

Convexity Recognition of the Union of Polyhedra

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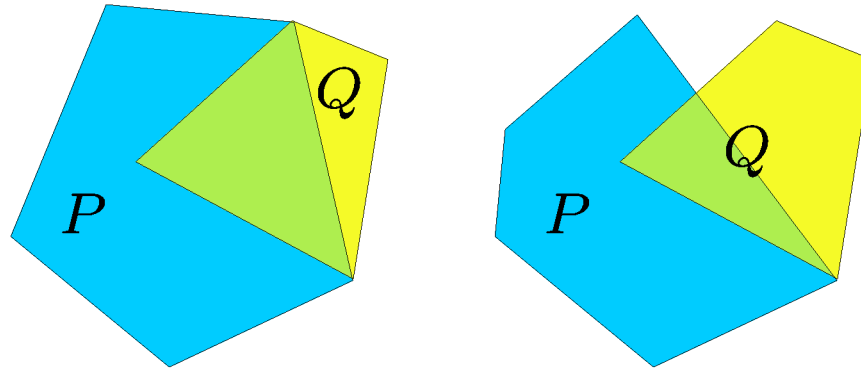
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Joint Work with

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Union of Polyhedra



Problem: Given two convex polyhedra P, Q in \mathbb{R}^d :

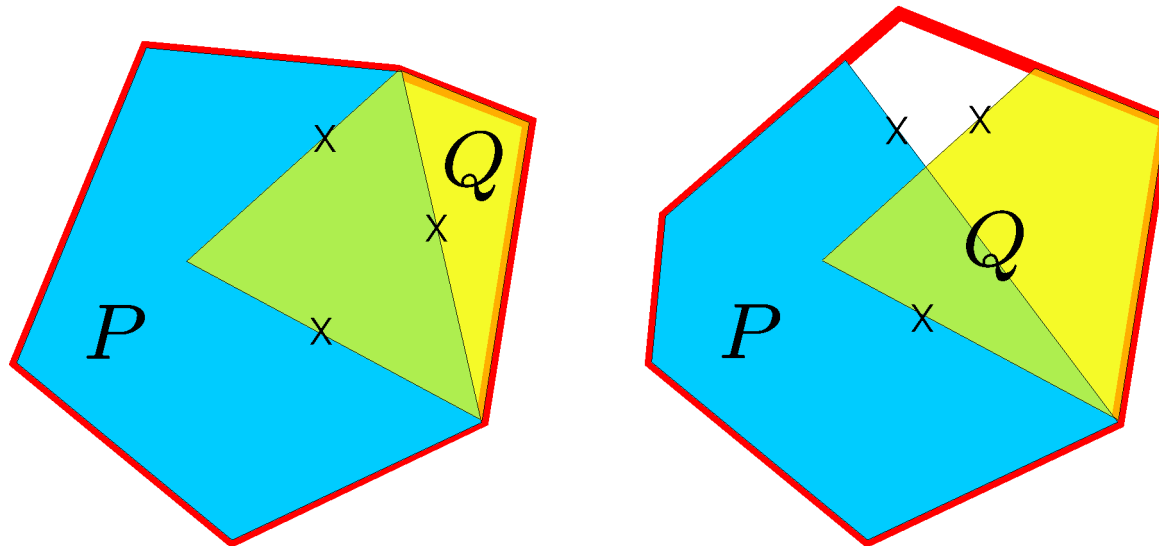
- Recognize if $P \cup Q$ is convex
- If yes, find a minimal representation for $P \cup Q$

Three natural cases:

1. P, Q , are in H-representation $P = \{x : Ax \leq \alpha\}$, $Q = \{x : Bx \leq \beta\}$
2. P, Q , are in V-representation $P = \text{conv}(V) + \text{cone}(R)$,
 $Q = \text{conv}(W) + \text{cone}(S)$
3. P, Q , are in VH-representation

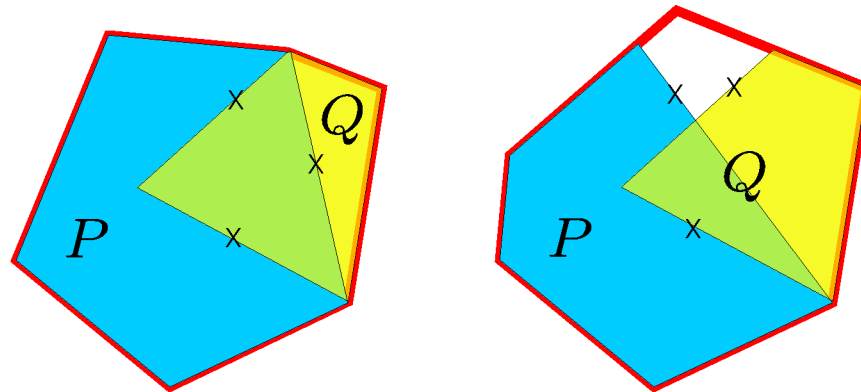
Envelope of Polyhedra

$$P = \{x : Ax \leq \alpha\}, \quad Q = \{x : Bx \leq \beta\}$$



$$\text{env}(P, Q) \triangleq \{x : \bar{A}x \leq \bar{\alpha}, \bar{B}x \leq \bar{\beta}\}$$

Key Theorem for H-Polyhedra



Theorem 1 $P \cup Q$ is convex $\Leftrightarrow P \cup Q = \text{env}(P, Q)$.

Proof.

\Leftarrow trivial (env is a convex object)

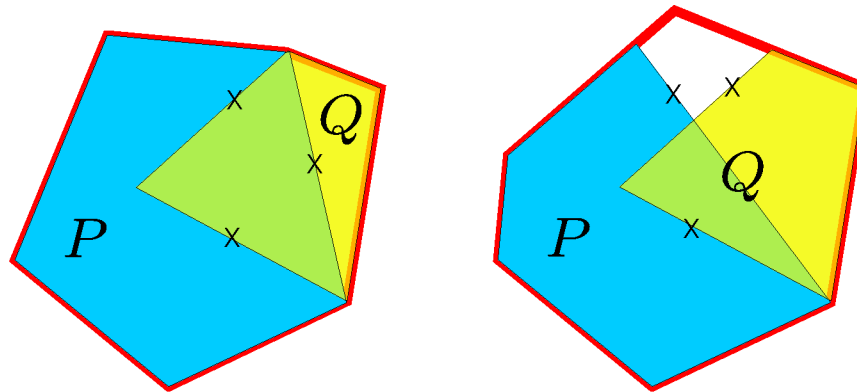
\Rightarrow – Consider the H-representation of $P \cup Q$

– Clearly, $P \cup Q \subseteq \text{env}(P, Q)$

– Show that $P \cup Q \supseteq \text{env}(P, Q)$, by contradiction

- Holds for unbounded polyhedra

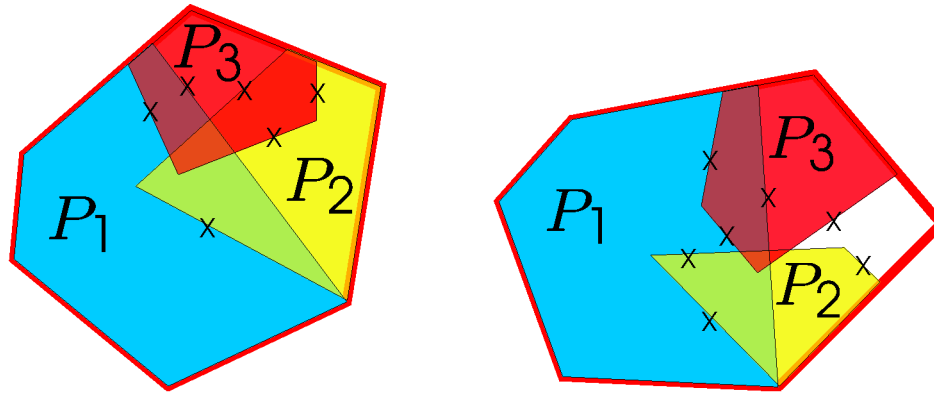
Key Theorem for H-Polyhedra



Proof.

- $P \cup Q = \text{env}(P, Q) \Rightarrow P \cup Q$ is convex trivial
- $P \cup Q$ is convex $\Rightarrow P \cup Q = \text{env}(P, Q)$
 - Let $K \triangleq P \cup Q$, $K \subseteq \text{env}(P, Q)$, show that $K \supseteq \text{env}(P, Q)$
 - assume by contradiction that a facet inequality $r'x \leq s$ of K is not in $\text{env}(P, Q)$
 - Let $H = \{x : r'x = s\}$, then $\dim(P \cap H) \leq d - 2$ and $\dim(Q \cap H) \leq d - 2$, $\dim(K \cap H) = d - 1$
 - $K \cap H = (P \cap H) \cup (Q \cap H)$, contradiction

Generalization to k H-Polyhedra

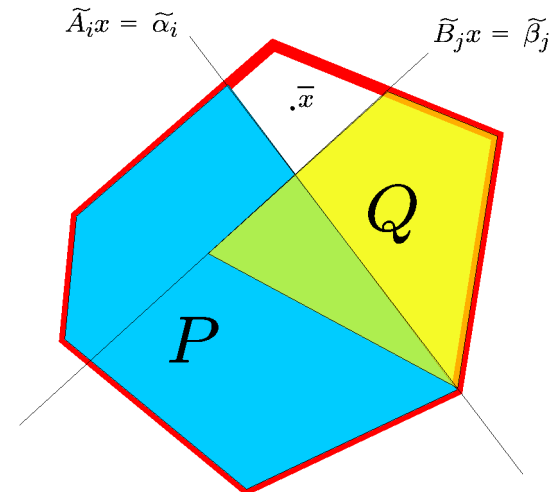


- Theorem 1 generalizes to $k \geq 3$ H-polyhedra

$$\bigcup_{i=1}^k P_i \text{ is convex} \Leftrightarrow \text{env}(P_1, P_2, \dots, P_k) = \bigcup_{i=1}^k P_i$$

Algorithm for H-Polyhedra

Main idea: Determine a point $\bar{x} \in \text{env}(P, Q)$, $\bar{x} \notin P$, $\bar{x} \notin Q$

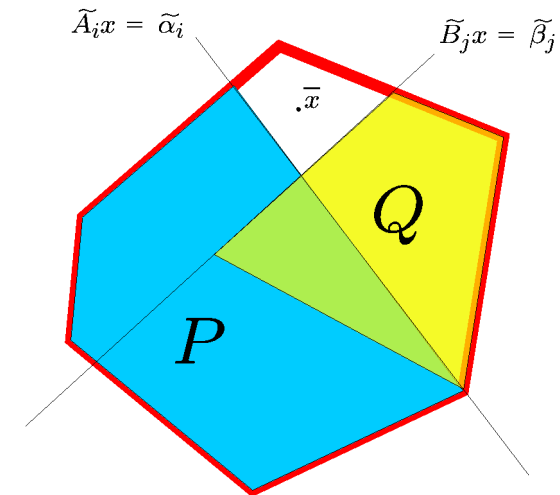


- 1 Construct $\text{env}(P, Q) = \{x : Cx \leq \gamma\}$
let $\tilde{A}x \leq \tilde{\alpha}$, $\tilde{B}x \leq \tilde{\beta}$ be the set of removed constraints
- 2 Remove from $\text{env}(P, Q)$ possible duplicates
- 3 **for** each pair $\tilde{A}_i x \leq \tilde{\alpha}_i$, $\tilde{B}_j x \leq \tilde{\beta}_j$ **do**
- 4 Determine ϵ^* by solving the linear program

$$\begin{aligned} \epsilon^* = \max_{(x, \epsilon)} \quad & \epsilon \\ \text{subj. to} \quad & \tilde{A}_i x = \tilde{\alpha}_i + \epsilon \\ & \tilde{B}_j x = \tilde{\beta}_j + \epsilon \\ & Cx \leq \gamma \end{aligned}$$

- 5 **if** $\epsilon^* > 0$, **stop**; **return False**; /* $P \cup Q$ is nonconvex.*/
- 6 **return** $\text{env}(P, Q)$. /* $P \cup Q$ is convex.*/

Algorithm for H-Polyhedra



$$P = \{Ax \leq \alpha\}, Q = \{Bx \leq \beta\}$$

Input: $(A, \alpha), (B, \beta)$ minimal H-representation

Output: Minimal H-representation of $P \cup Q$

Complexity: $O(m_1 m_2 \mathbf{lp}(d, m_1 + m_2))$

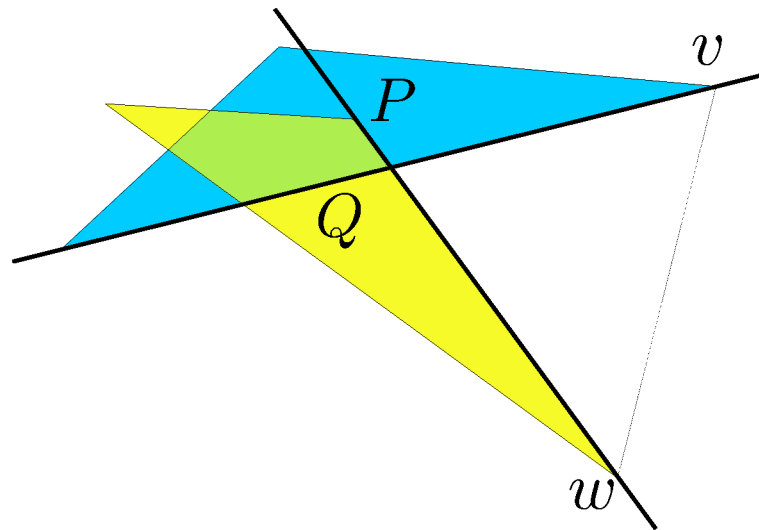
m_1 = number of rows of A

m_2 = number of rows of B

In general: $O(\prod_{i=1}^k m_i \mathbf{lp}(d, \sum_{i=1}^k m_i))$

Key Theorem for V-Polytopes

$$P = \text{conv}(V), \quad Q = \text{conv}(W)$$



Theorem 2 Let P, Q be polytopes with V -representation V and W , respectively. Then

$$P \cup Q \text{ is convex} \Leftrightarrow [v, w] \subseteq P \cup Q, \quad \forall v \in V, \quad \forall w \in W.$$

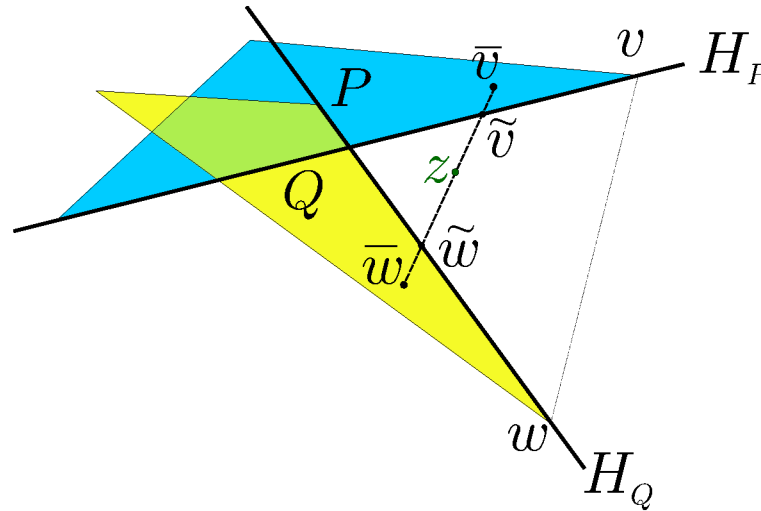
Moreover, a stronger characterization of convexity holds

$$\exists v \in V, \quad w \in W: (v, w) \cap (P \cup Q) = \emptyset \Leftrightarrow P \cup Q \text{ is nonconvex.}$$

Generalizes to unbounded polyhedra by homogenization

Key Theorem for V-Polytopes

$$P = \text{conv}(V), \quad Q = \text{conv}(W)$$



Proof.

$\Rightarrow P \cup Q$ is convex $\Rightarrow [v, w] \subseteq P \cup Q, \forall v \in V, \forall w \in W$, trivial

$\Leftarrow [v, w] \subseteq P \cup Q, \forall v \in V, \forall w \in W \Rightarrow P \cup Q$ is convex

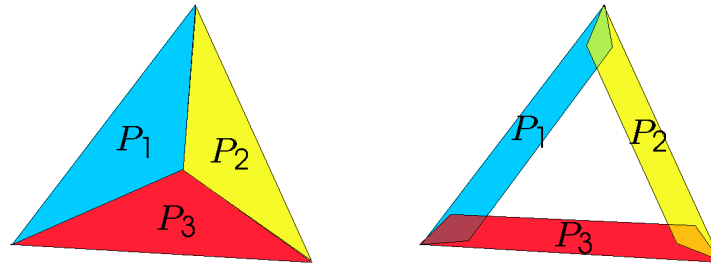
by contradiction assume $\exists z = (1 - \gamma)\bar{v} + \gamma\bar{w}, 0 \leq \gamma \leq 1, \bar{v} \in P, \bar{w} \in Q$ s.t. $z \notin P \cup Q$

$\Rightarrow \exists \tilde{v}, \tilde{w}$ s.t. $(\tilde{v}, \tilde{w}) \subset [\bar{v}, \bar{w}], (\tilde{v}, \tilde{w}) \not\subseteq P \cup Q, \tilde{v} \in H_P, \bar{w} \in H_P^-$

$\Rightarrow \exists w \in W, w \in H_P^-,$ similarly $\exists v \in V, v \in H_Q^-$

$\Rightarrow (v, w) \cap P \cup Q = \emptyset$, contradiction

Generalization to k V-Polyhedra



Open problem:

- Generalization to $k \geq 3$ V-polyhedra

Conjecture (false):

- Θ union of the vertices of P_1, P_2, P_3
- Check $\forall \theta_i, \theta_j \in \Theta, [\theta_i, \theta_j] \subset P_1 \cup P_2 \cup P_3$

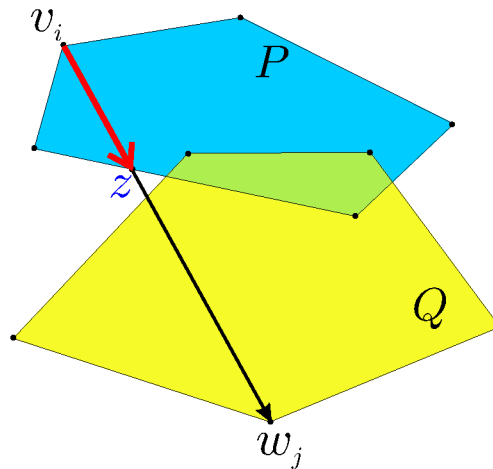
Alternative: (Carathodory) Consider the convex hull of $(d + 1)$ vertices and check if it is contained in $P_1 \cup P_2 \cup P_3$

Theorem: The convex hull of k vertices is enough

Proof. By Carateodory's Theorem (Finschi, Torrisi)

Can we exploit this characterization in an algorithm?

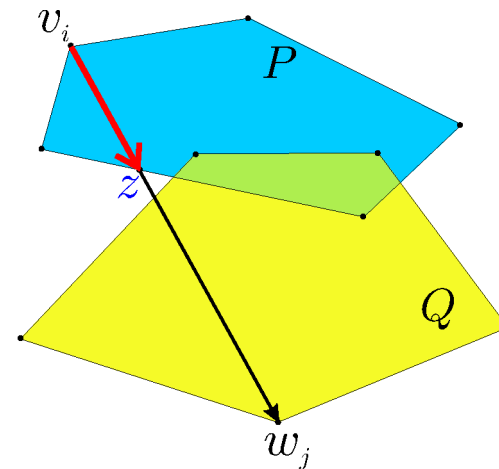
Algorithm for V-Polyhedra (1)



Main idea: Ray shooting from v_i towards w_j , and check $z \in Q$

- 1 Remove vertices of P which are in Q , and vice-versa, and
let \bar{V} , \bar{W} the sets of remaining vertices;
- 2 **if** no vertex has been removed, **return False**; /*disjoint.*/
- 3 **for** each pair $v_i \in \bar{V}$, $w_j \in \bar{W}$ **do**
- 4 Find the corresponding vector $z \triangleq v_i + \lambda_0^*(w_j - v_i)$,
where $\lambda_0^* = \max \lambda_0$ s.t. $v_i + \lambda_0(w_j - v_i) \in P$
- 5 Determine if $z \in Q$ (via LFT)
- 6 **if** $z \notin Q$, **return False**; /* $P \cup Q$ is nonconvex.*/
- 7 **let** X be the set of points in $V \cap W$ that are extreme in $P \cup Q$;
- 8 **return** $\text{conv}(\bar{V} \cup \bar{W} \cup X)$.

Algorithm for V-Polyhedra (1)



$$P = \text{conv}(V), \quad V = \{v_1, \dots, v_{n_1}\}$$

$$Q = \text{conv}(W), \quad W = \{w_1, \dots, w_{n_2}\}$$

Input: V, W minimal V-representation

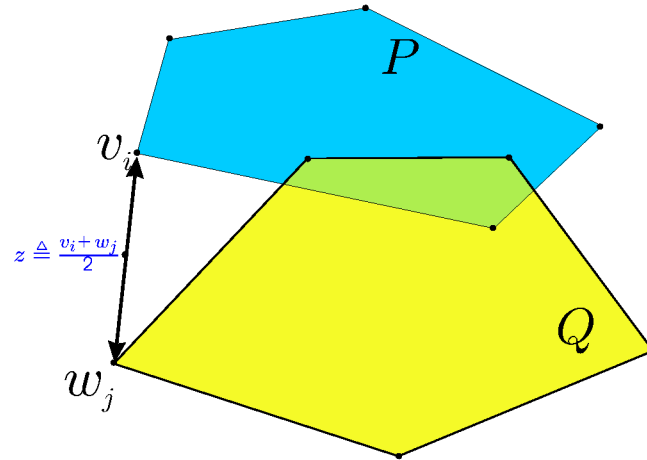
Output: Minimal V-representation of $P \cup Q$

Complexity: $O(n_1 n_2 (\mathbf{lp}(d, n_1) + \mathbf{lp}(d, n_2)))$

n_1 = number of vertices of P

n_2 = number of vertices of Q

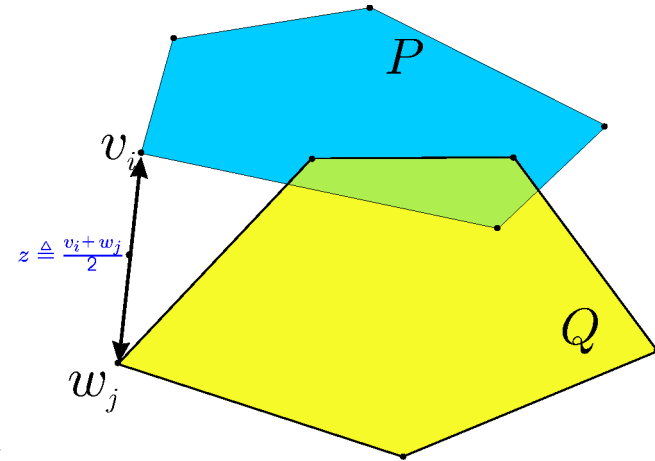
Algorithm for V-Polyhedra (2)



Main idea: Exploit the stronger converse result in Theorem 2, and try to find a segment $(v_i, w_j) \notin P \cup Q$ by checking the middle point $z = \frac{v_i + w_j}{2}$

- 1 Remove vertices of P which are in Q , and vice-versa, and
let \bar{V} , \bar{W} the sets of remaining vertices;
- 2 **if** no vertex has been removed, **return False**; /*disjoint*/
- 3 **for** each pair $v_i \in \bar{V}$, $w_i \in \bar{W}$ **do**
- 4 **let** $z \triangleq \frac{v_i + w_j}{2}$;
- 5 Determine if $z \in P \cup Q$ (via LFT)
if $z \notin P \cup Q$, **return False**; /* $P \cup Q$ is not convex.*/
- 7 **let** X be the set of points in $V \cap W$ that are extreme in $P \cup Q$;
- 8 **return** $\text{conv}(\bar{V} \cup \bar{W} \cup X)$.

Algorithm for V-Polyhedra (2)



$$P = \text{conv}(V), \quad V = \{v_1, \dots, v_{n_1}\}$$

$$Q = \text{conv}(W), \quad W = \{w_1, \dots, w_{n_2}\}$$

Input: V, W minimal V-representation

Output: Minimal V-representation of $P \cup Q$

Complexity: $O(n_1 n_2 (\text{lp}(d, n_1) + \text{lp}(d, n_2)))$

n_1 = number of vertices of P

n_2 = number of vertices of Q

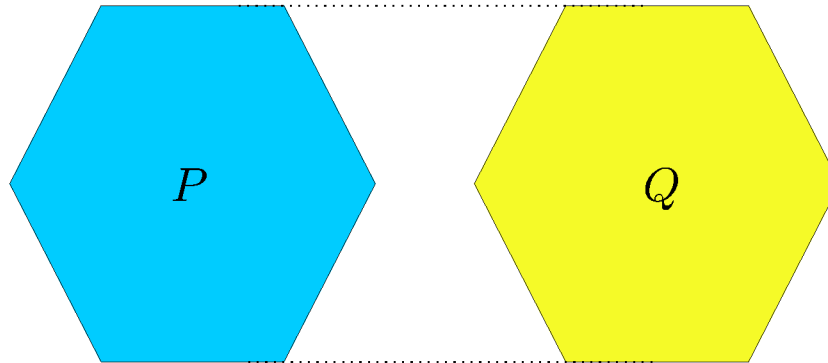
Comparison:

- Algorithm 1 might stop earlier if $P \cup Q$ is not convex
→ performance depends on inputs P and Q

VH-Polytopes

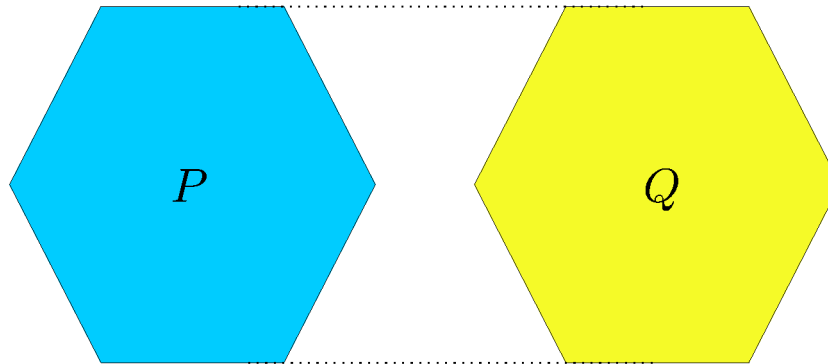
Theorem 3 *Let P and Q be VH-polytopes, $P \cup Q$ is convex $\Rightarrow \text{conv}(V \cup W) = \text{env}(P, Q)$, moreover $\text{conv}(V \cup W) = \text{env}(P, Q) = P \cup Q$.*

The converse is not true:



Not useful for convexity recognition, a converse result can be proved under additional assumptions

VH-Polytopes



Theorem 4 Let $\bar{V} = \{v \in V : v \notin Q\}$, $\bar{W} = \{w \in W : w \notin P\}$. If $\bar{V} \cup \bar{W} \cup (V \cap W)$ and $\text{env}(P, Q)$ are *minimal* V- and H-representations, respectively, of the *same polytope* then $P \cup Q$ is convex, and $P \cup Q = \text{conv}(\bar{V} \cup \bar{W} \cup (V \cap W)) = \text{env}(P, Q)$.
Proof. by contradiction.

The result can not be exploited for convexity recognition of the union of polyhedra, as checking coherence of given V- and H-representation might be a hard task

Algorithm for VH-Polyhedra

Assuming that P and Q are given by **coherent VH-representations**, an efficient algorithm can be proposed by modifying 2nd Algorithm for V-polytopes to exploit information coming from H-representations

In this case the algorithm computes the solution without solving any LP

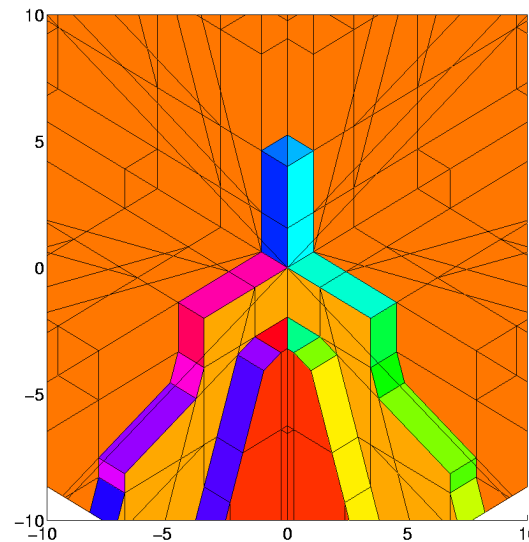
Complexity: $O(n_1 n_2 d(m_1 + m_2))$ **strongly polynomial**

Related Work (1): Reduction

Multiparametric Programming amounts to solve for all x

$$\begin{aligned} \min_z \{ & Rz + Qx \}, \\ \text{subj. to } & Gz \leq W + Kx \end{aligned}$$

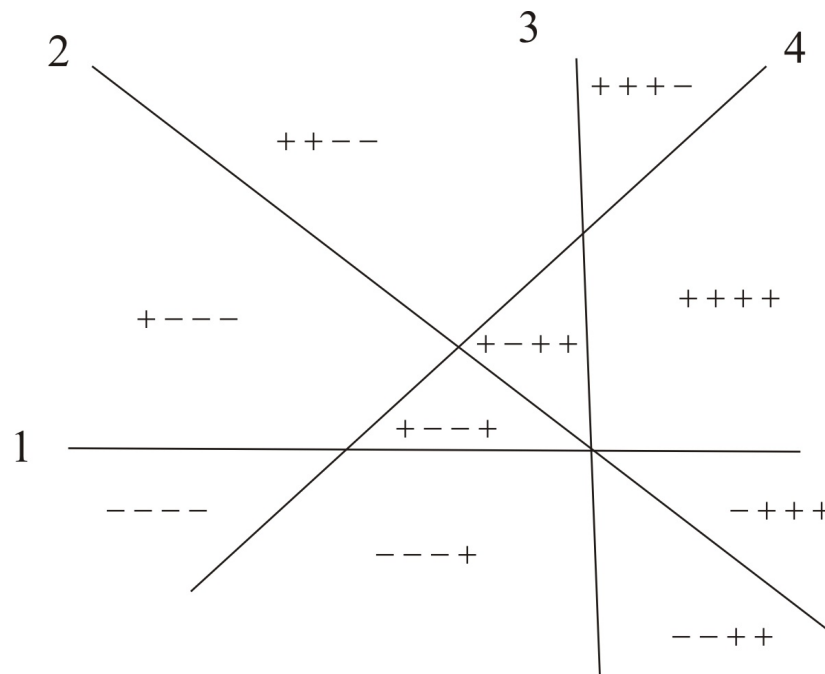
Optimal solution looks like



Problem: Find a “minimal” representation for the solution

Hyperplane Arrangements

Buck 1943,
Edelsbrunner 1987,
Fukuda 1996



Let $\mathcal{A} = \{H_i\}_{i=\{1, \dots, n\}}$, $H_i = \{x: a_i x - b_i = 0\}$ be a collection of n hyperplanes in \mathbb{R}^d

Theorem Each polyhedral region (or cell) is associated to a sign marking

Theorem The total number of cells is bounded by Buck's formula

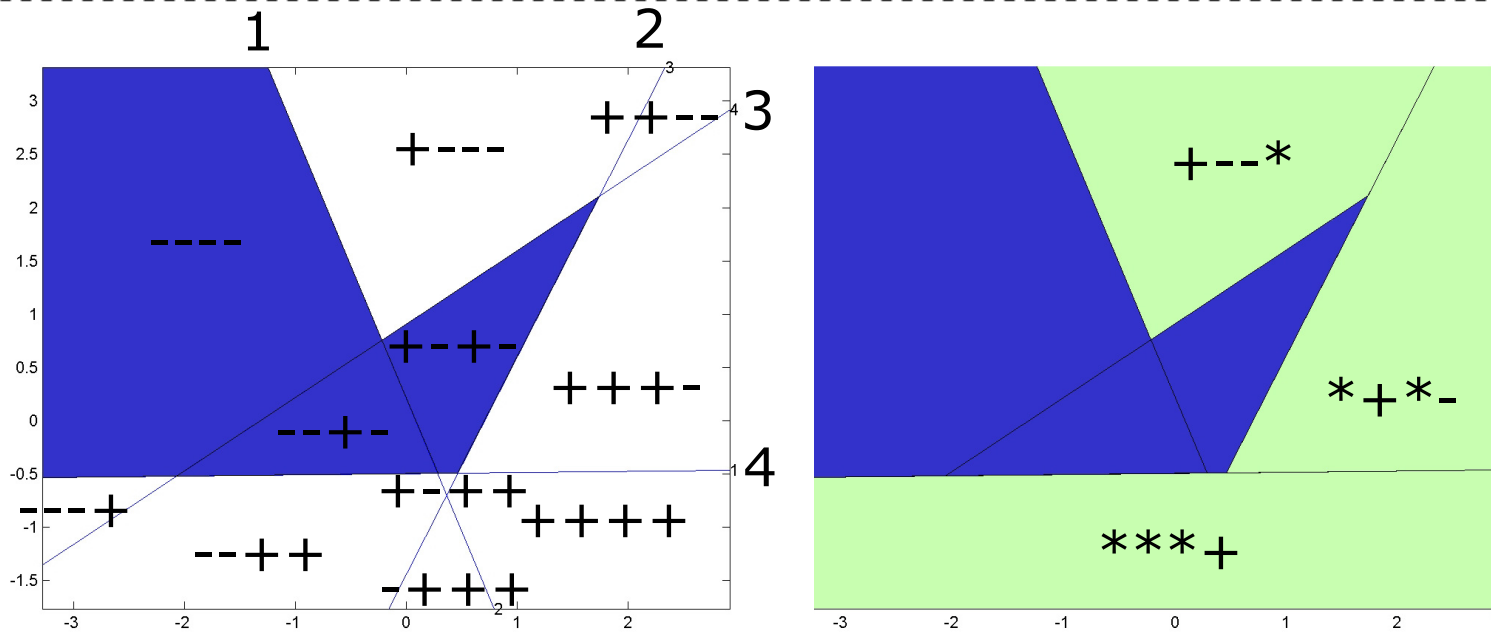
$$\#M \leq \sum_{i=0}^d \binom{n}{i}$$

Hyperplane Arrangements - Algorithms

There is an **optimal** algorithm for enumeration of hyperplane arrangements with time and space complexity $O(n^d)$ (Edelsbrunner '87)

There is **reverse search** algorithm (Fukuda '96,'01) for enumeration of hyperplane arrangements that runs in $O(n \text{lp}(n,d) \#M)$ time and $O(n,d)$ space, where $\text{lp}(n,d)$ is the complexity of solving a linear program with d variables and n constraints

Optimal Merging Of Cells - Idea



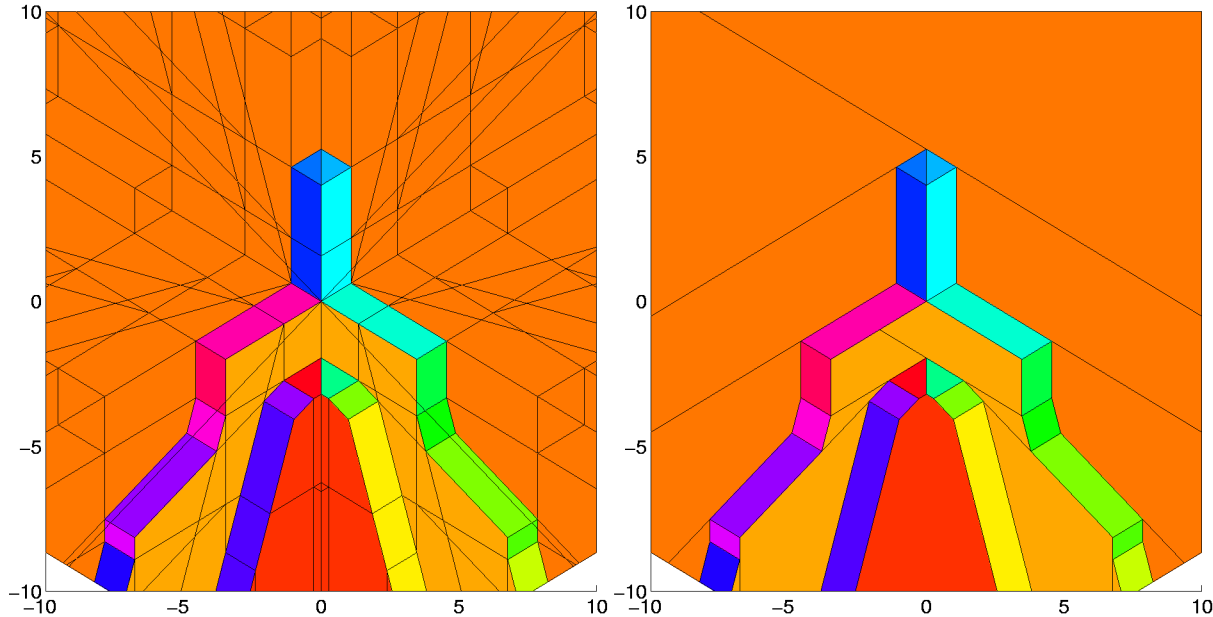
Markings allows **easy** merging of the region

- o Convexity recognition by marking comparison
- o Redundancy removal by marking comparison
- o Branch&Bound guarantees minimum by trying several combinations
- o **Future research**: Fast suboptimal algorithms

Optimal Merging

Geyer Torrisi '03

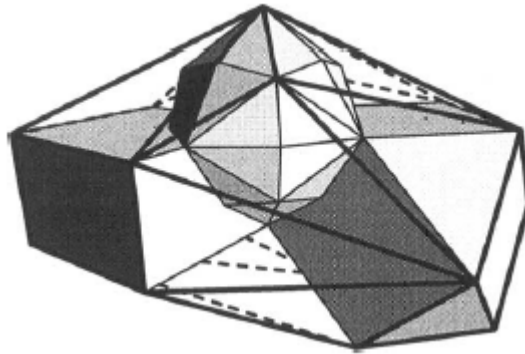
252 regions are reduced to 39 regions in 2' on Pentium IV
2.8 GHz machine



Related Work (2): Extended hull

Fukuda et al. 01

Compute the convex hull of the union of k H-polytopes



Conclusions

- We have provided:
 - Key theorems for characterizing the union of H-, V- and VH-polyhedra
 - Algorithms for computing the union of H-, V- and VH-polyhedra
- Similar work:
 - Efficient algorithms for adjacent H-polyhedra. (useful for multiparametric programming) (Geyer and Torrisi '03)
 - Convex hull of k H-polyhedra (Fukuda, Liebling and Lütolf '01)
- Open problems:
 - Generalization to k V-polyhedra

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