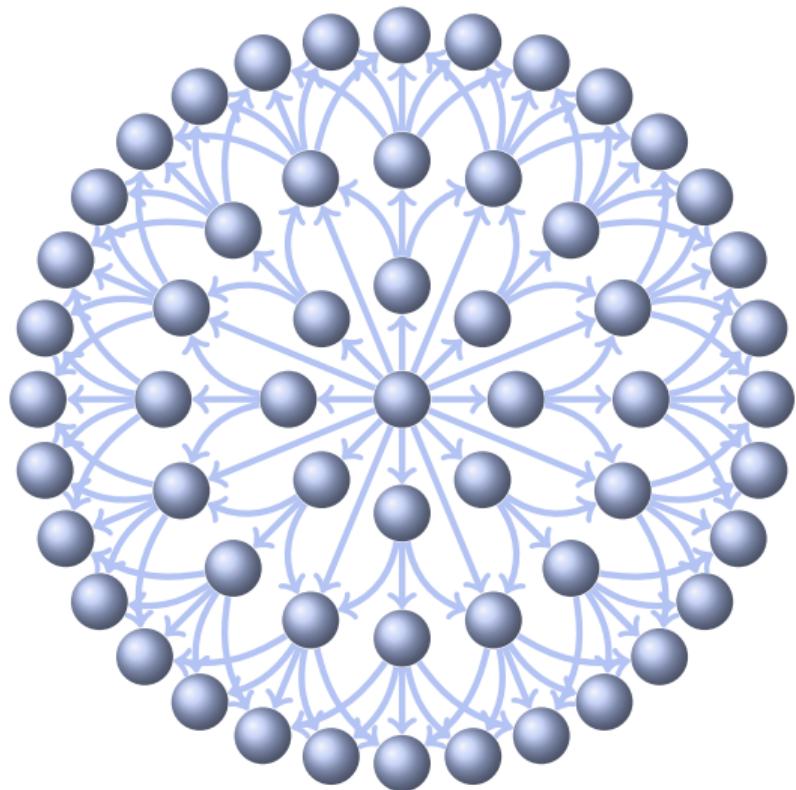
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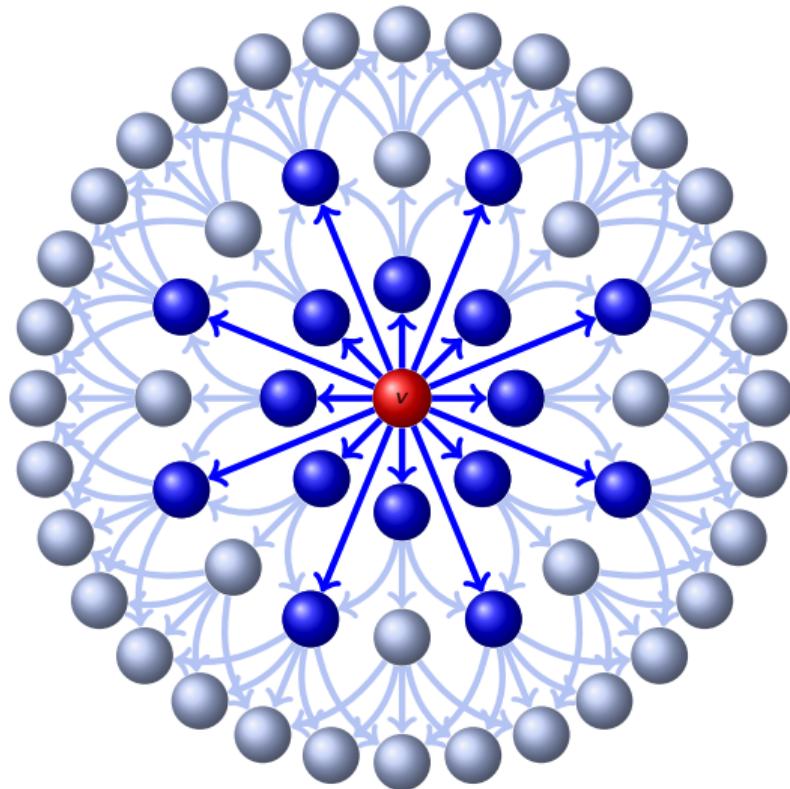
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Branched Combinatorial/Polyhedral Systems



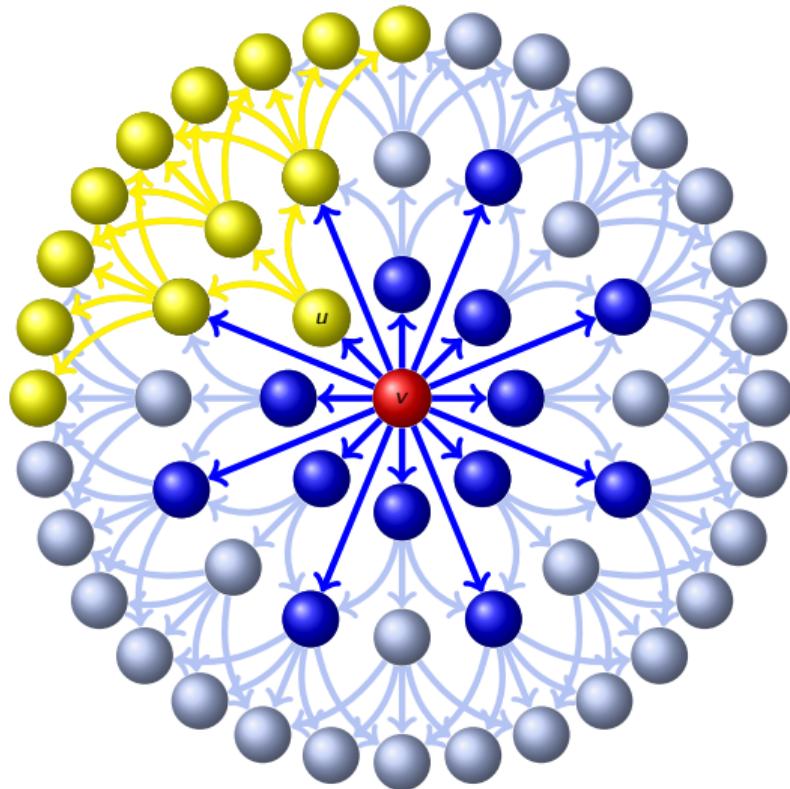
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For each non-sink v

$$\mathcal{S}^{(v)} \subseteq 2^{\text{N}^{\text{out}}(v)}$$

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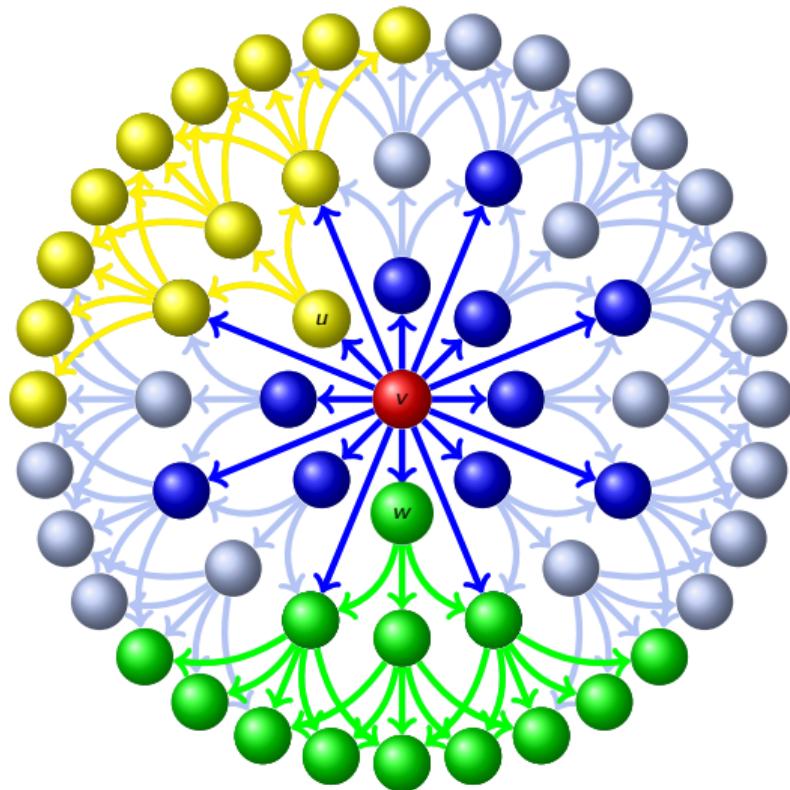
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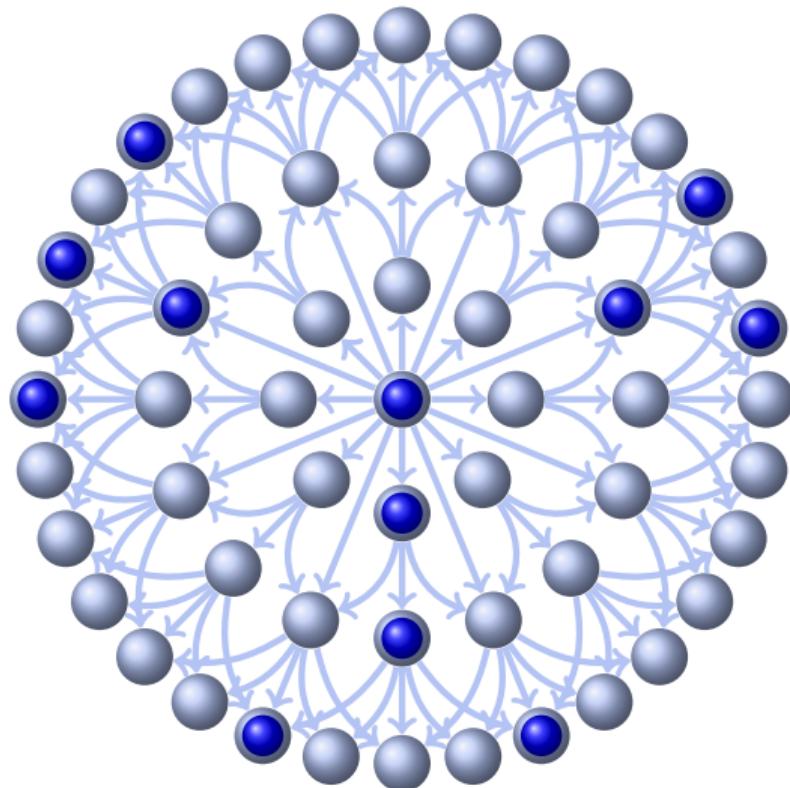
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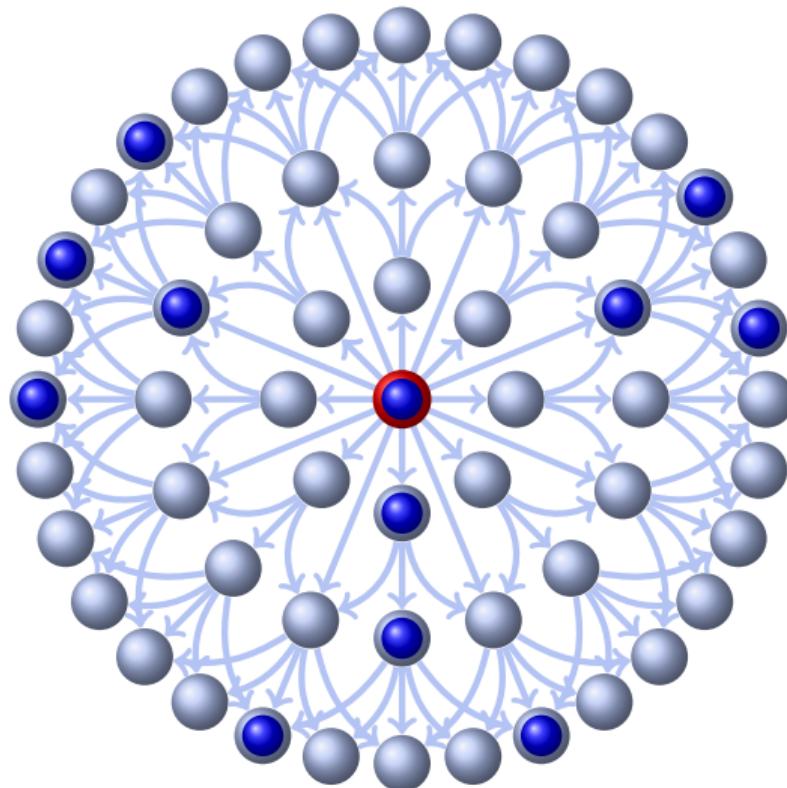
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Feasible set $F \subseteq V$

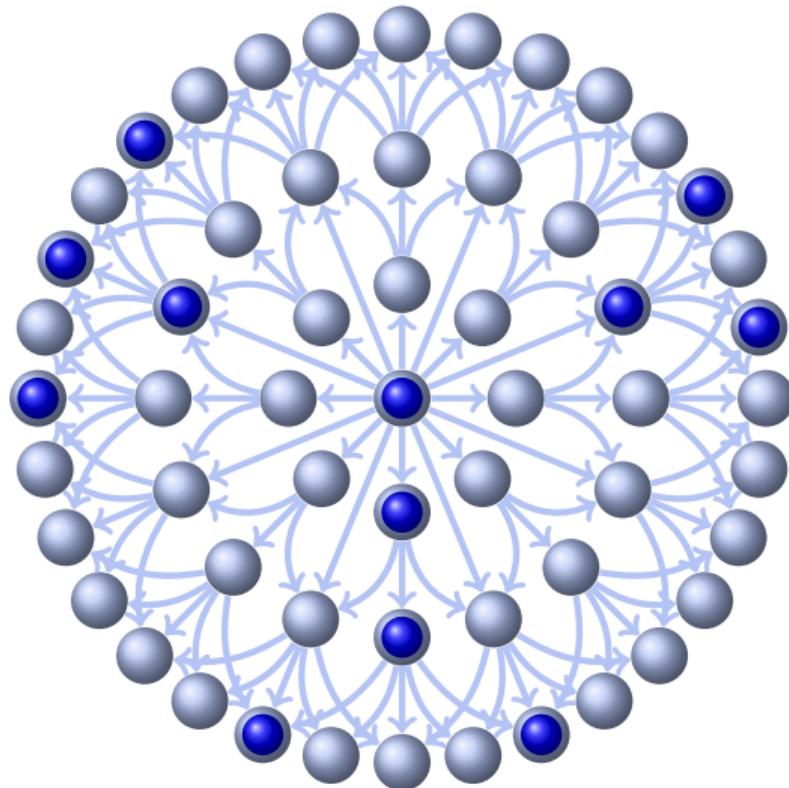
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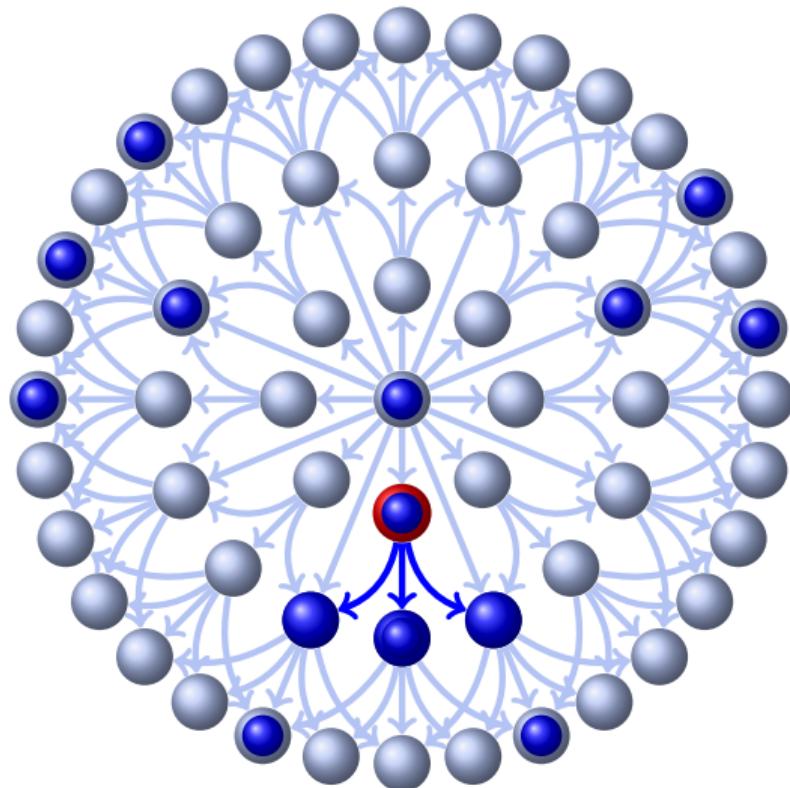
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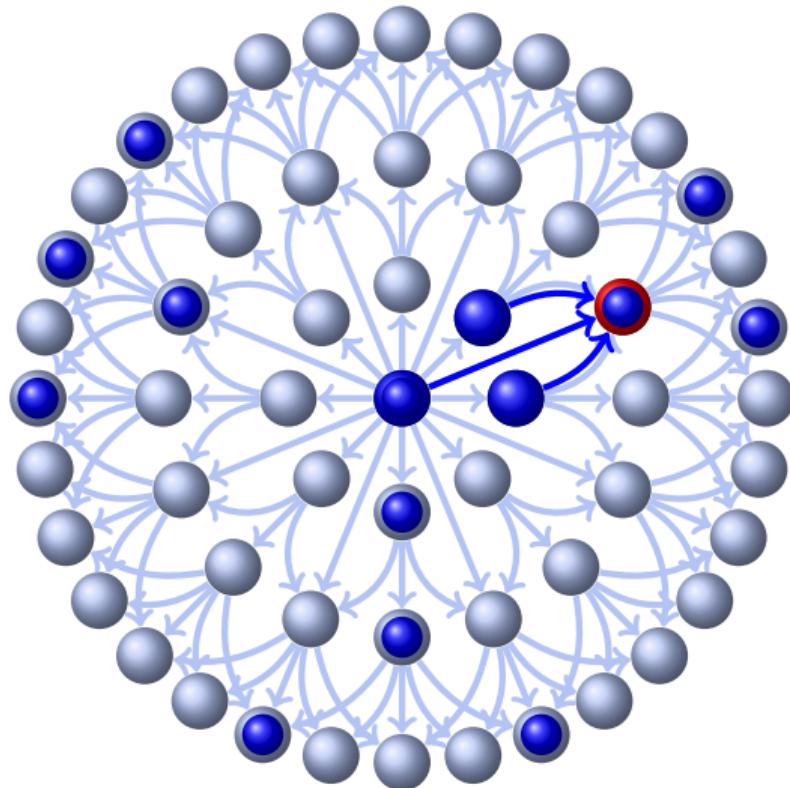
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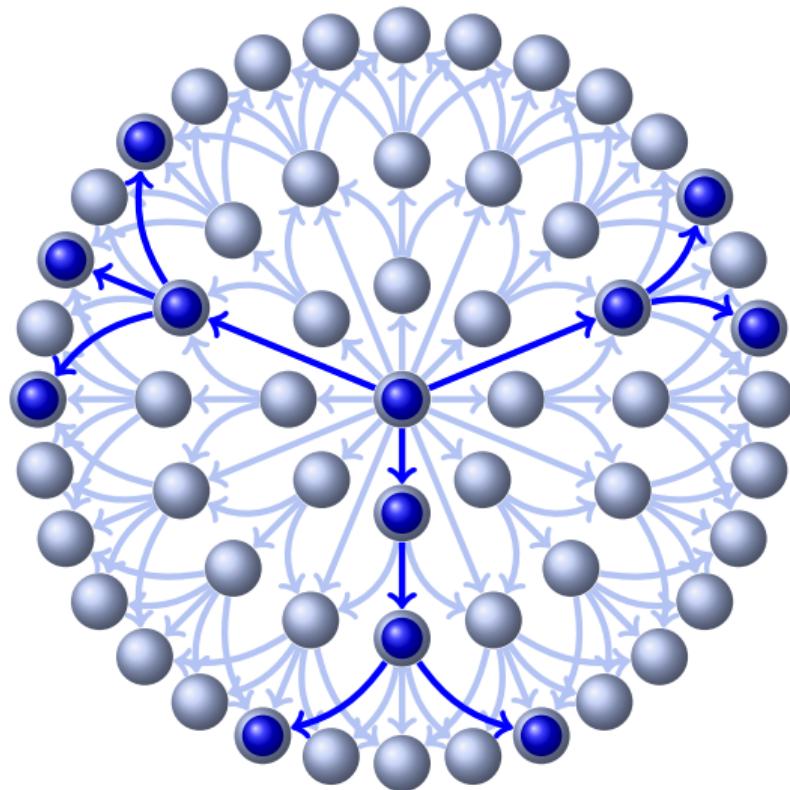
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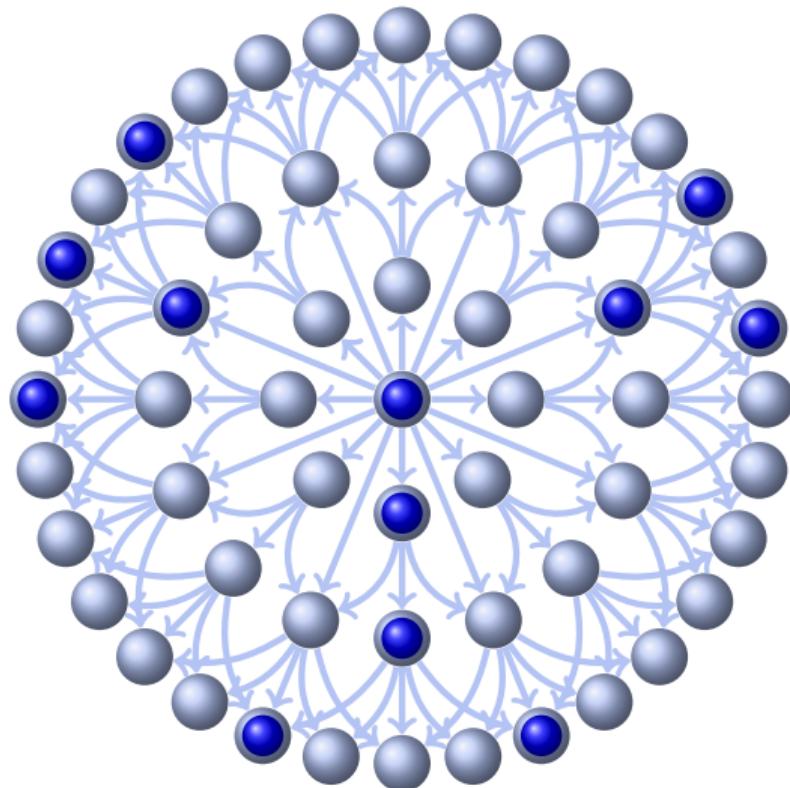
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0/1-Polytope $P(\mathcal{B})$

$\text{conv}(\{\chi(F) : F \text{ feasible}\})$

Extended Formulation

K & Loos 2010

$P(\mathcal{B})$ is described by the following extended formulation:

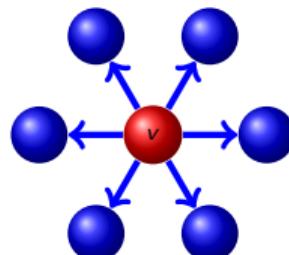
$$x_s = 1$$

$$x_v = y(\delta^{\text{in}}(v)) \quad \text{for all } v \neq s$$

$$A^{(v)} y_{\delta^{\text{out}}(v)} - x_v b^{(v)} \leq 0 \quad \text{for all non-sinks } v$$

$$x_v \geq 0 \quad \text{for all non-sinks } v$$

(if $A^{(v)} z \leq b^{(v)}$ describe $P^{(v)} = \text{conv}(\{\chi(S) : S \in \mathcal{S}^{(v)}\})$)



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Extended Formulations via Duality

Observation due to Martin (1991)

$$\text{xc}\{x : Ax \leq \beta \cdot \mathbf{1}\} \leq \text{xc}(\text{conv}\{\text{rows of } A\}) + 1$$

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Application to Spanning Tree Polytopes

$$\mathbb{P}_{\text{spt}}(G) = \{x \in \mathbb{R}_+^E : x(E) = |V| - 1, x(E(U)) \leq |U| - 1 \text{ for all } \emptyset \neq U \subsetneq V\}$$

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Extended Formulations for Non-Empty-Subgraphs Polytopes

All-Subgraphs Polytope of G

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Martin 1991

$$\text{xc}(P_{\text{spt}}(G)) \leq |V| \cdot (2|V| + 2|E|) + |E| + 1 = 2|V||E| + 2|V|^2 + |E| + 1$$

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Conforti, K, Walter, Weltge (2015)

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Corollary

$$|xc(P_{\text{spt}}(G)) - xc(P_{\text{ne-sub}}(G))| \leq 2|V| + |E|$$

Graphs of Bounded Genus

Djidjev & Venkatesan (1995), Hutchinson & Miller (1987)

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Fiorini, Huynh, Joret, Pashkovich (2016)

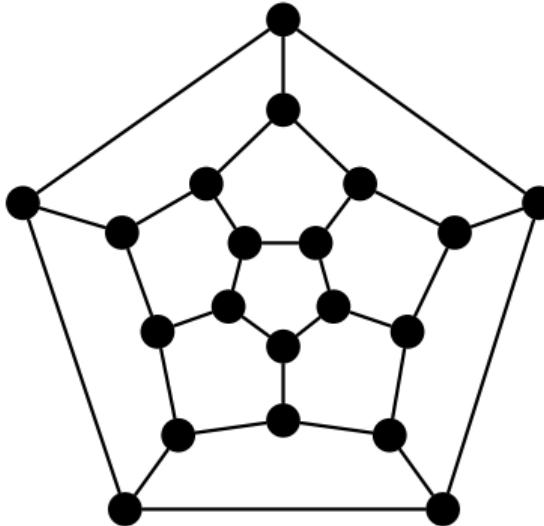
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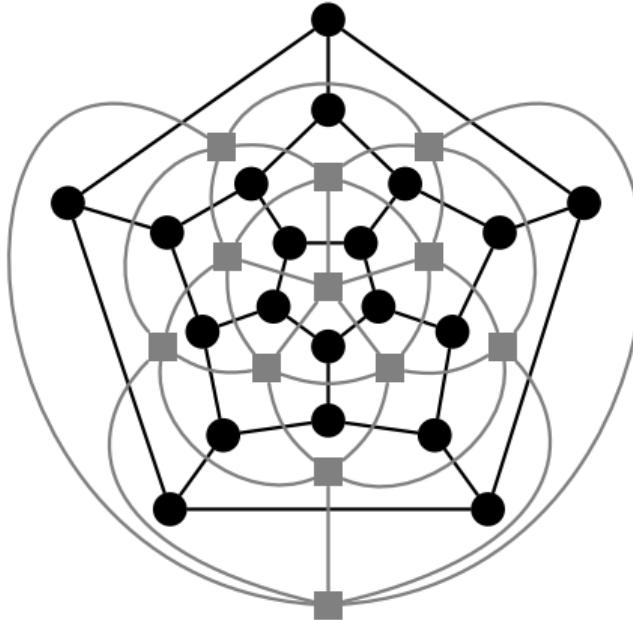
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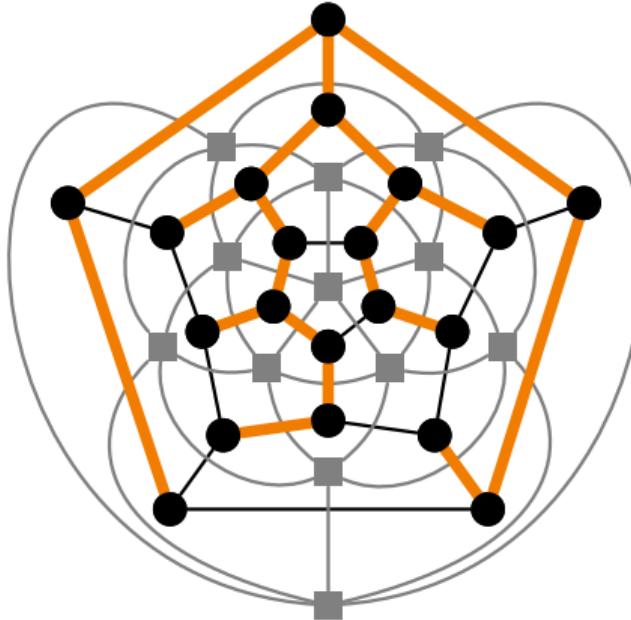
Spanning Trees in Planar Graphs



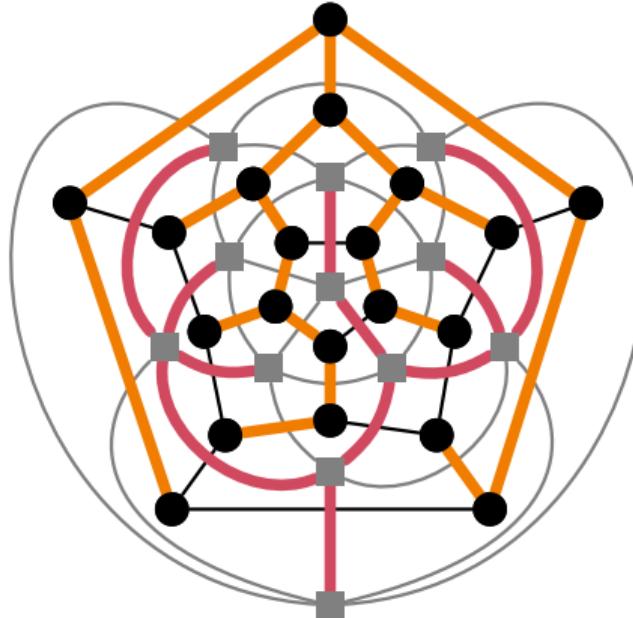
Spanning Trees in Planar Graphs



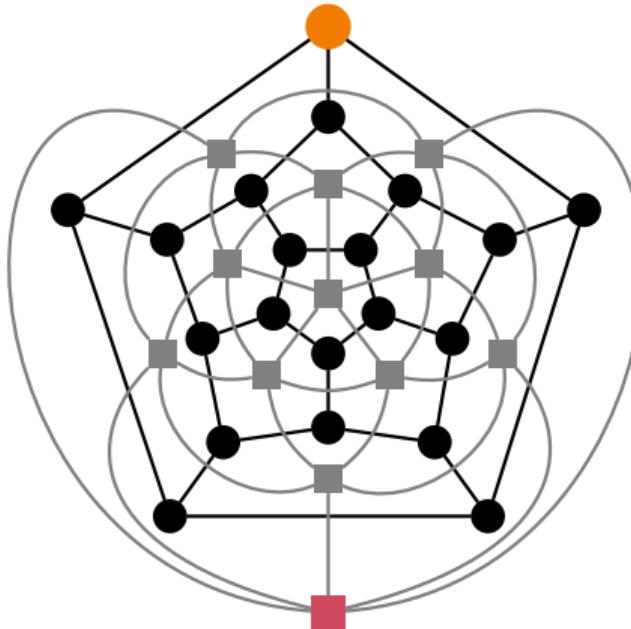
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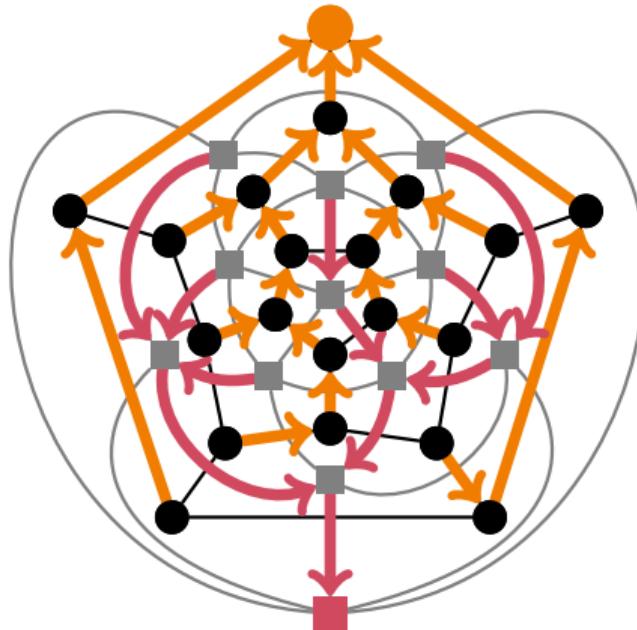
Spanning Trees in Planar Graphs



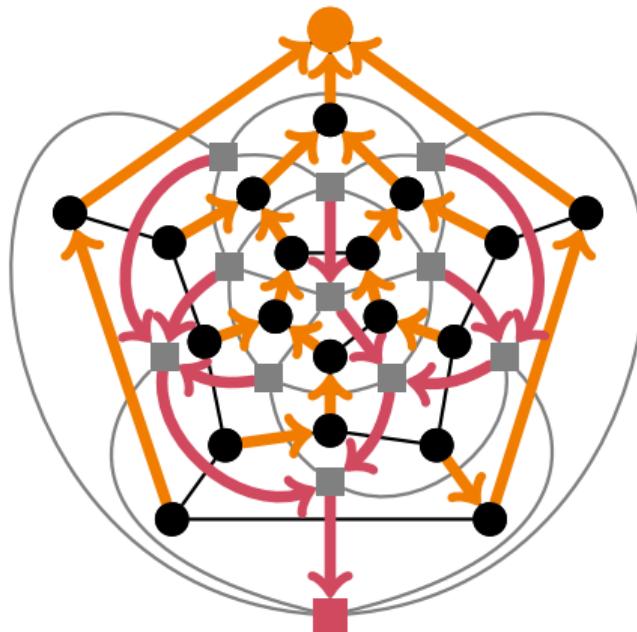
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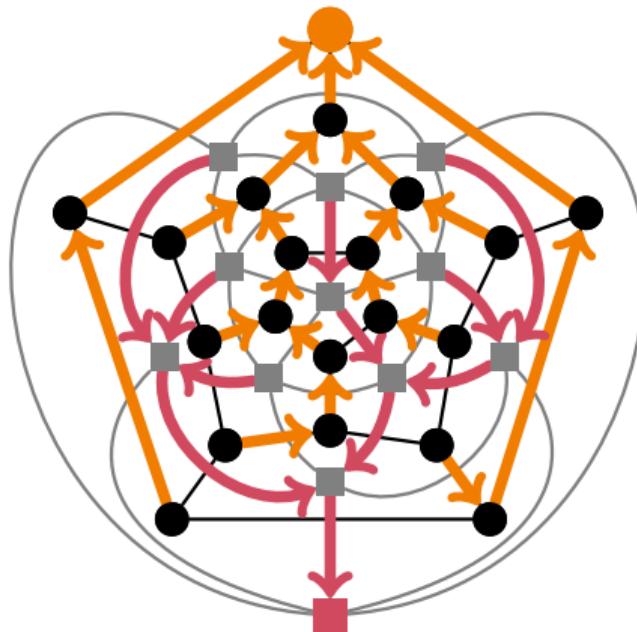


Spanning Trees in Planar Graphs



$$\mathbb{P}_{\text{arb}}(G) = \text{conv}\{(\chi(\vec{T}), \chi(\vec{T}^*)) : T \text{ spanning tree of } G\}$$

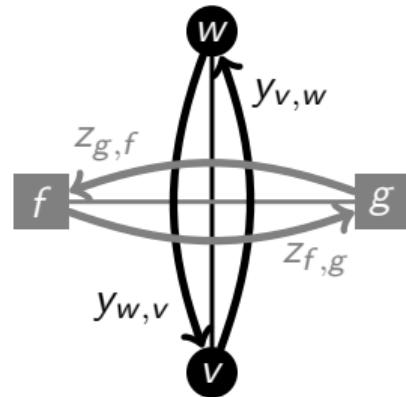
Spanning Trees in Planar Graphs



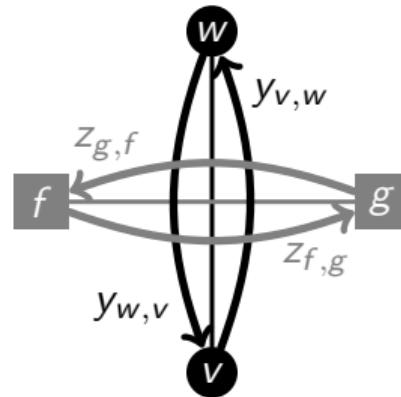
$$\mathcal{P}_{\text{arb}}(G) = \text{conv}\{\chi(\vec{T}), \chi(\vec{T}^*) : T \text{ spanning tree of } G\}$$

$$\mathcal{P}_{\text{spt}}(G) = p(\mathcal{P}_{\text{arb}}(G)) \text{ with linear projection } p$$

Linear Description of $P_{\text{arb}}(\cdot)$



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WILLIAMS 2001

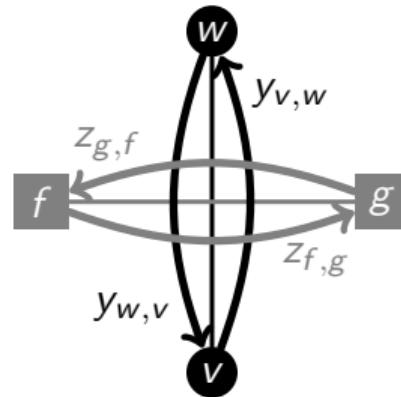
$$y_{v,w} + y_{w,v} + z_{f,g} + z_{g,f} = 1 \quad \forall \{v, w\} \in E$$

$$\sum_w y_{v,w} = 1 \quad \forall v \neq \text{root}$$

$$\sum_g z_{f,g} = 1 \quad \forall w \neq \text{root}$$

$$y_{v,w}, y_{w,v}, z_{f,g}, z_{g,f} \geq 0 \quad \forall \{v, w\} \in E$$

Linear Description of $P_{\text{arb}}(\cdot)$



WILLIAMS 2001

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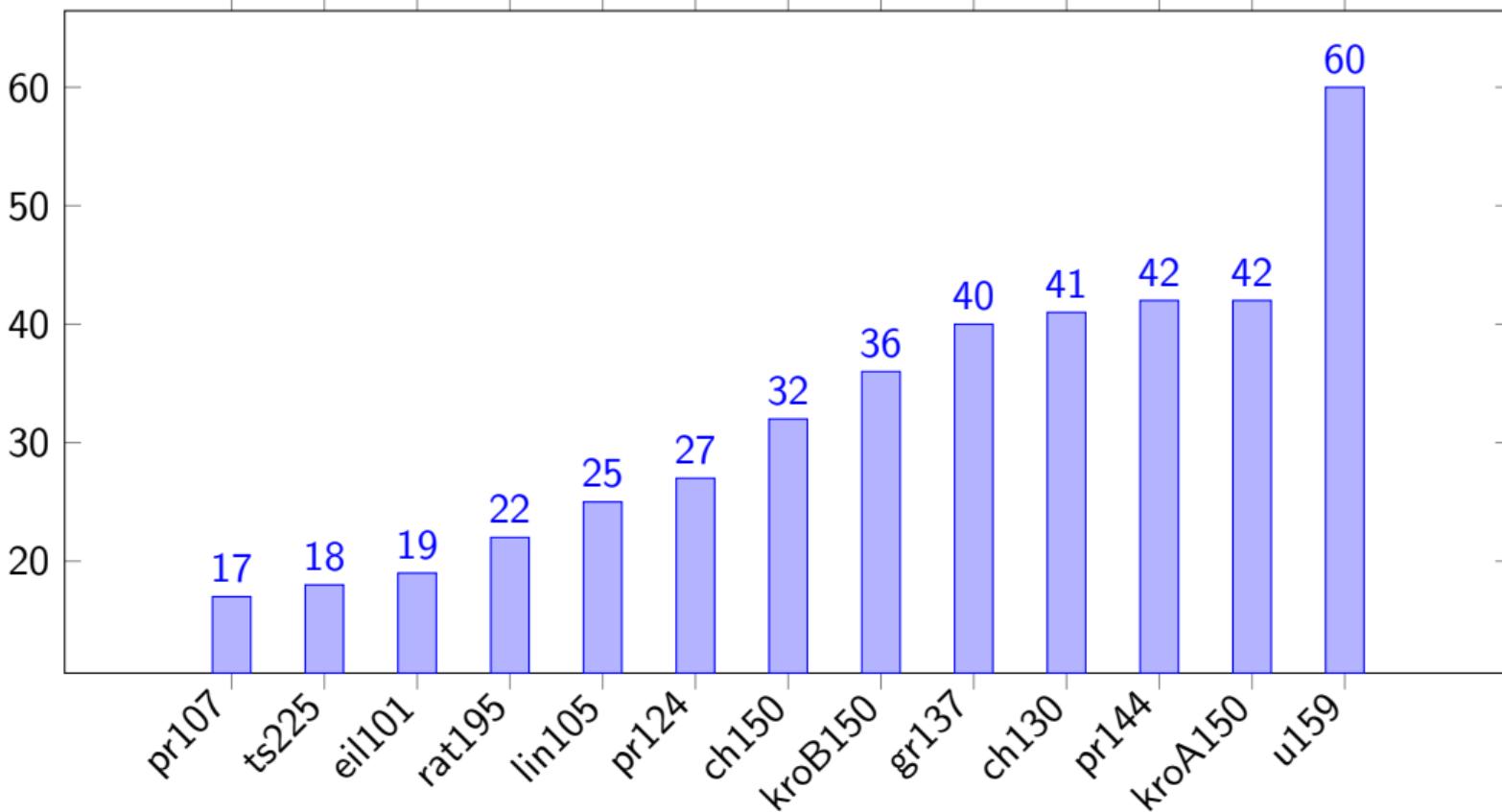
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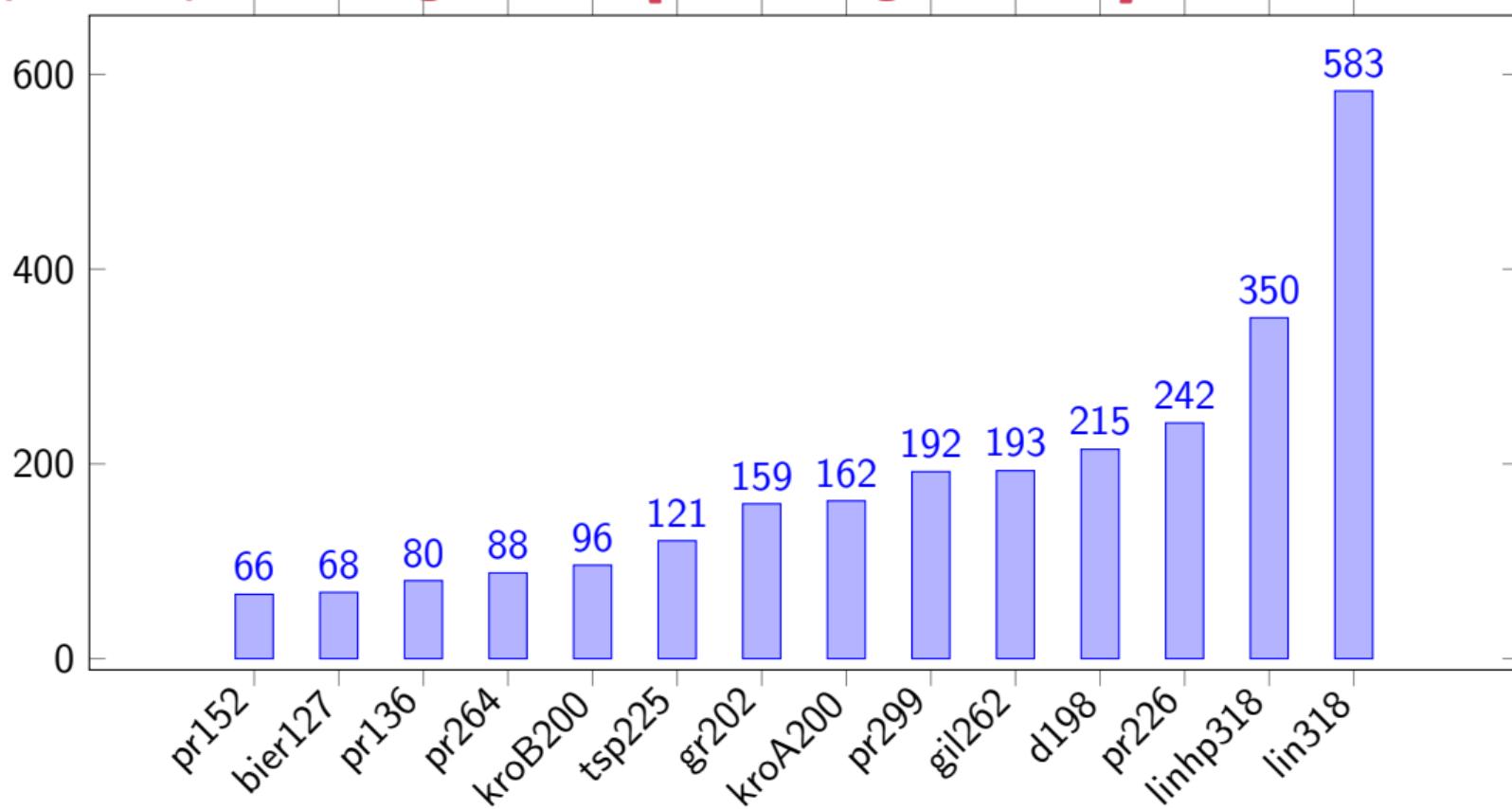
$$y_{v,w}, y_{w,v}, z_{f,g}, z_{g,f} \geq 0 \quad \forall \{v, w\} \in E$$

Thus $\text{xc}(P_{\text{spt}}(G)) \leq O(n)$ for planar G on n nodes.

Speed-Ups for Degree ≤ 3 [K & Sorgatz 2012]



Speed-Ups for Degree ≤ 3 [K & Sorgatz 2012]



Polynomial Spanning Tree Optimization

Setup for $G = (V, E)$, $\mathcal{M} \subseteq 2^E$ (acyclic subsets)

- $\Omega(\mathcal{M}) := \{(x, (y_M)_{M \in \mathcal{M}}) : x \text{ incidence vector of a spanning tree in } G, y_M = \prod_{e \in M} x_e \text{ for all } M \in \mathcal{M}\}$
- Linear optimization over $\Omega(\mathcal{M}) \iff$ Optimization of polynomials with support in \mathcal{M}
- $P(\mathcal{M}) := \text{conv}(\Omega(\mathcal{M}))$

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Relaxations

- $\mathcal{M}' \subseteq \mathcal{M}$: $\Omega(\mathcal{M}', \mathcal{M}) := \{(x, (y_M)_{M \in \mathcal{M}}) : x \in \dots, y_M = \prod_{e \in M} x_e \text{ for all } M \in \mathcal{M}'\}$
- $P(\mathcal{M}', \mathcal{M}) := \text{conv}(\Omega(\mathcal{M}', \mathcal{M})) (\cong P(\mathcal{M}') \times \mathbb{R}^{\mathcal{M} \setminus \mathcal{M}'})$
- For $\mathcal{M} = \mathcal{M}_1 \cup \dots \cup \mathcal{M}_t$: $P(\mathcal{M}) \subseteq P(\mathcal{M}_1, \mathcal{M}) \cap \dots \cap P(\mathcal{M}_t, \mathcal{M})$

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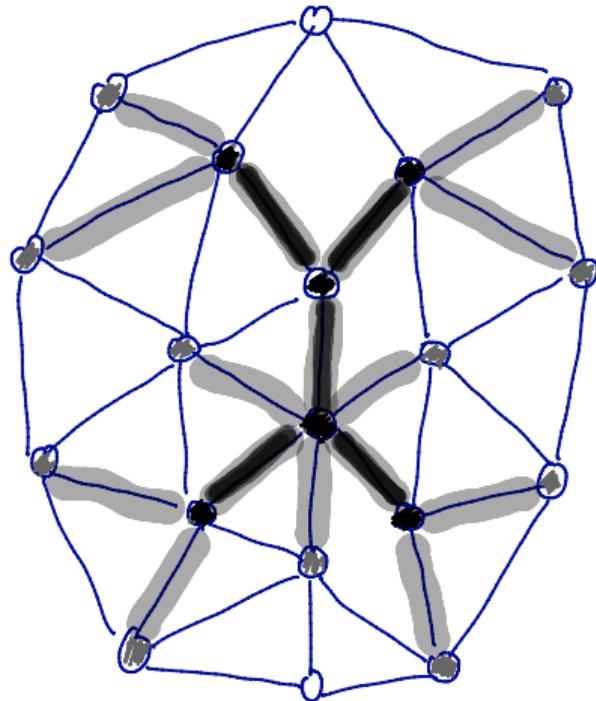
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Disjunctive extended formulation

$$\text{xc}(P(\mathcal{M})) \leq |V|^{\text{width of } \mathcal{M}} \cdot \text{xc}(\mathsf{P}_{\text{spt}}(G))$$

A Single Chain \mathcal{M} of Trees



$$[M_1 \leq M_2 = \bar{M}]$$

Fischer, Fischer, McCormick 2016

$P(\mathcal{M})$ is described by

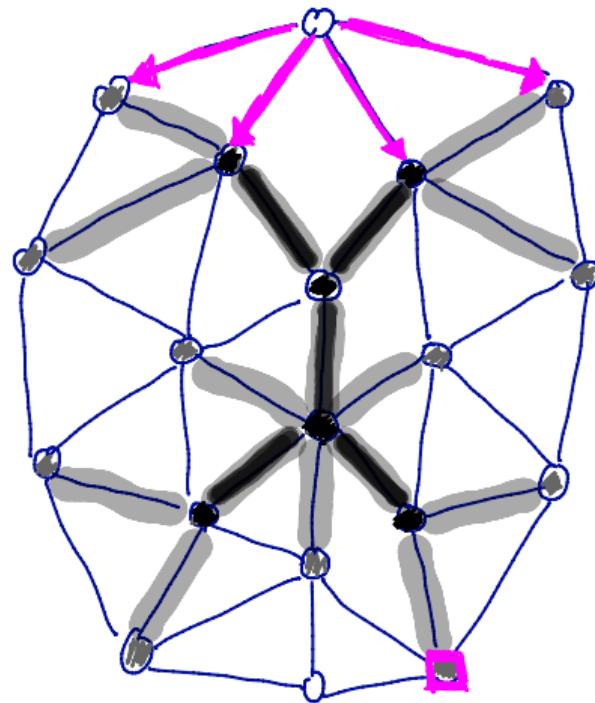
- $x \in P_{\text{spt}}(G)$, $y \geq \mathbf{0}$
- $y_i \leq x_e$ for all $i, e \in M_i \setminus M_{i-1}$
- $y_i - y_{i-1} \geq \sum_{e \in M_i \setminus M_{i-1}} -|M_i \setminus M_{i-1}|$ for all i
- $y_i \leq y_{i-1}$ for all i
- $x(\cup_j E[S_j]) + \sum_i \beta_i y_i \leq \sum_j (|S_j| - 1)$ for all pairwise disjoint S_j

(Similarly even for general matroids.)

For a single pair of edges see also:

- Buchheim & Klein 2014
- Fischer & Fischer 2013

The Extended Formulation $Q(\mathcal{M})$

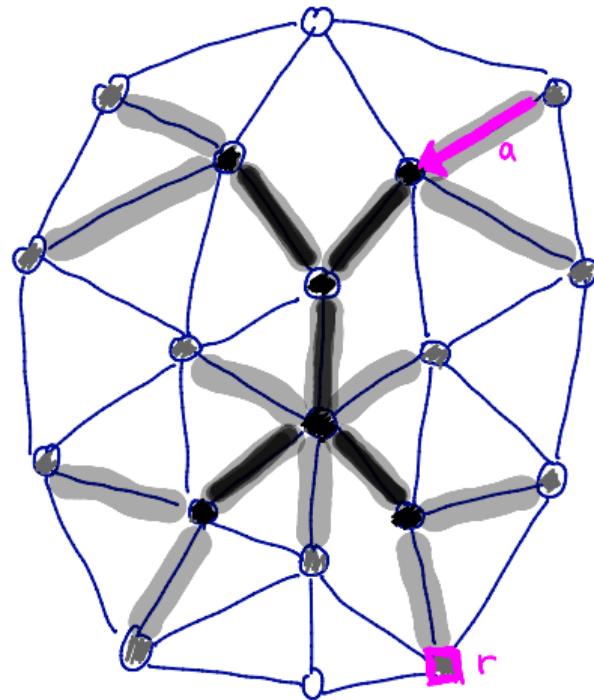


$$[M_1 \leq M_2 =: \bar{M}]$$

Friesen & K 2017

$$\text{xc}(P(\mathcal{M})) \leq \text{xc}(\mathsf{P}_{\text{spt}}(G)) + O(|\bar{M}| \cdot |E|)$$

The Extended Formulation $Q(\mathcal{M})$

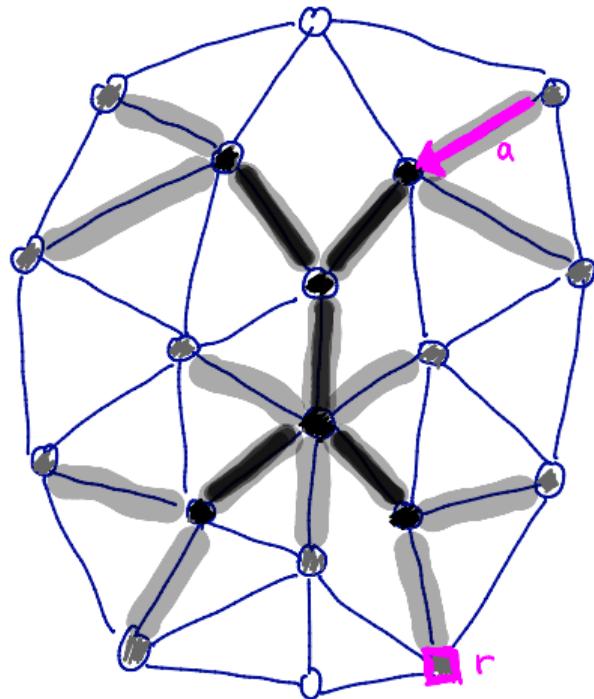


Friesen & K 2017

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$$[M_1 \leq M_2 = \bar{M}] \quad \chi_2 \leq z_a^r$$

The Extended Formulation $Q(\mathcal{M})$



$$[M_1 \leq M_2 = \bar{M}] \quad \chi_2 \leq z_\alpha^r$$

Friesen & K 2017

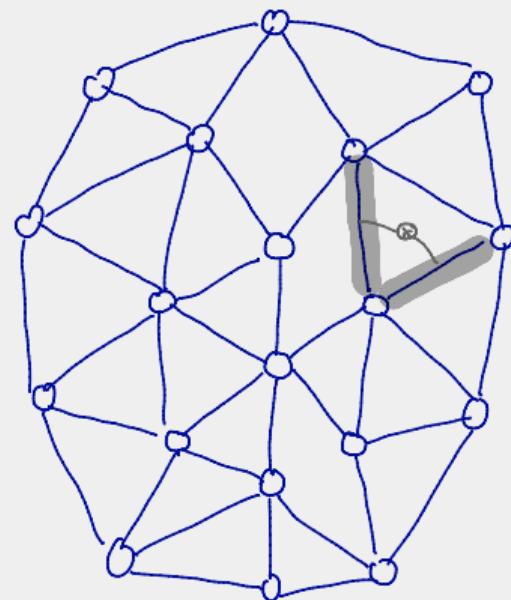
$$\text{xc}(P(\mathcal{M})) \leq \text{xc}(\text{P}_{\text{spt}}(G)) + O(|\bar{M}| \cdot |E|)$$

Disjunction yields:

$$\text{xc}(P(\mathcal{M})) \leq O(|\bar{M}|) \cdot \text{xc}(\text{P}_{\text{spt}}(G))$$

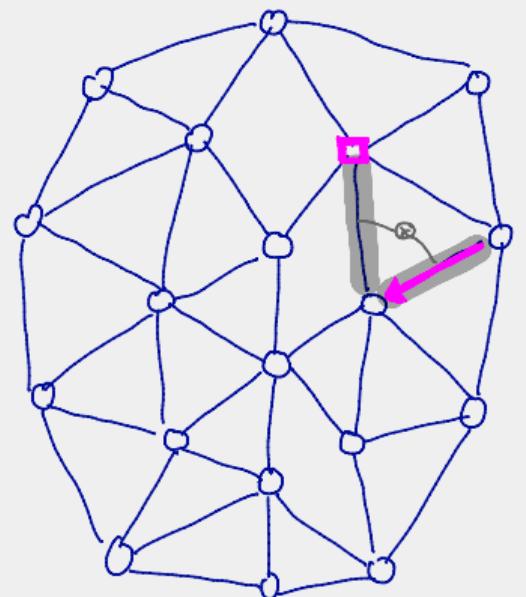
A Single Product

One adjacent pair



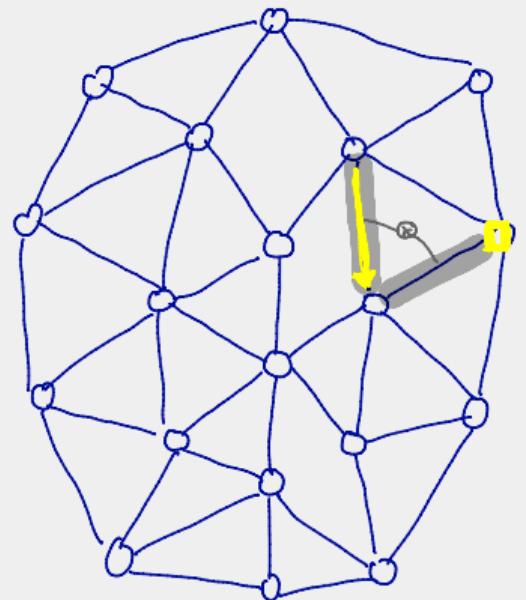
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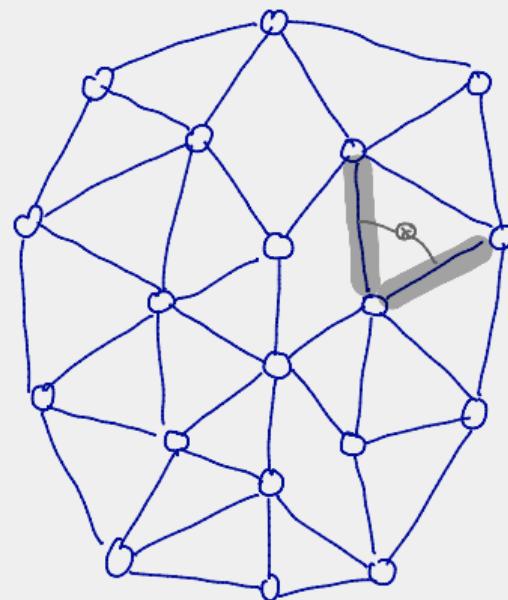
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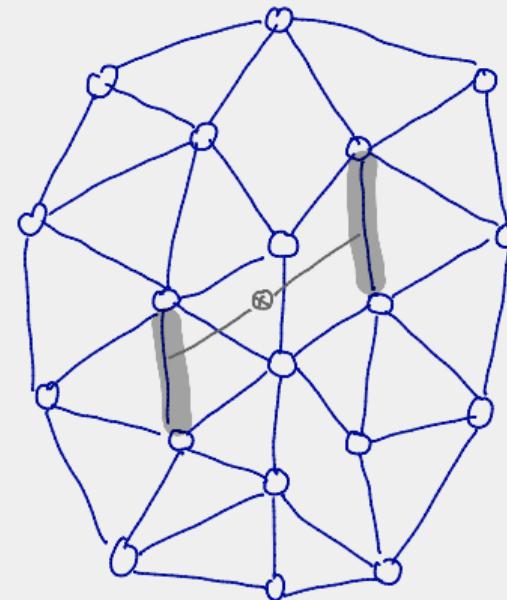


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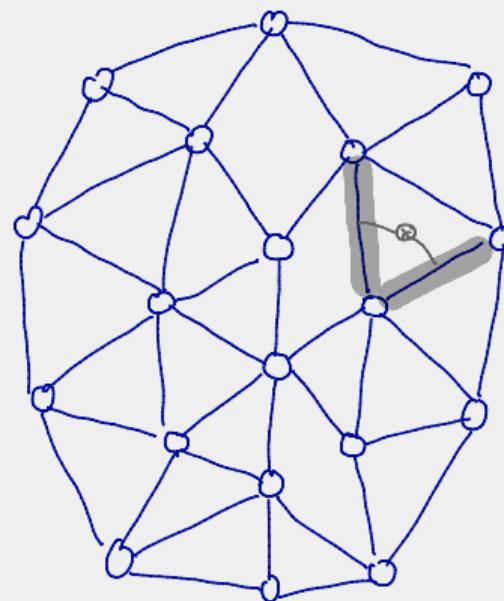


One non-adjacent pair

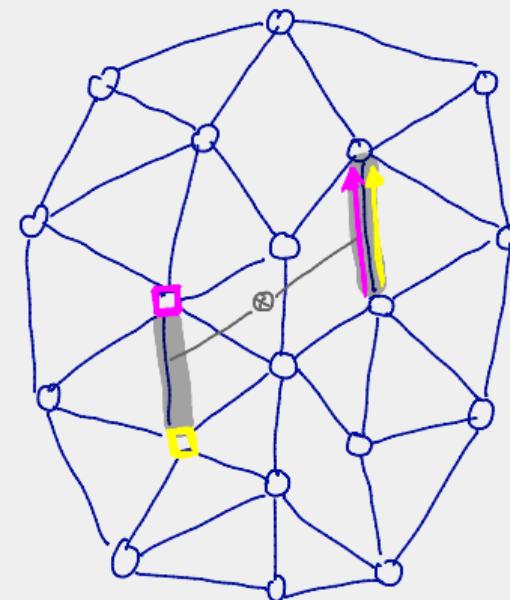


A Single Product

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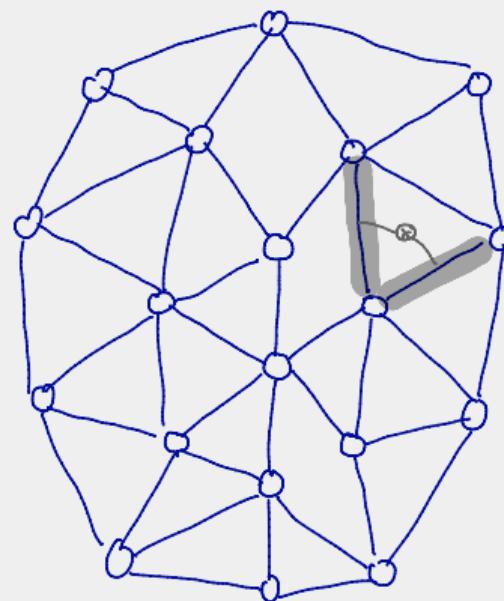


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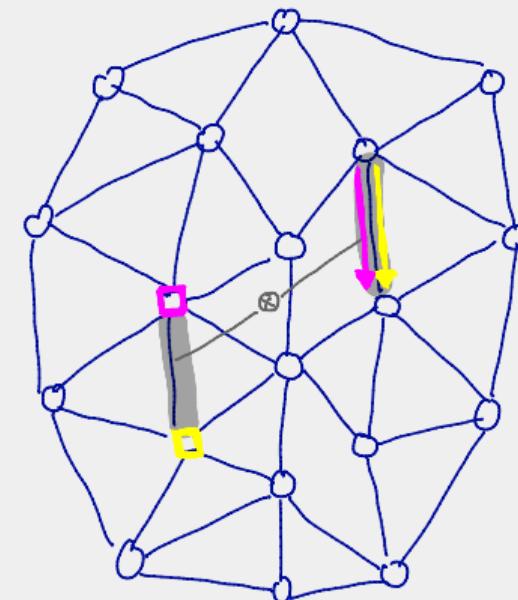


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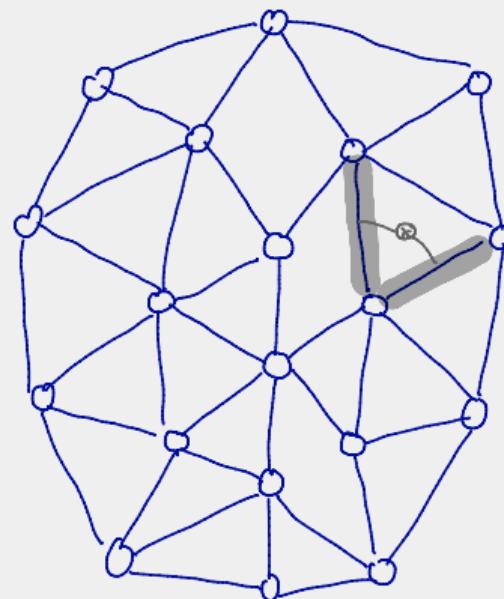


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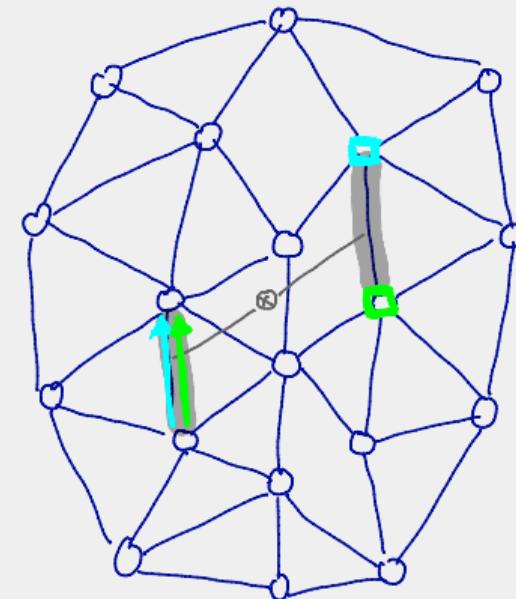


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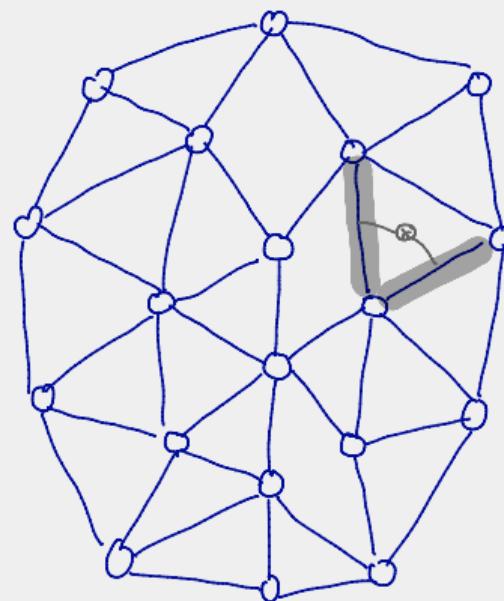


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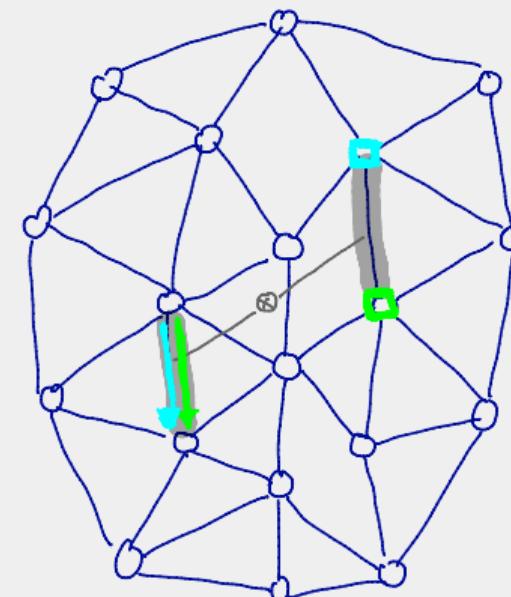


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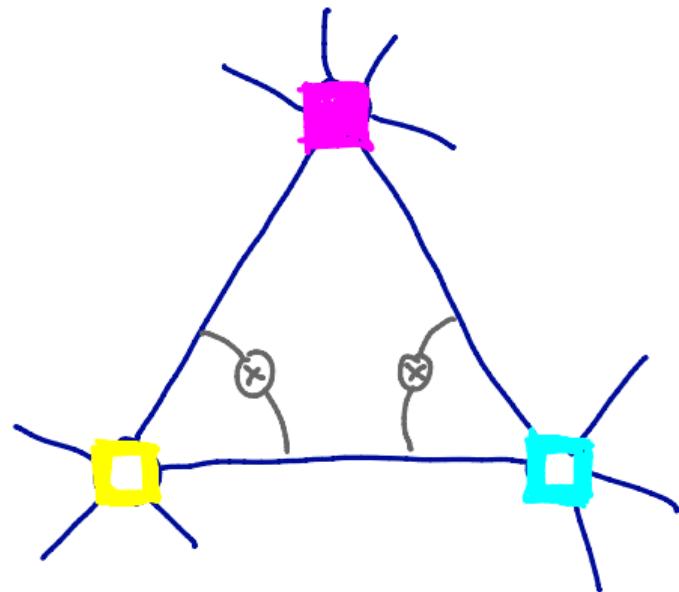
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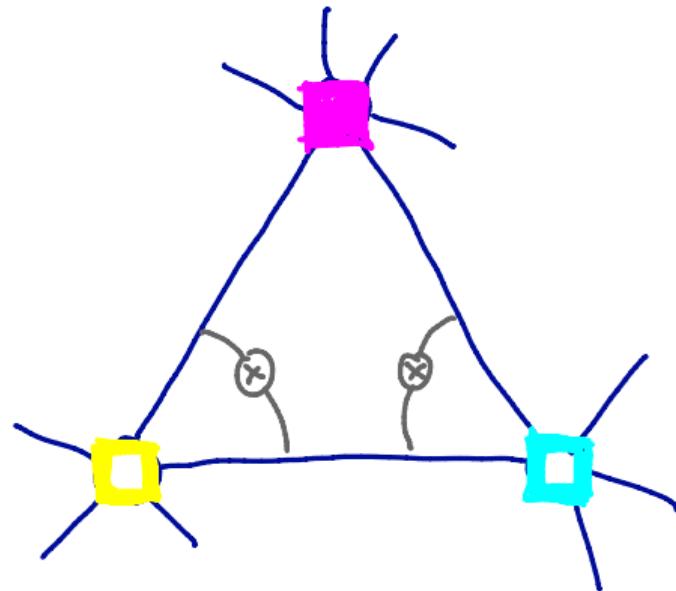
One non-adjacent pair



A Strengthened Relaxation



A Strengthened Relaxation



Friesen & K 2017

The relaxation provided by

$$Q(\mathcal{M}_1, \mathcal{M}) \cap \cdots \cap Q(\mathcal{M}_t, \mathcal{M})$$

with re-used *z*-variables is in general stronger than $P(\mathcal{M}_1, \mathcal{M}) \cap \cdots \cap P(\mathcal{M}_t, \mathcal{M})$.

An Approach to Obtain Relaxations

For polytope $P \subseteq \mathbb{R}^d$

- ① Choose $P_1, \dots, P_r \supseteq P$.
- ② Construct extensions Q_i of P_i with preimages $z_i(v)$ of the vertices v of P .
- ③ Identify valid linear inequalities for the $(v, z_1(v), \dots, z_r(v))$'s defining a polyhedron S .

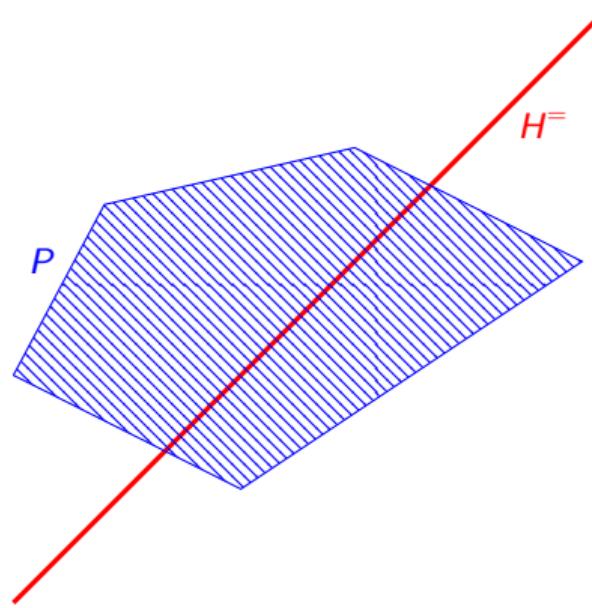
Then $\{(x, z_1, \dots, z_r) \in S : z_i \in Q_i\}$ is an extension of a polyhedron R with

$$P \subseteq R \subseteq P_1 \cap \dots \cap P_r.$$

Outline

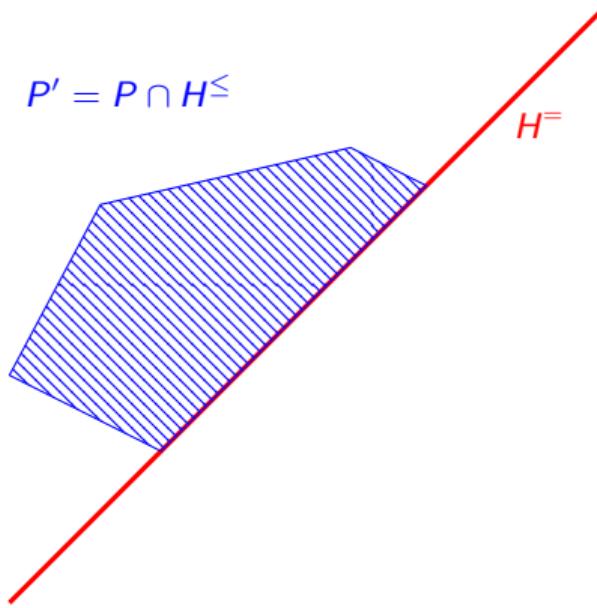
- ① The Concept
- ② Disjunctive Programming
- ③ Dynamic Programming
- ④ Branched Polyhedral Systems
- ⑤ Dualization
- ⑥ Redundant Information
- ⑦ Reflections

The Reflection Operation

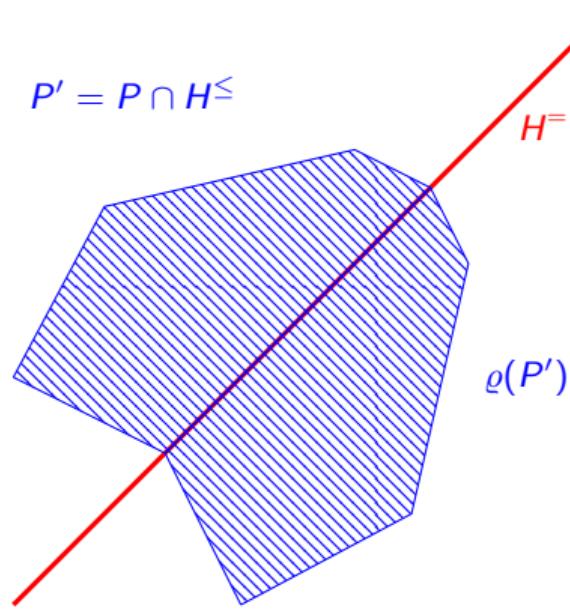


The Reflection Operation

$$P' = P \cap H^{\leq}$$



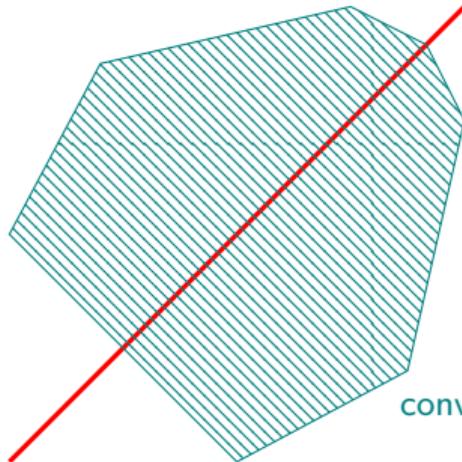
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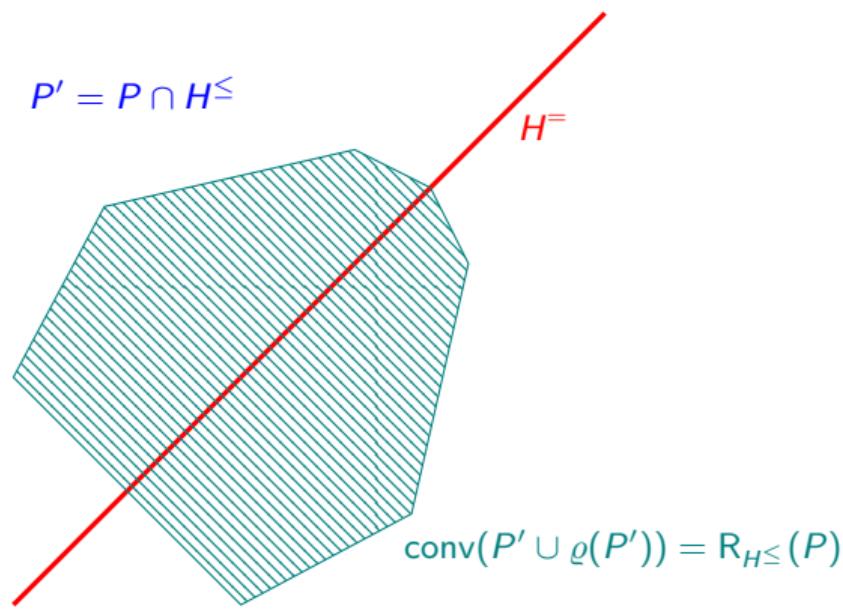
$$P' = P \cap H^{\leq}$$

$$H^=$$



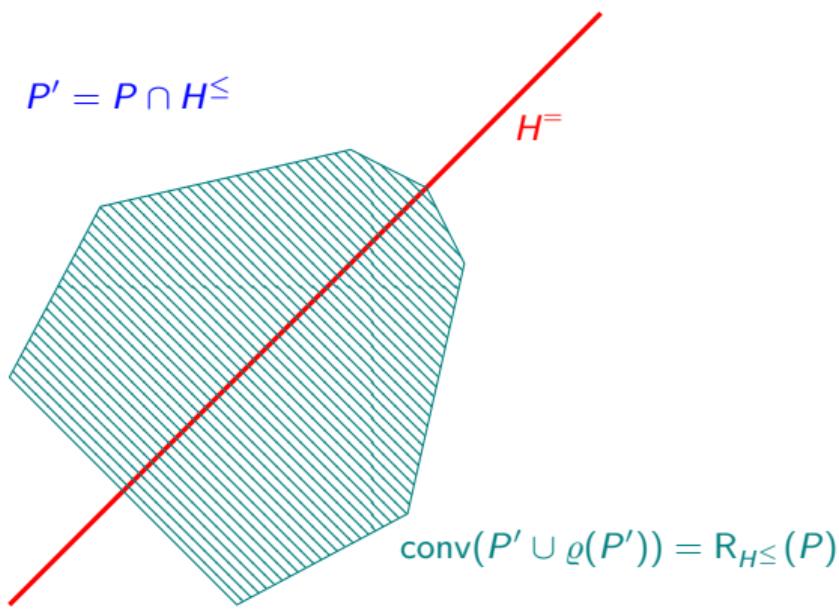
$$\text{conv}(P' \cup \varrho(P')) = R_{H^{\leq}}(P)$$

The Reflection Operation



- $R_{H^{\leq}}(P) = \{x + \lambda a : x \in P, \langle a, x \rangle \leq \langle a, x + \lambda a \rangle \leq 2b - \langle a, x \rangle\}$

The Reflection Operation



- $R_{H^<}(P) = \{x + \lambda a : x \in P, \langle a, x \rangle \leq \langle a, x + \lambda a \rangle \leq 2b - \langle a, x \rangle\}$
- Thus: $\text{xc}(R_{H^<}(P)) \leq \text{xc}(P) + 2$

Sequences of Reflection Operations

Consequence

For each sequence $H_1^{\leq}, \dots, H_r^{\leq} \subseteq \mathbb{R}^n$ of halfspaces and for each polytope $P \subseteq \mathbb{R}^n$, the polytope

$$\mathcal{R}_{H_1^{\leq}, \dots, H_r^{\leq}}(P) = R_{H_r^{\leq}}(R_{H_{r-1}^{\leq}}(\dots R_{H_1^{\leq}}(P)\dots))$$

satisfies

$$xc(\mathcal{R}_{H_1^{\leq}, \dots, H_r^{\leq}}(P)) \leq xc(P) + 2r.$$

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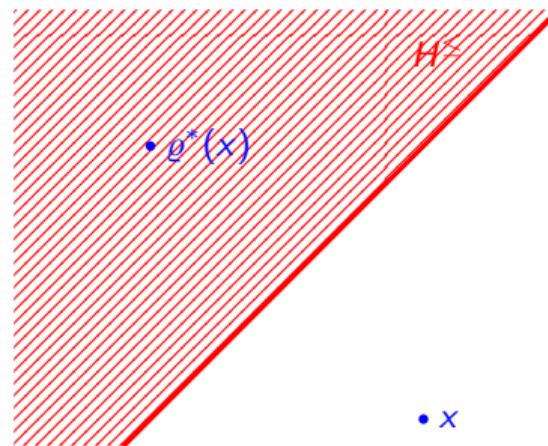
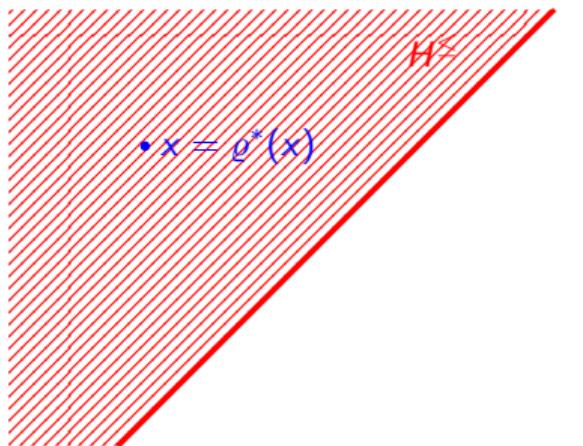
Task for target polytope Q

Find (and describe) P , design sequence $H_1^{\leq}, \dots, H_r^{\leq}$, and prove

$$Q = \mathcal{R}_{H_r^{\leq}, \dots, H_1^{\leq}}(P).$$

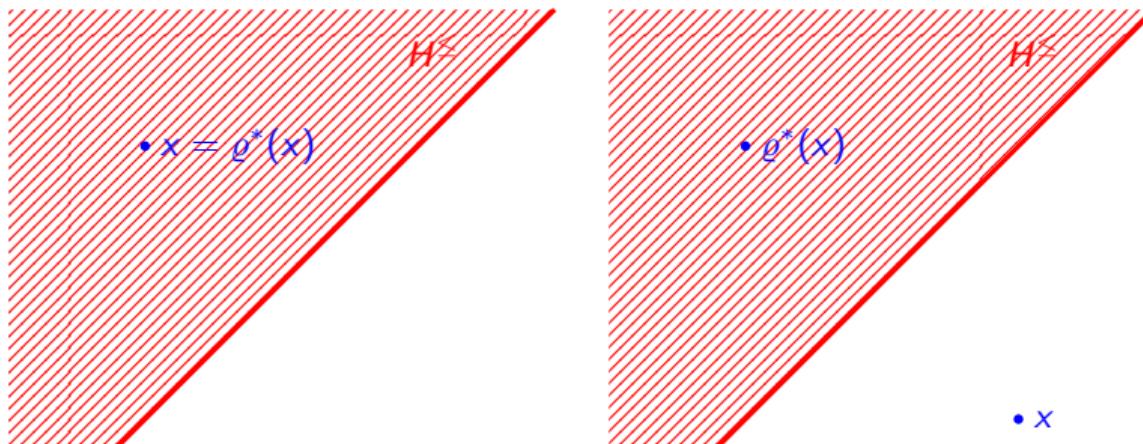
Conditional Reflections

Define $\varrho^* : \mathbb{R}^n \rightarrow \mathbb{R}^n$ via $\varrho^*(x) = \begin{cases} x & \text{if } x \in H^\leq \\ \varrho(x) & \text{otherwise} \end{cases}$.



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$$\varrho^*(x) \in P \quad \Rightarrow \quad x \in R_{H^\leq}(P)$$

Generating the Target Polytope

K & PASHKOVICH 11

Let

- $Q = \text{conv}(W)$ be some (target) polytope,
- $H_1^{\leq}, \dots, H_r^{\leq} \subseteq \mathbb{R}^n$ be a sequence of halfspaces, and
- ϱ_i (and ϱ_i^*) the associated (conditional) reflections.

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- ② $\varrho_1^*(\varrho_2^*(\cdots(\varrho_r^*(w)\cdots)) \in P$ for all $w \in W$.

Reflection Groups

Finite Reflection Group G

A *finite* group generated by a (finite) family $\varrho^{H_i} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ ($i \in I$) of reflections at hyperplanes $\mathbf{0} \in H_i \subseteq \mathbb{R}^n$.

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Coxeter-Arrangement of G

The set of all hyperplanes $\mathbf{0} \in H \subseteq \mathbb{R}^n$ with $\varrho^H \in G$.

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G acts transitively on the *regions* of the Coxeter-Arrangement.

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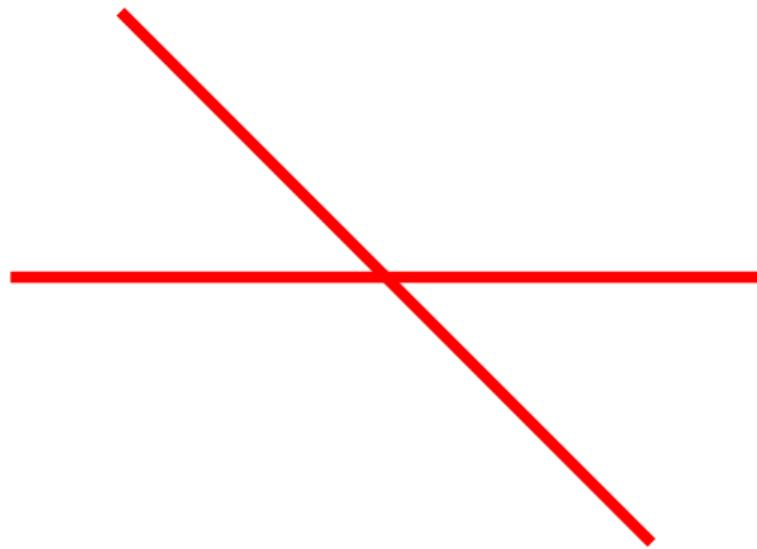
Action on \mathbb{R}^n

G acts transitively on the *regions* of the Coxeter-Arrangement.

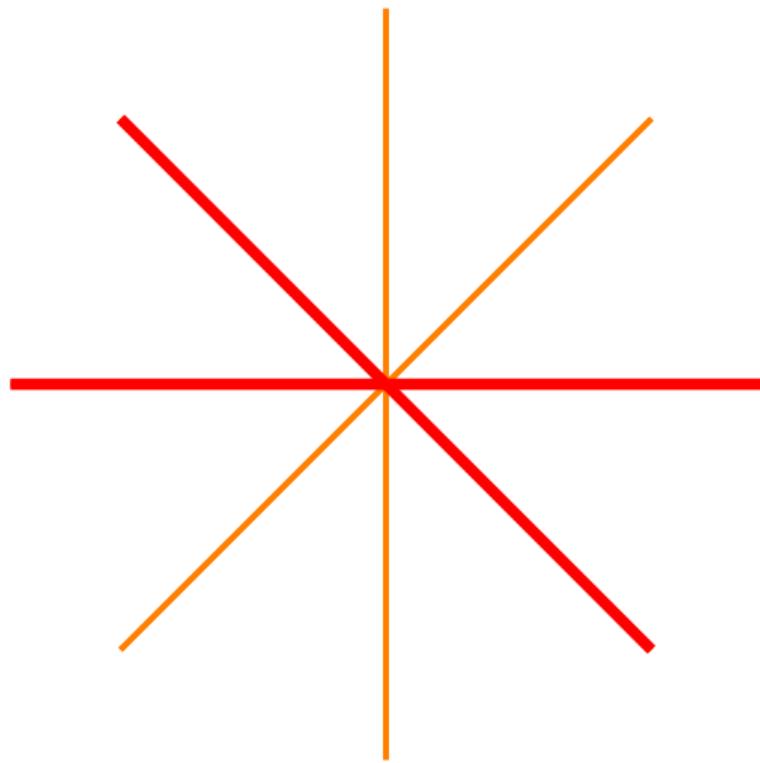
G -Permutahedron of polytope P in one region

$$\mathsf{P}_{\text{perm}}^G(P) = \text{conv}(\bigcup_{g \in G} g.P)$$

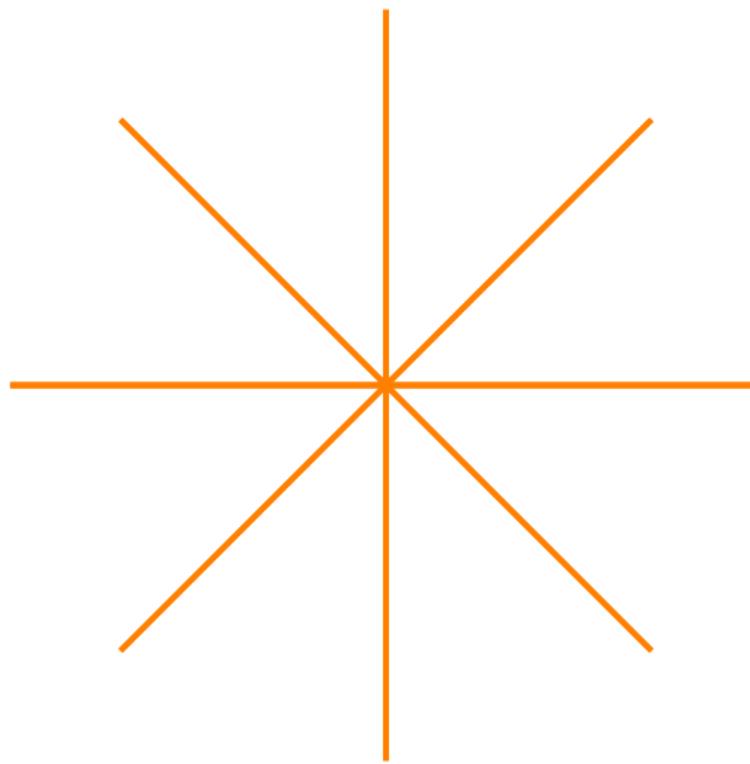
Example: $I_2(4)$



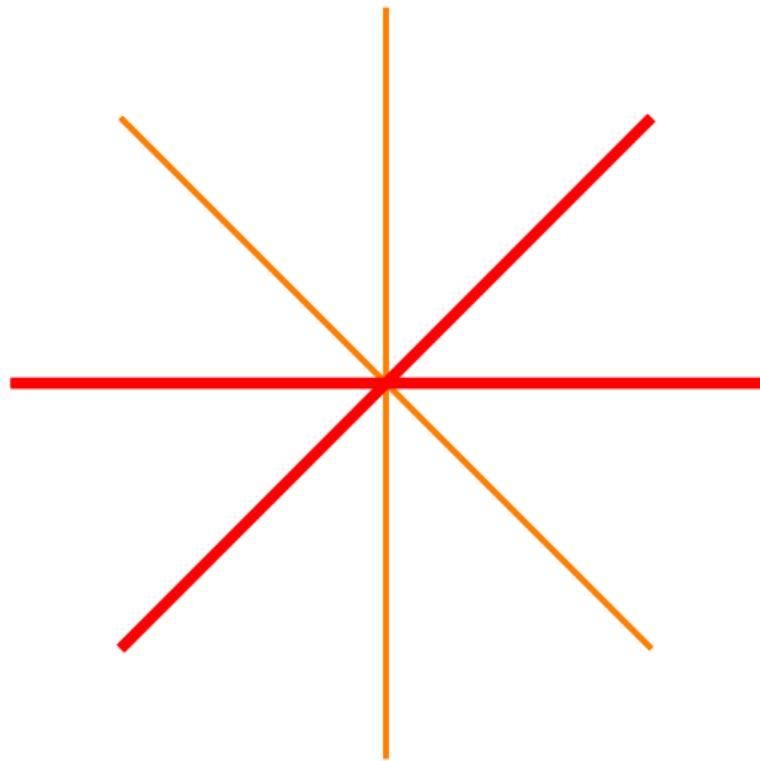
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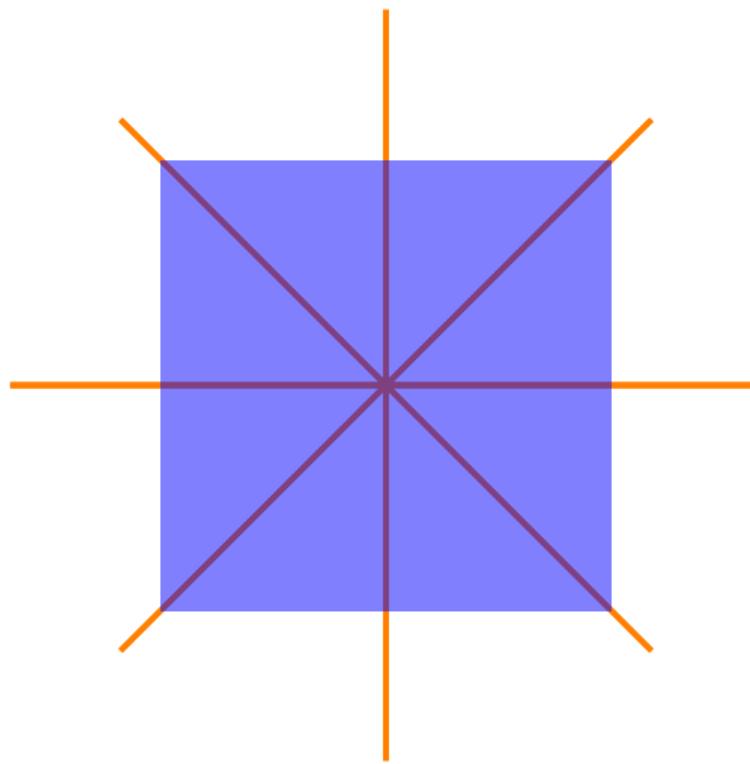
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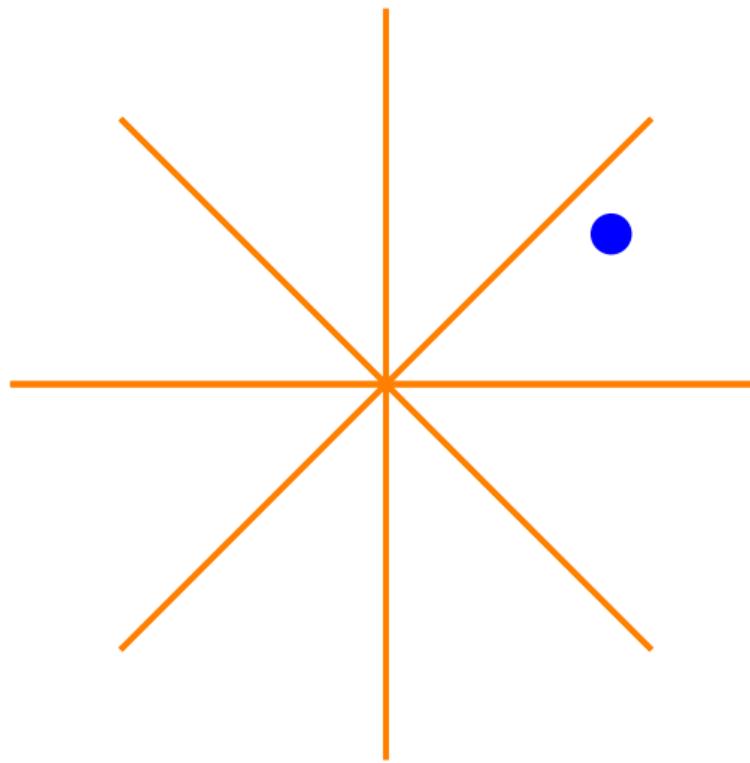
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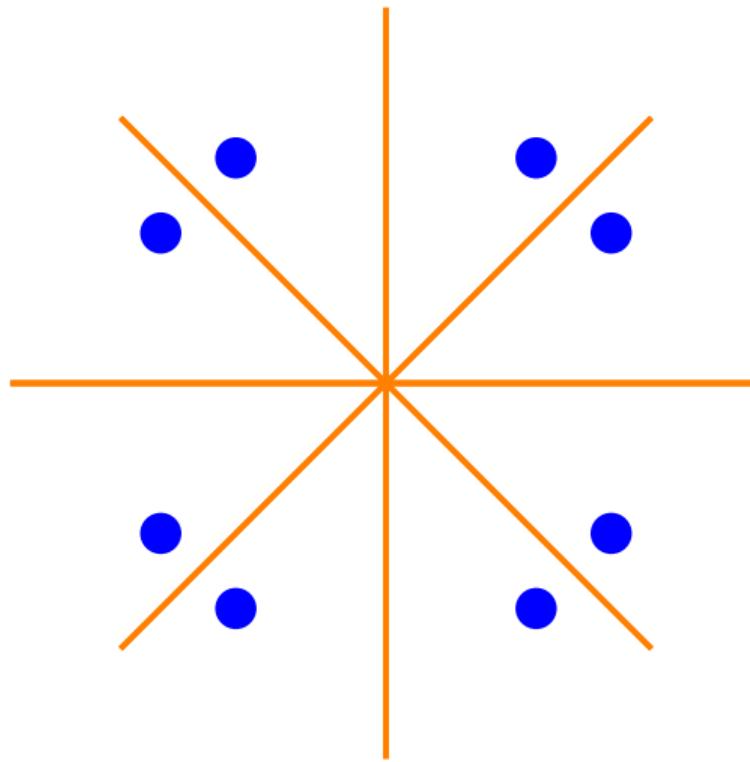
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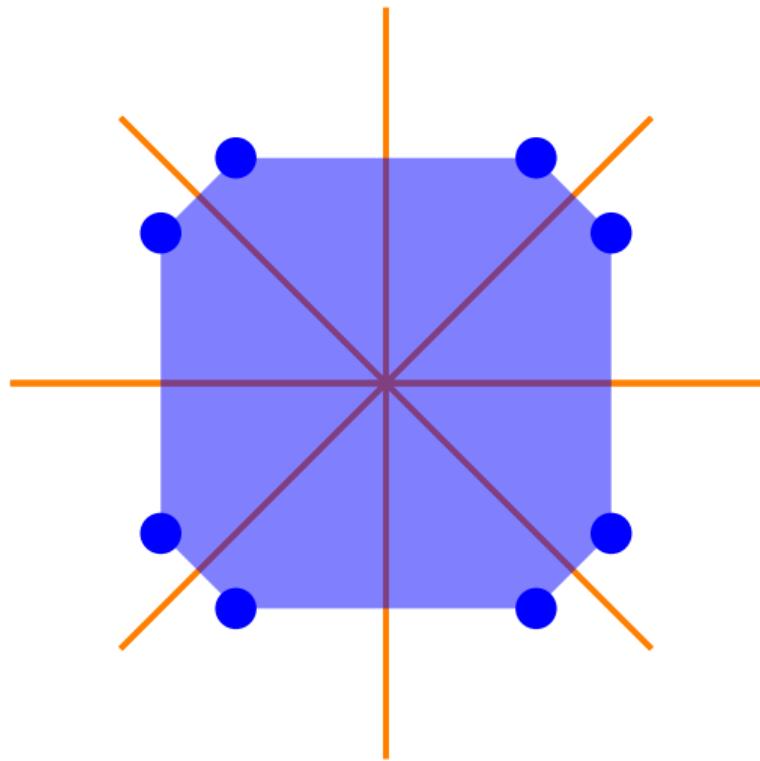
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Classification of Irreducible Reflection Groups

Name	Dynkin Diagram	Regular Polytope
$I_2(m)$		m -gon
A_{n-1}		$(n - 1)$ -simplex
B_n		n -cube, n -cross polytope
D_n		
E_6		
E_7		
E_8		
F_4		24-cell
H_3		dodecahedron, icosahedron

The Reflection Group $I_2(m)$

The group

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$$P_{\text{perm}}^{I_2(m)}(P) = \mathcal{R}_{H_{r\pi/m}^\leq, \dots, H_{4\pi/m}^\leq, H_{2\pi/m}^\leq, H_{\pi/m}^\leq}(P) \quad \text{with } r = \lceil \log(m) \rceil$$

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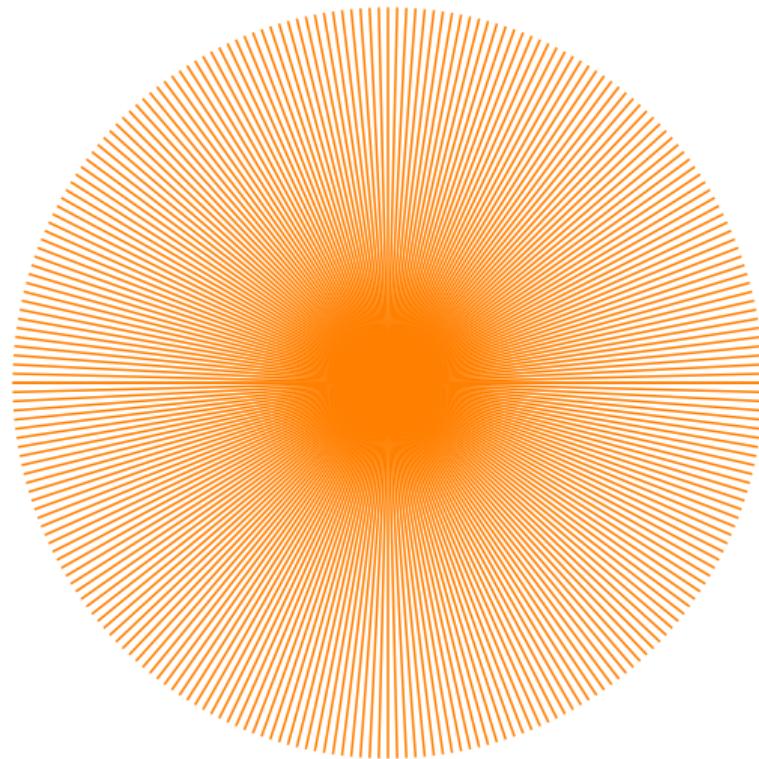
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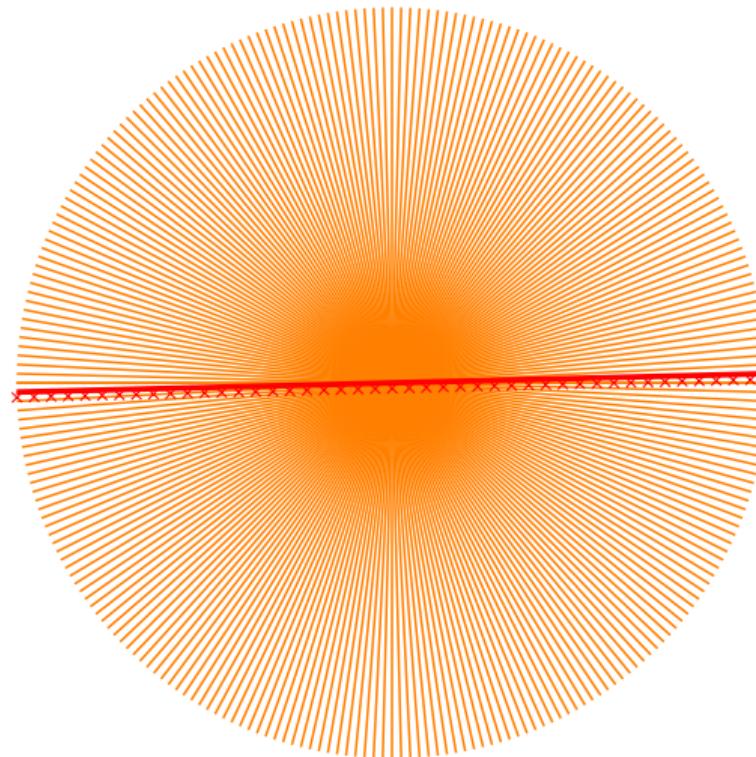
Thus we have:

$$xc(P_{\text{perm}}^{I_2(m)}(P)) \leq xc(P) + 2\lceil \log(m) \rceil + 2$$

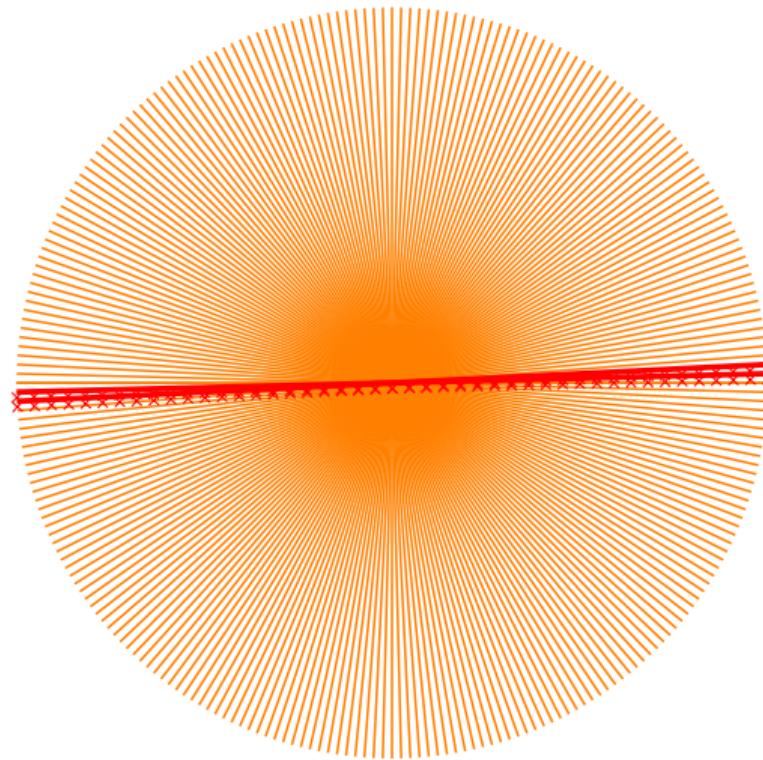
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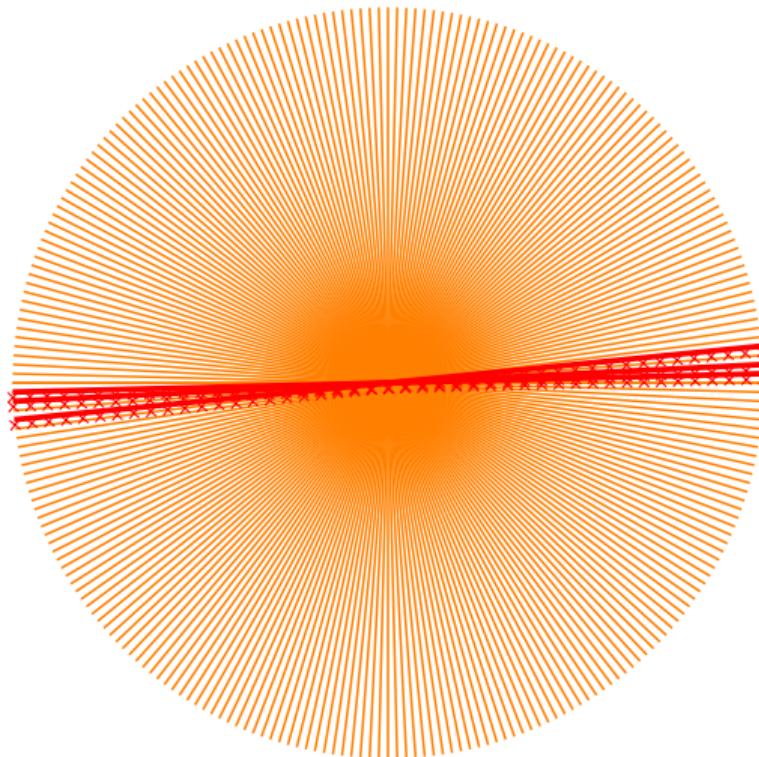
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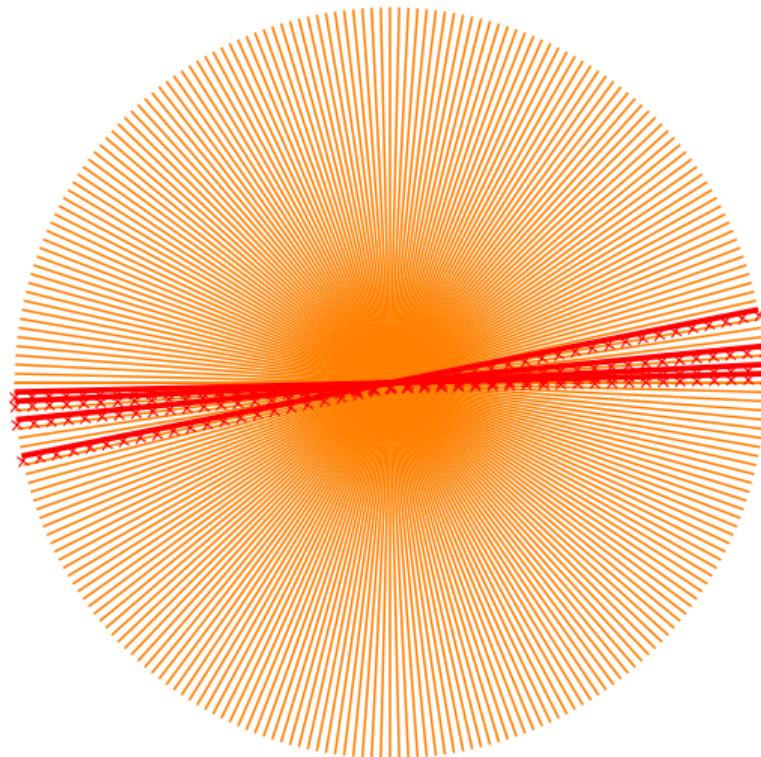
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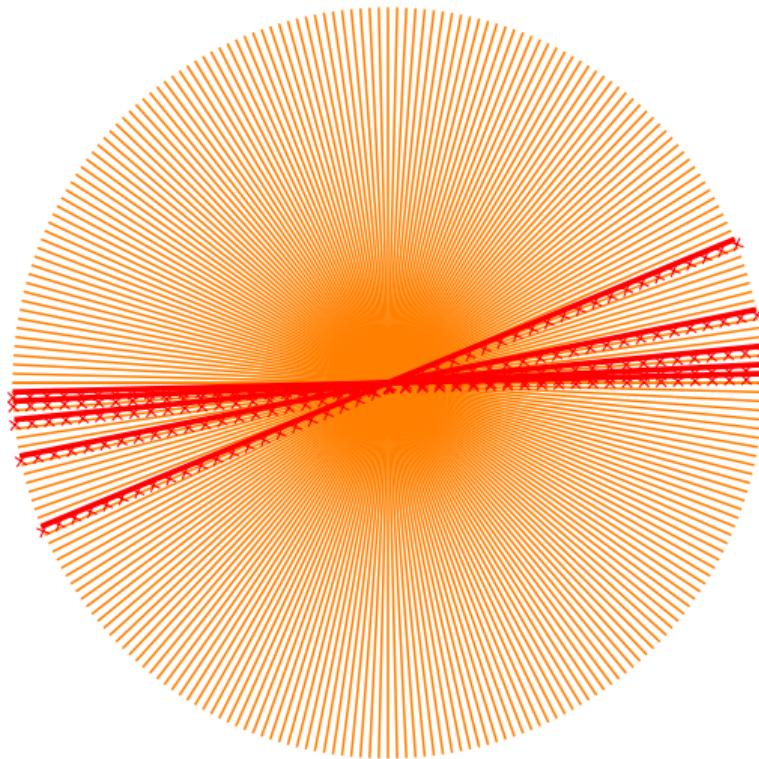
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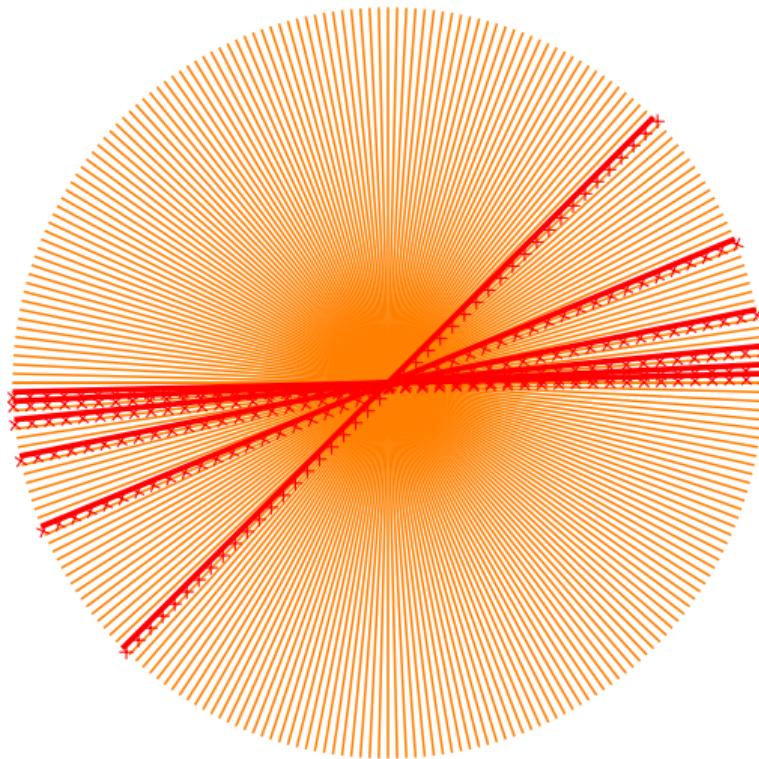
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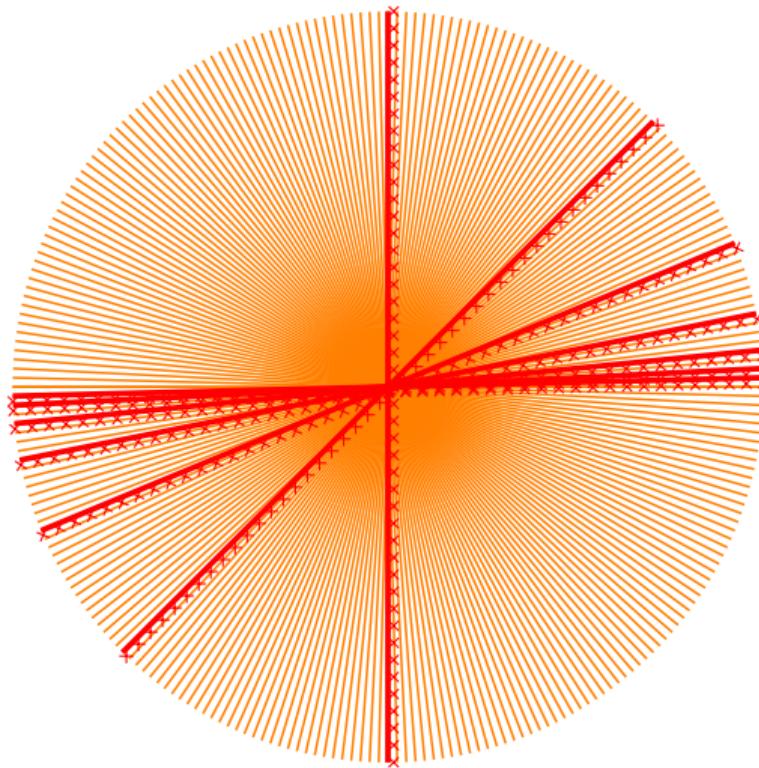
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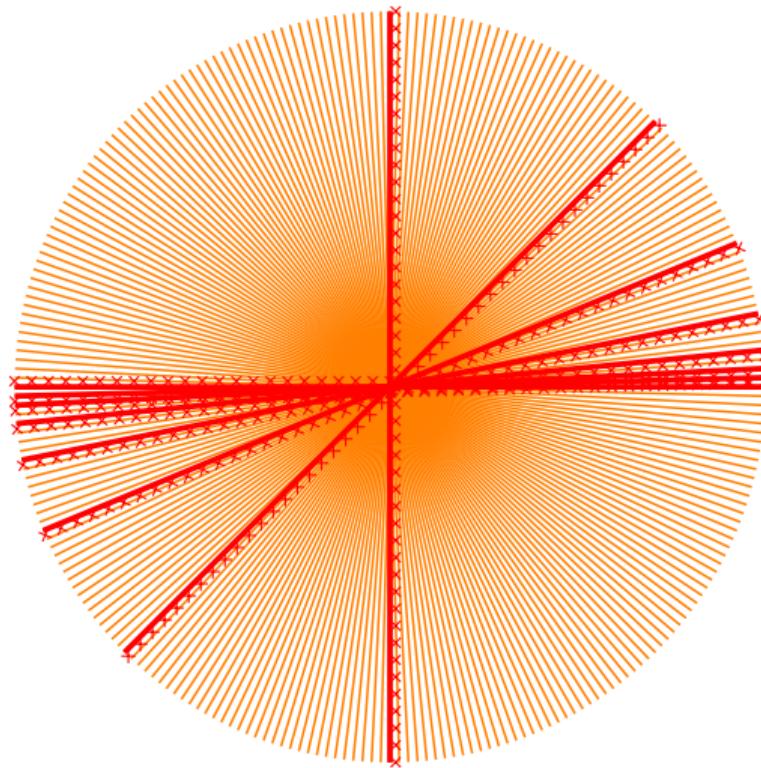
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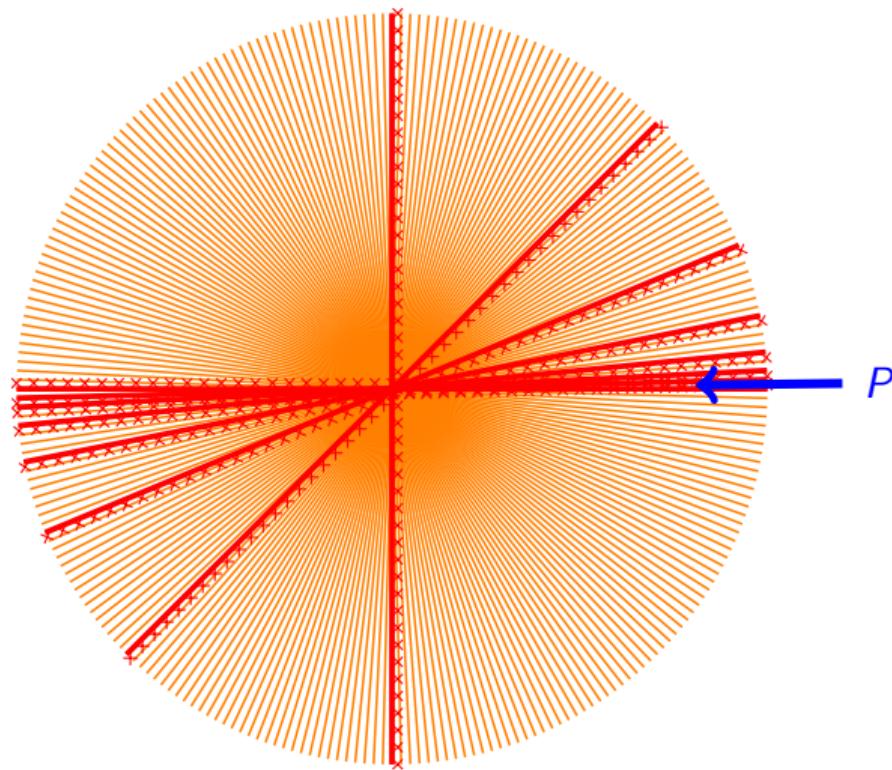
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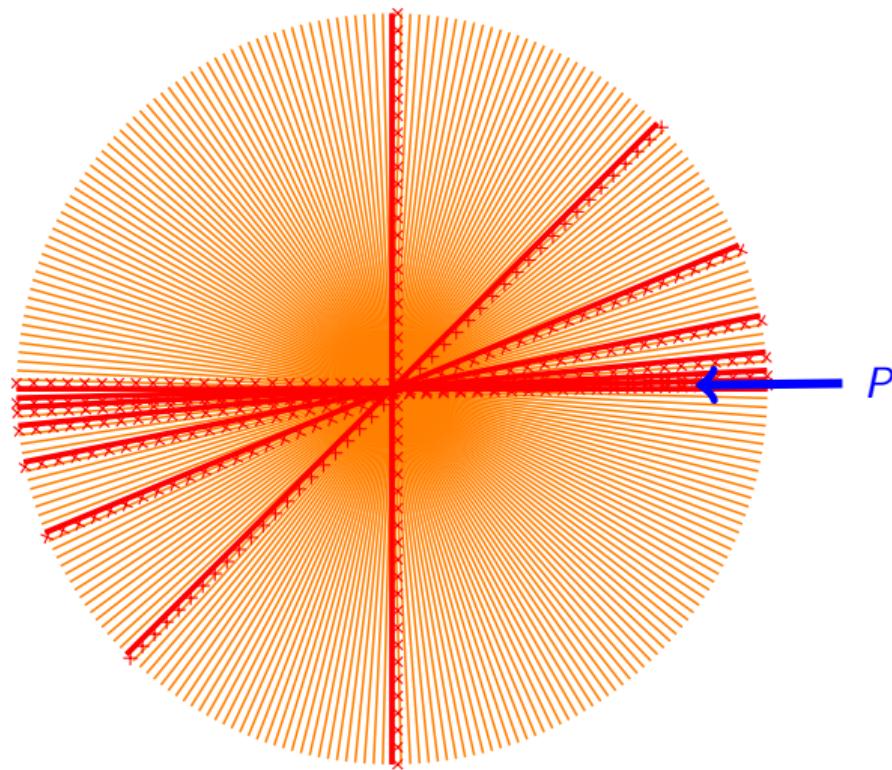
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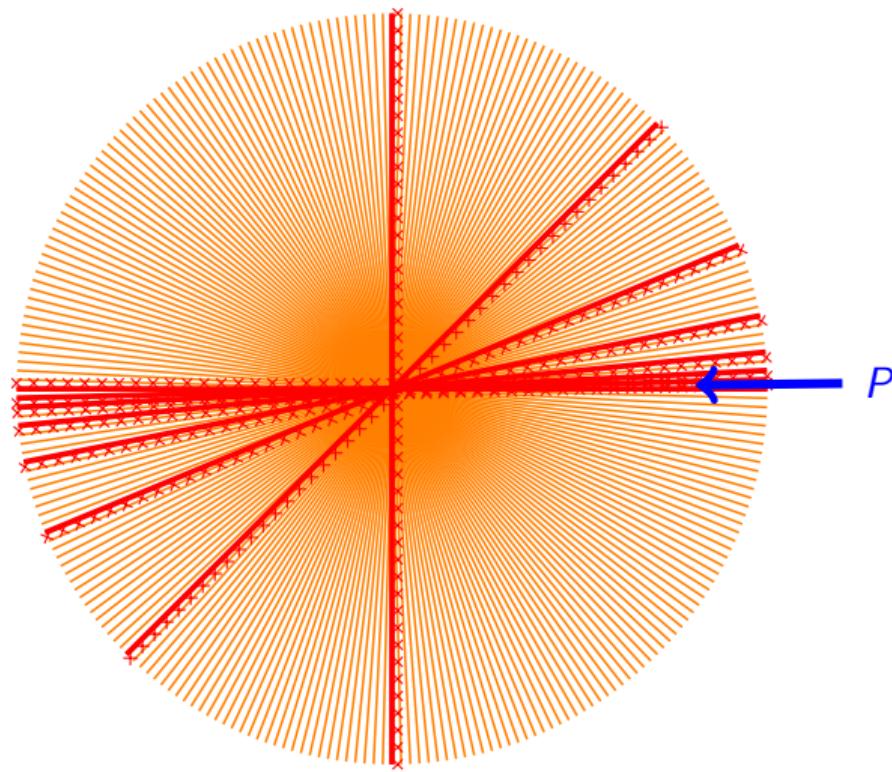
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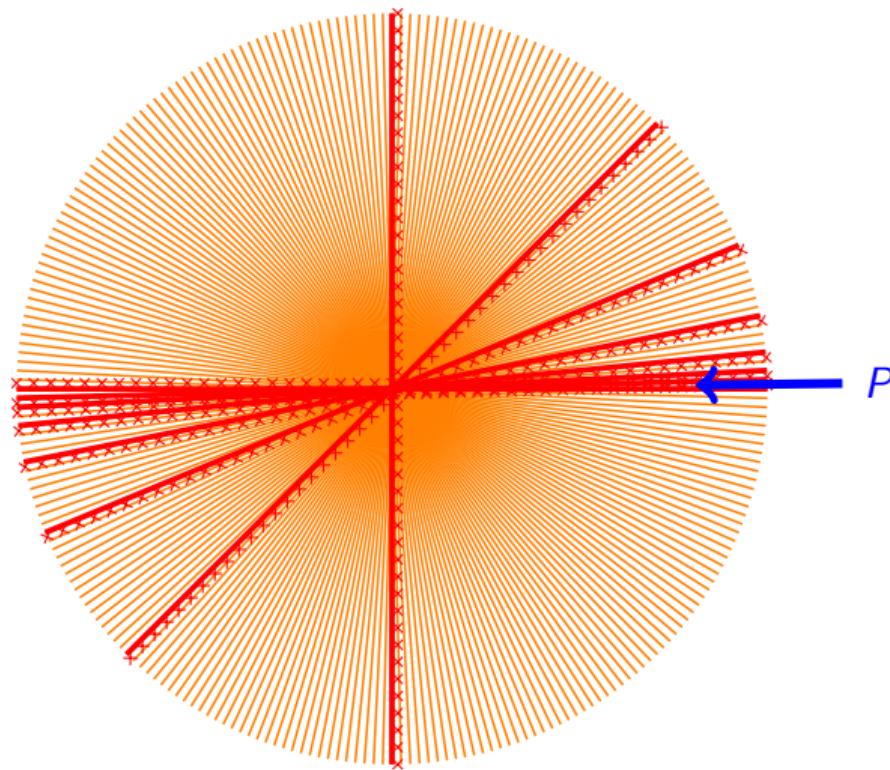


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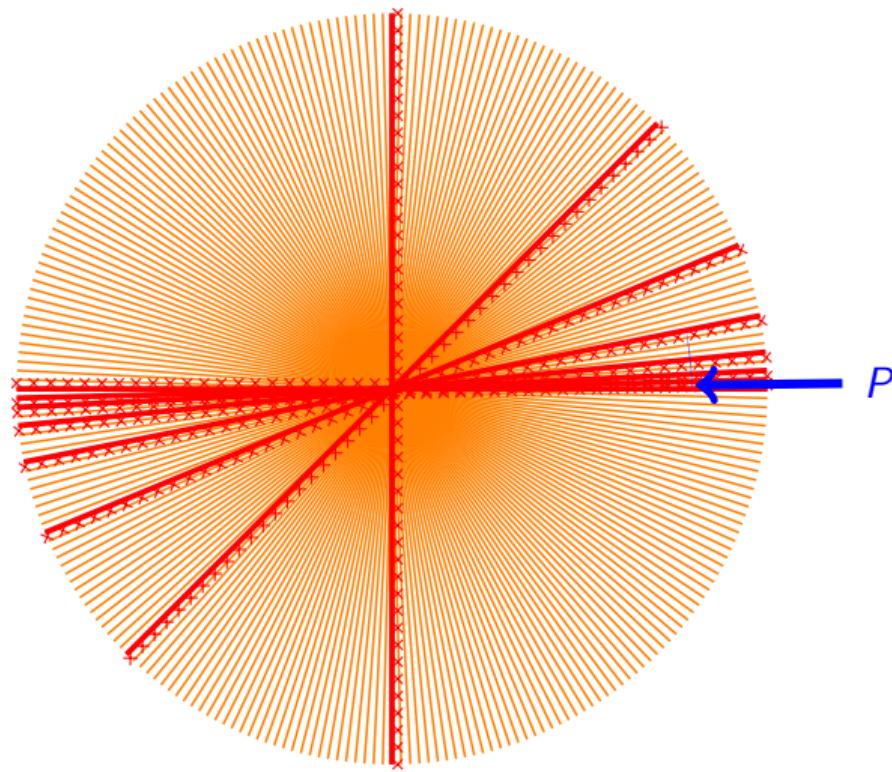
$$\mathcal{R}_{H_{\pi/128}^{\leq}}(P)$$

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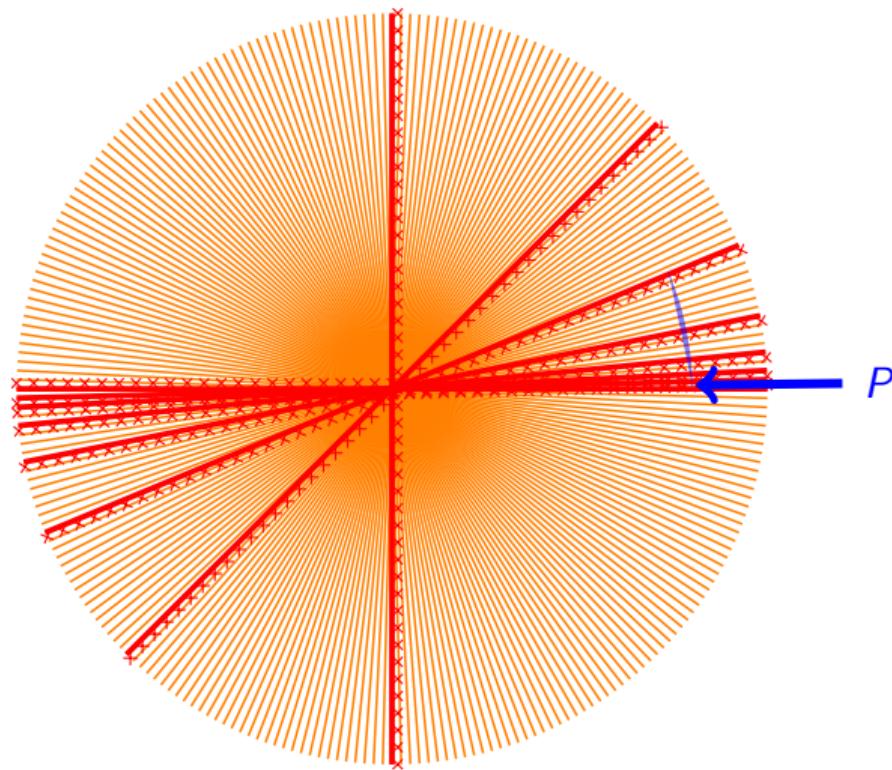
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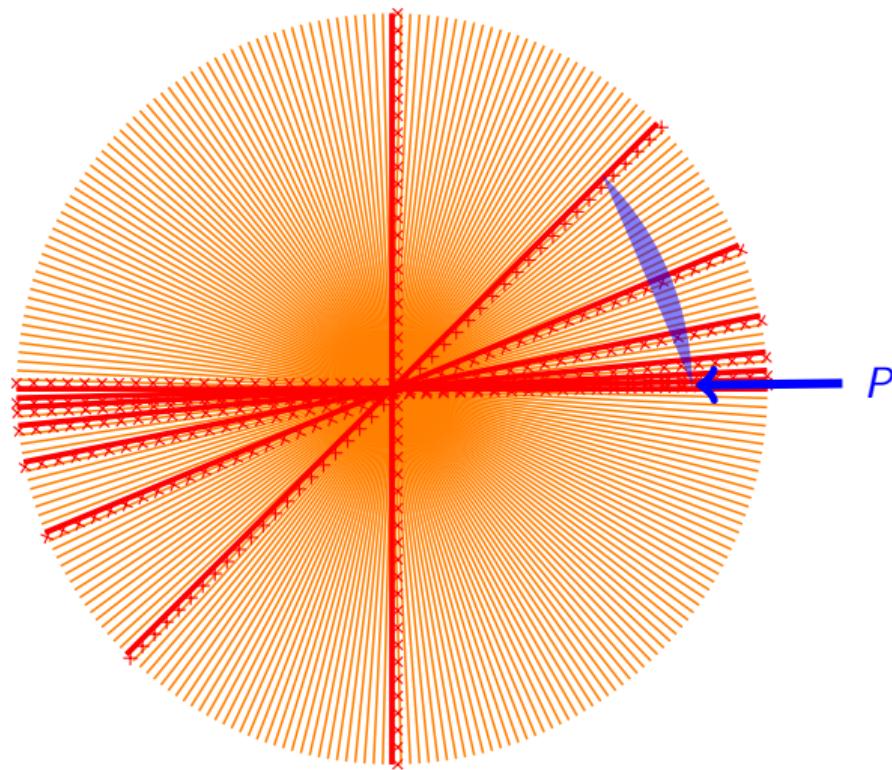
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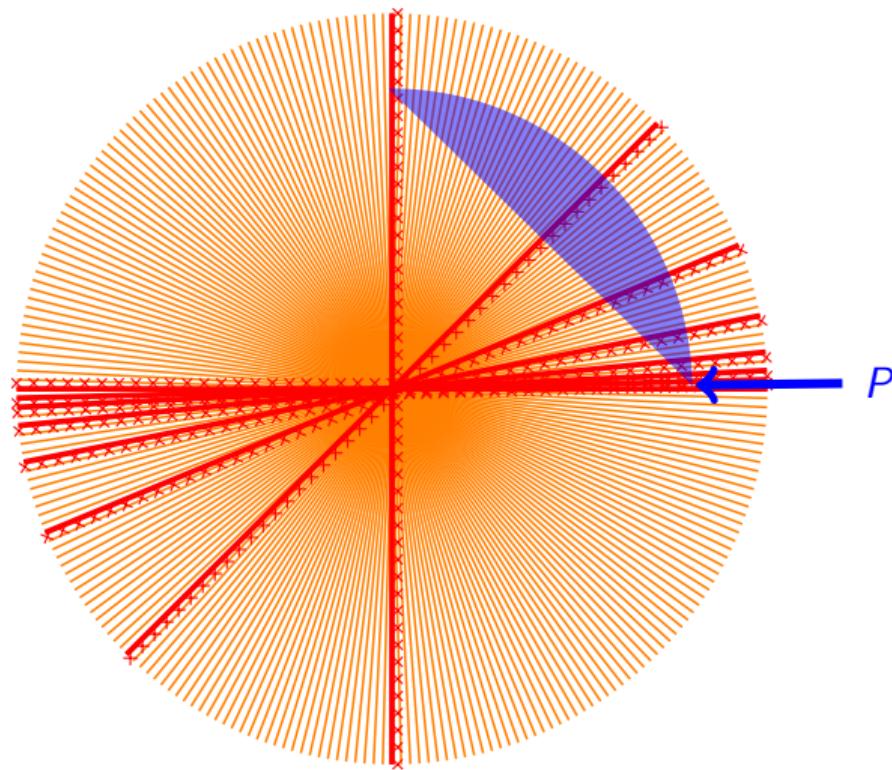
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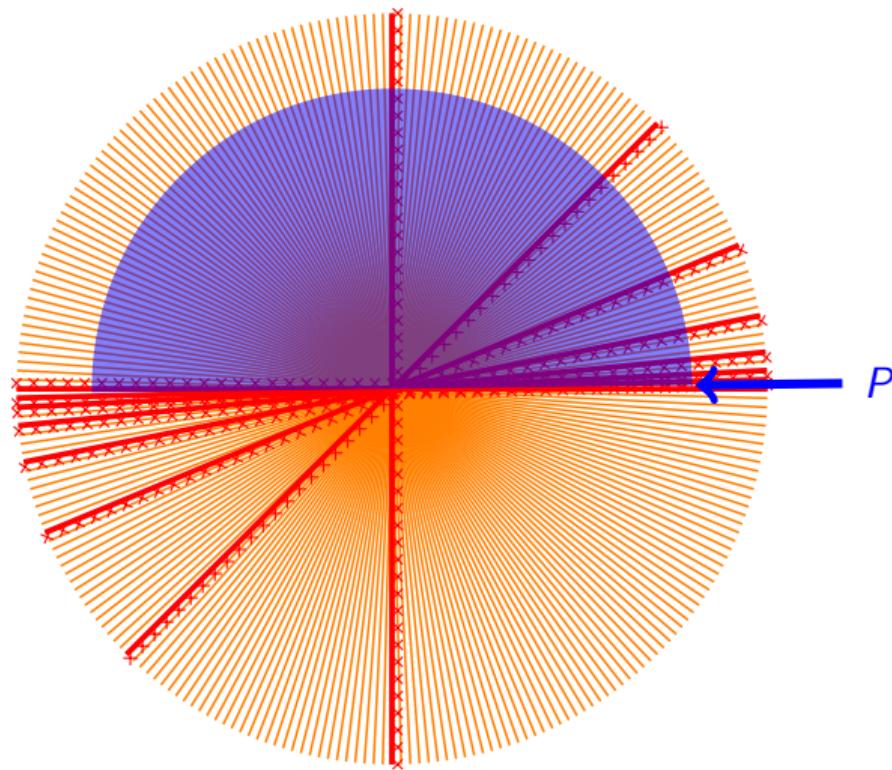
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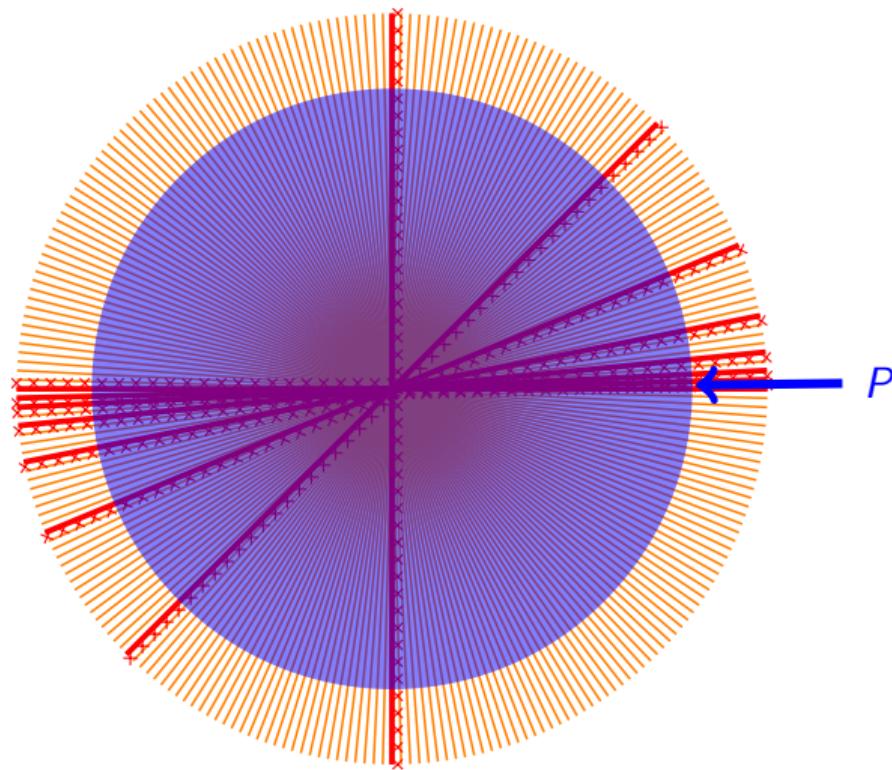
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Example: $I_2(128)$



$$\mathcal{R}_{H_{64\pi/128}^{\leq}, H_{32\pi/128}^{\leq}, H_{16\pi/128}^{\leq}, H_{8\pi/128}^{\leq}, H_{4\pi/128}^{\leq}, H_{2\pi/128}^{\leq}, H_{\pi/128}^{\leq}}(P)$$

Example: $I_2(128)$



$$\mathcal{R}_{H_{128\pi/128}^{\leq}, H_{64\pi/128}^{\leq}, H_{32\pi/128}^{\leq}, H_{16\pi/128}^{\leq}, H_{8\pi/128}^{\leq}, H_{4\pi/128}^{\leq}, H_{2\pi/128}^{\leq}, H_{\pi/128}^{\leq}} (P)$$

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- $H_{k,\ell} = H^=(\mathbb{e}_k - \mathbb{e}_\ell, 0), H_{k,\ell}^\leq = H^\leq(\mathbb{e}_k - \mathbb{e}_\ell, 0) \subseteq \mathbb{R}^n$

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Conditional reflections

$$\tau_{>}^{k,\ell}(y) = \begin{cases} \tau^{k,\ell}(y) & \text{if } y_k > y_\ell \\ y & \text{otherwise} \end{cases}$$

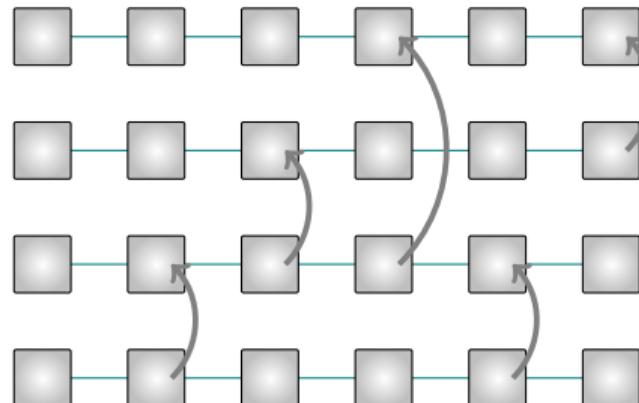
Sorting Networks

Sorting Network

Sequence $(k_1, \ell_1), \dots, (k_r, \ell_r)$ with

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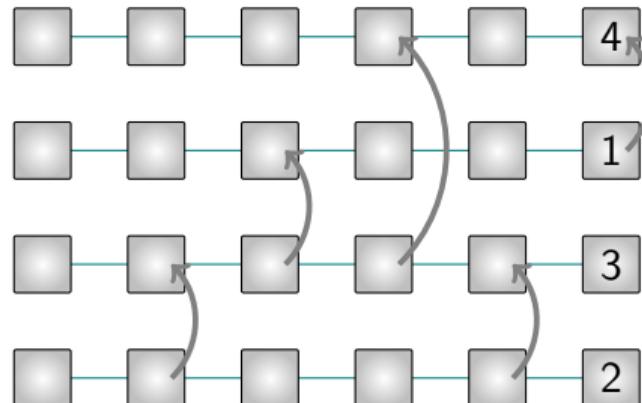
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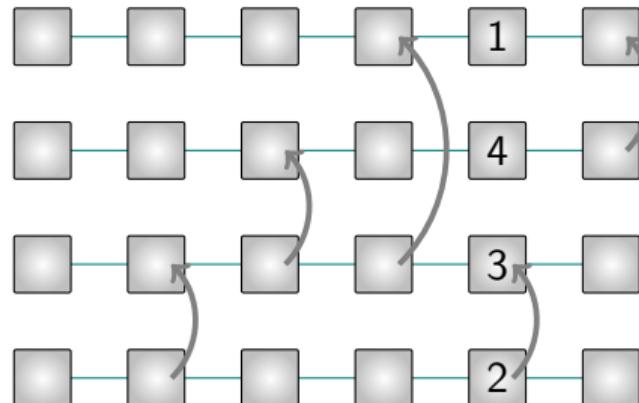
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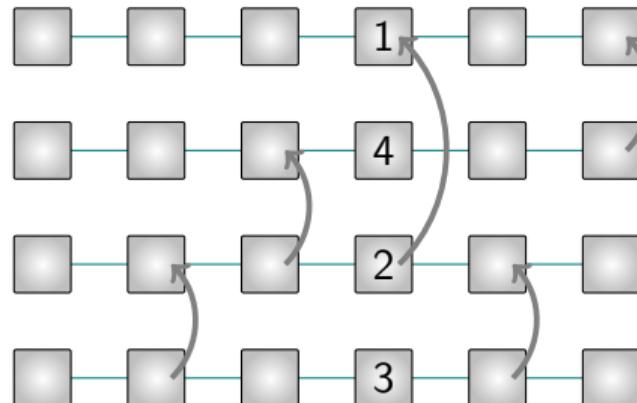
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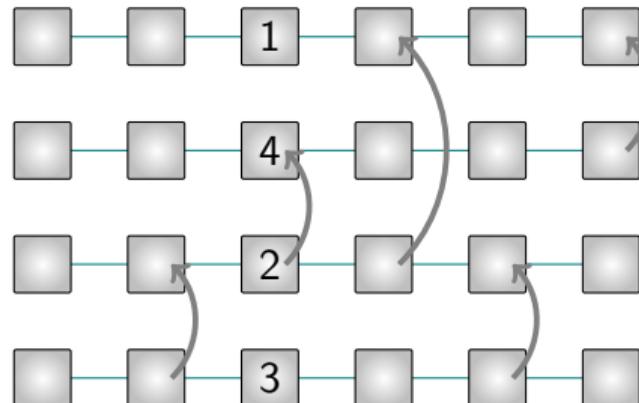
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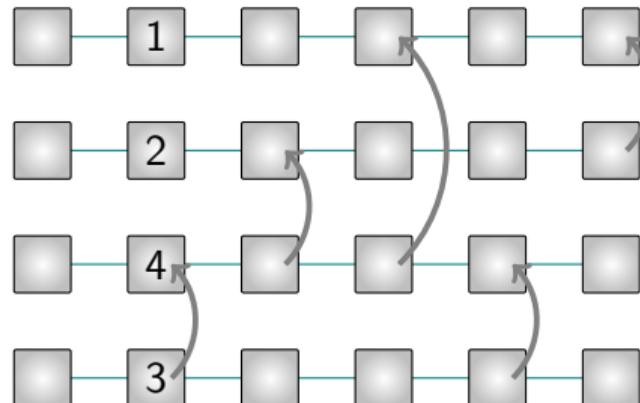
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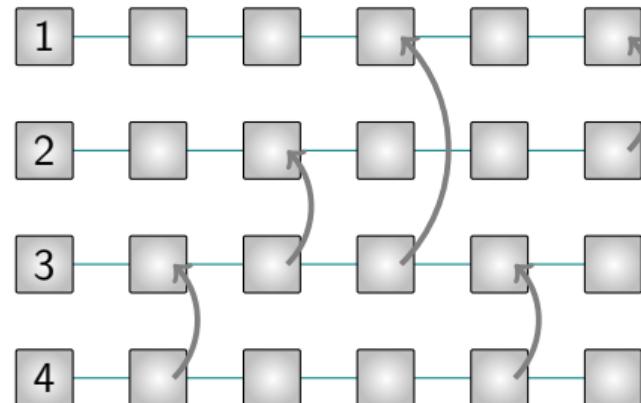
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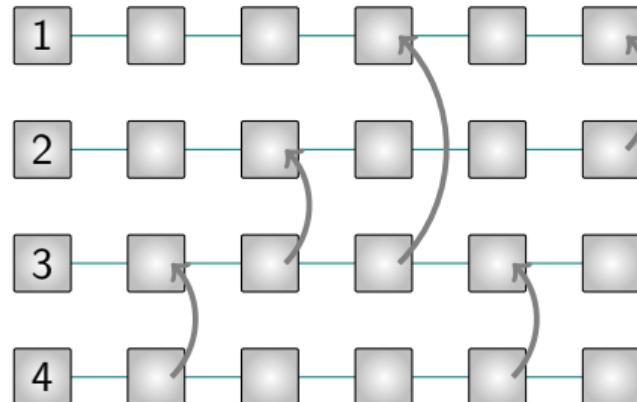
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AJTAI, KOMLÓS & SZEMERÉDI 1983

There are sorting networks of size $r = O(n \log n)$.

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GOEMANS 2009

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General G -Permutahedra

K & PASHKOVICH 11

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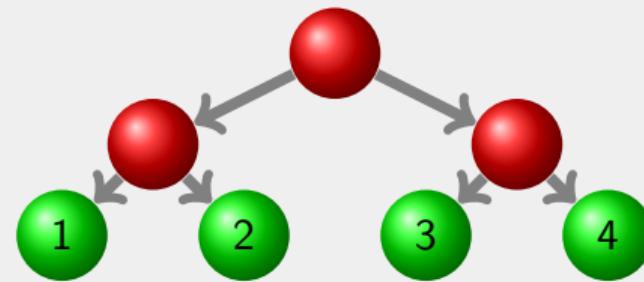
then we have:

$$\text{xc}(\mathcal{P}_{\text{perm}}^G(P)) \leq \text{xc}(P) + O(\log m + n \log n)$$

(where m is the largest number such that $I_2(m)$ is a factor of G).

Huffman Polytopes

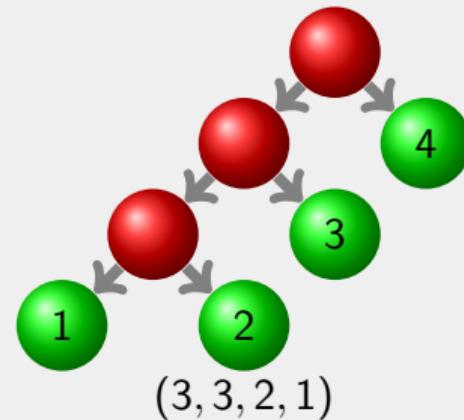
The set V_{huff}^n of Huffman vectors ($n = 4$)



$$(2, 2, 2, 2)$$

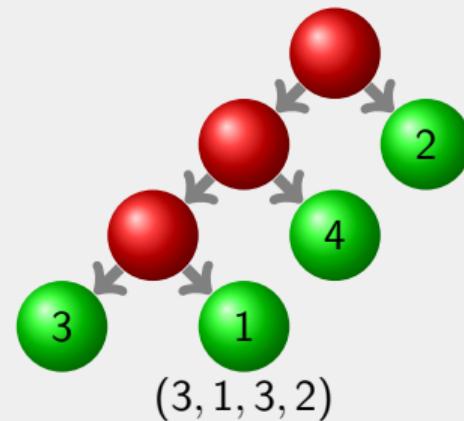
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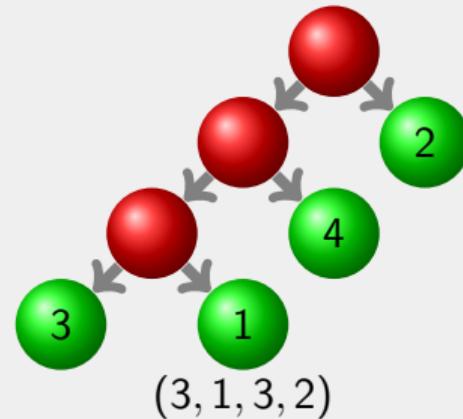
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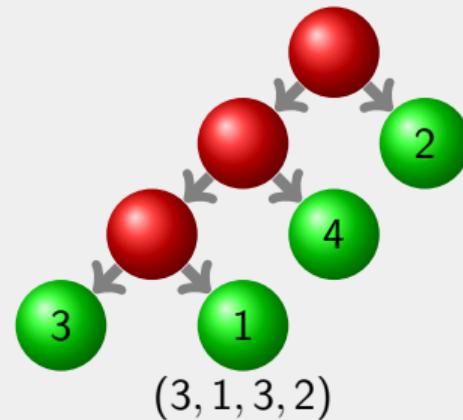
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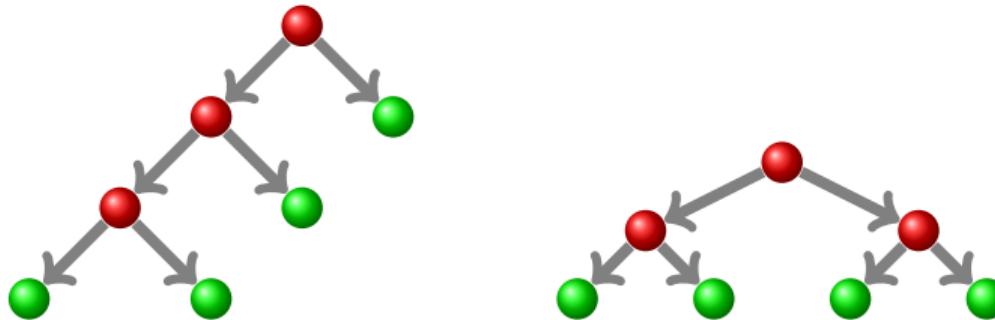


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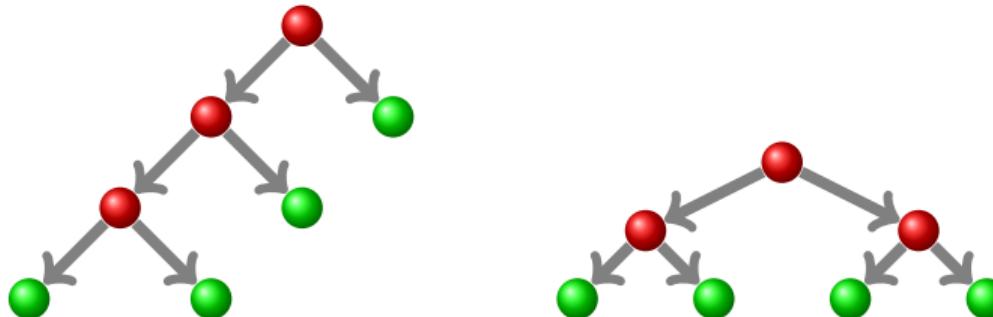
NGUYEN, NGUYEN, & MAURRAS 10

P_{huff}^n has at least $2^{\Omega(n \log n)}$ facets.

An Extended Formulation of Size $O(n^2)$

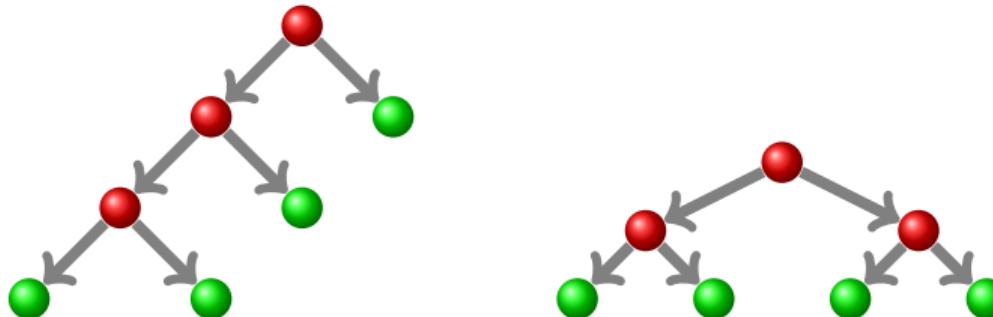


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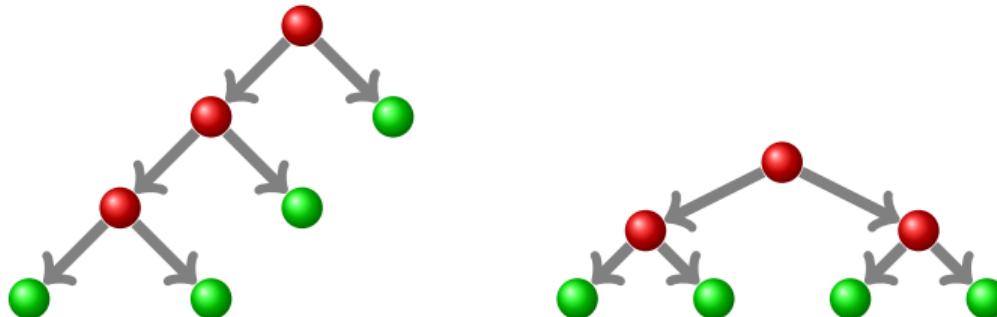
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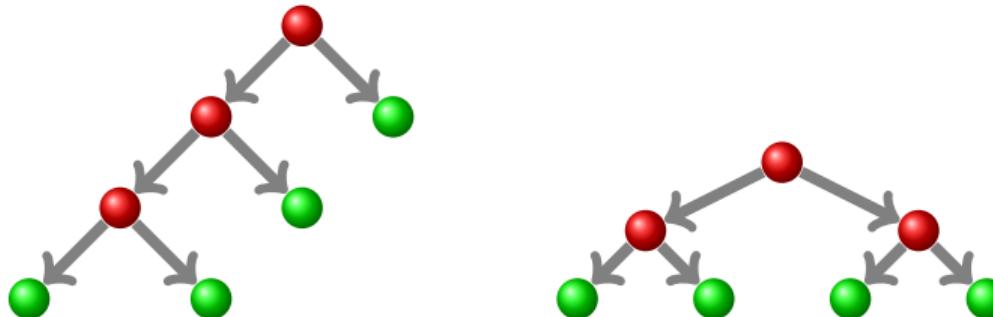
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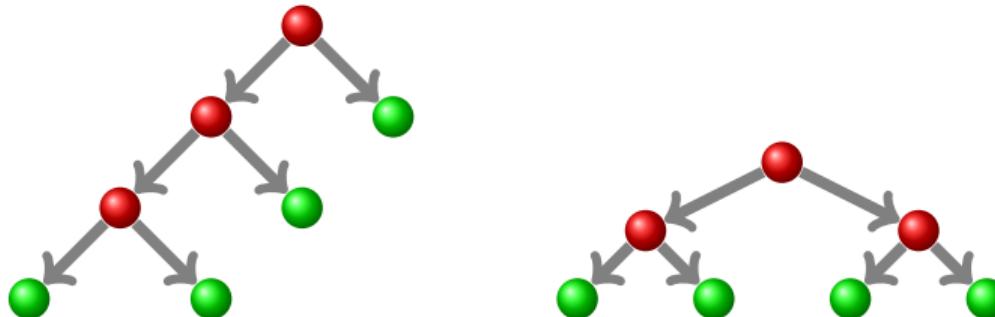
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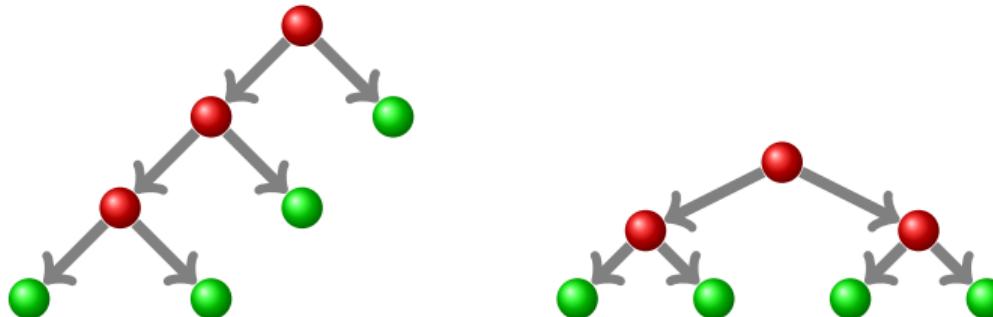
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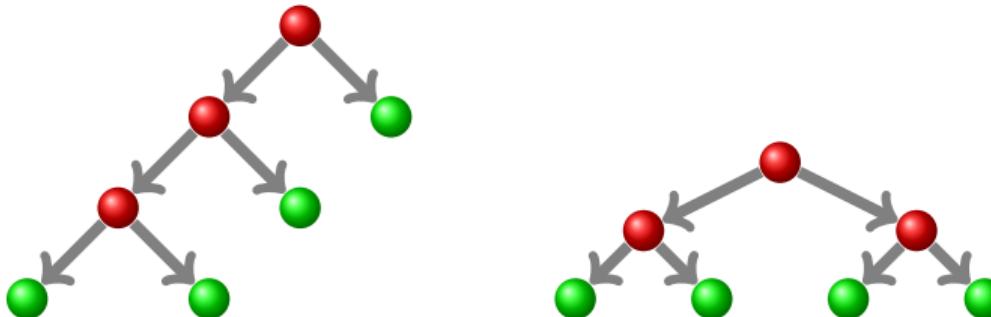
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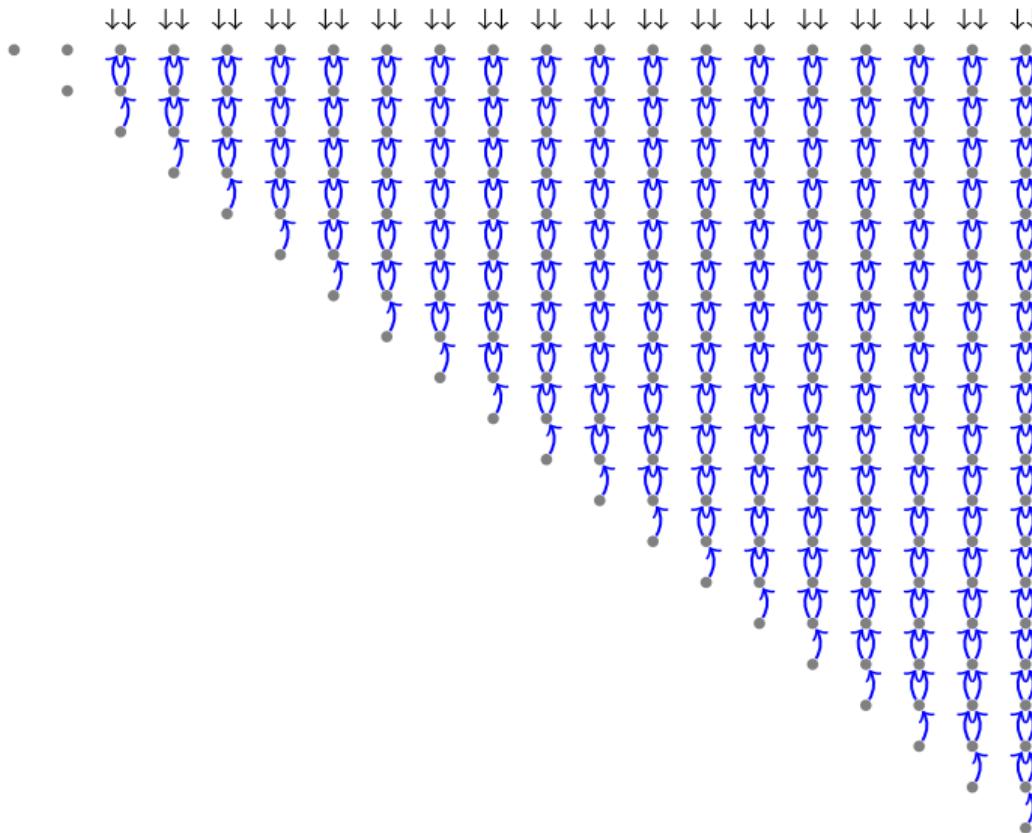


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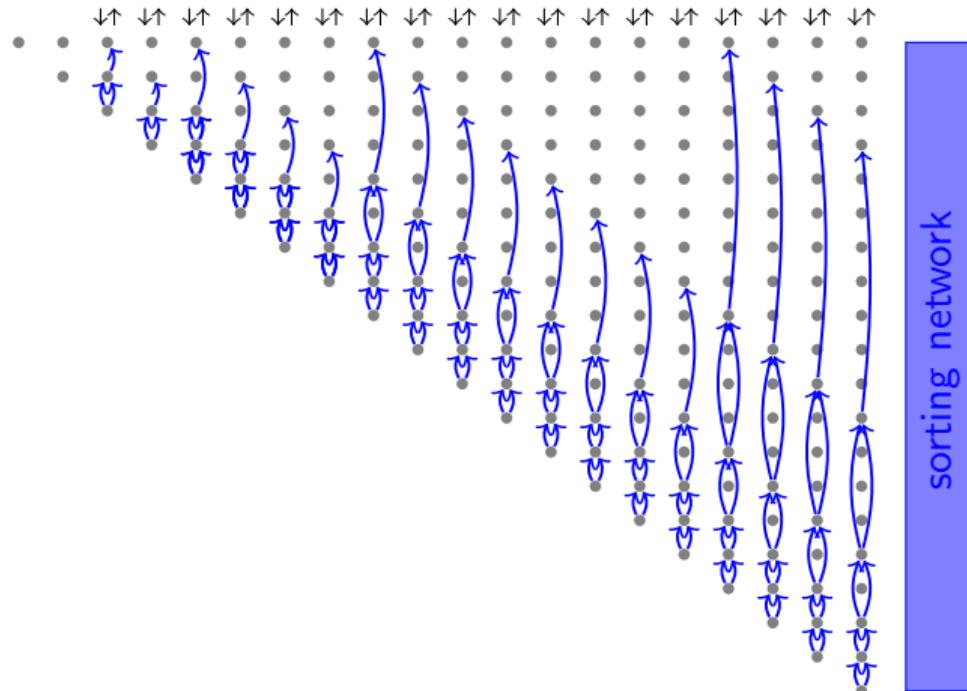
A first extended formulation

$$P_{\text{huff}}^n = \mathcal{R}_{H_{1,2}^{\leq}, \dots, H_{n-1,n}^{\leq}, H_{1,2}^{\leq}, \dots, H_{n-2,n-1}^{\leq}}(\varphi(P_{\text{huff}}^{n-1})).$$

Schematic View on the Construction



A Better Construction



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$$xc(P_{\text{huff}}^n) \leq O(n \log n)$$

Thanks for your attention.