

Algorithms for high-dimensional polytopes defined by oracles

Vissarion Fisikopoulos

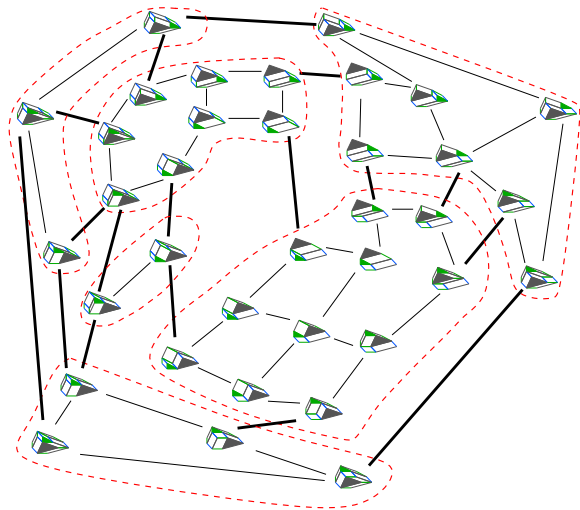
Joint work with I.Z. Emiris (UoA), B. Gärtner (ETHZ)

Dept. of Informatics & Telecommunications, University of Athens

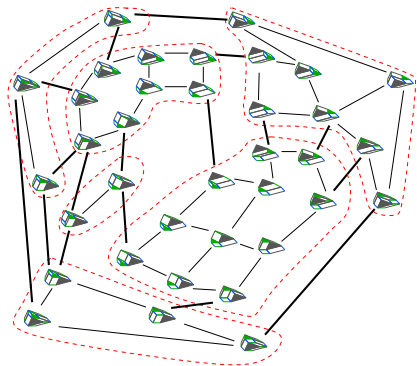


Advanced Geometric Computing and Critical Applications
Kickoff Meeting, Athens, 22.Feb.2013

Motivation: Secondary & Resultant polytopes

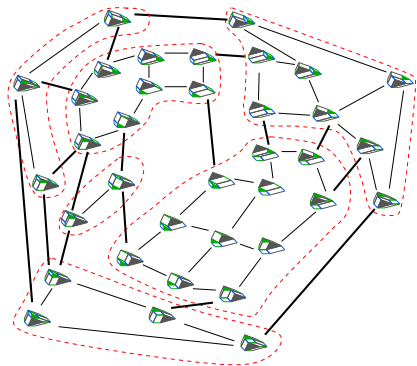


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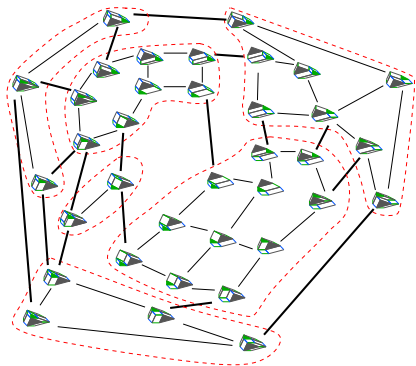
- **Algorithm:** [EFKP SoCG'12]
vertex oracle + incremental construction = output-sensitive

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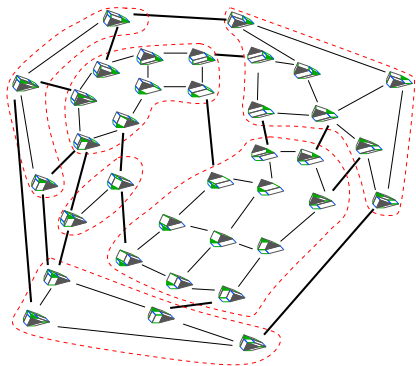
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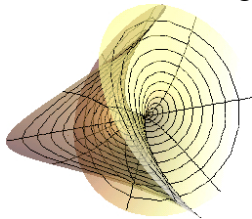
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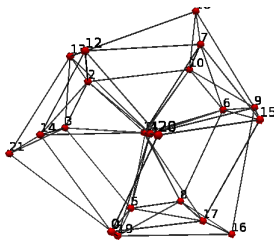
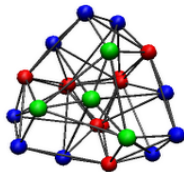
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- ▶ **Q:** More polytopes given by optimization oracles ?

Applications

- ▶ Geometric Modeling (Implicitization) [EKKL'12]



- ▶ Combinatorics of 4-d resultant polytopes (with Emiris & Dickenstein)



(figure courtesy of M.Joswig)

Facet and vertex graph of the largest 4-dimensional resultant polytope

How 4-d resultant polytopes look like?

(6, 15, 18, 9)

(8, 20, 21, 9)

(9, 22, 21, 8)

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(17, 48, 45, 14)

(17, 48, 46, 15)

(17, 48, 47, 16)

(17, 49, 47, 15)

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(17, 49, 49, 17)

(17, 50, 50, 17)

(18, 51, 48, 15)

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(19, 56, 54, 17)

(19, 56, 56, 19)

(19, 57, 57, 19)

(20, 58, 54, 16)

(20, 59, 57, 18)

(20, 60, 60, 20)

(21, 62, 60, 19)

(21, 63, 63, 21)

(22, 66, 66, 22)

Open problem

Almost symmetric f-vector?

Outline

Polytope Representation & Oracles

Edge Skeleton Computation

Geometric Random Walks: Optimization & Volume computation

Experimental Results

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Polytope representation

Convex polytope $P \in \mathbb{R}^n$.

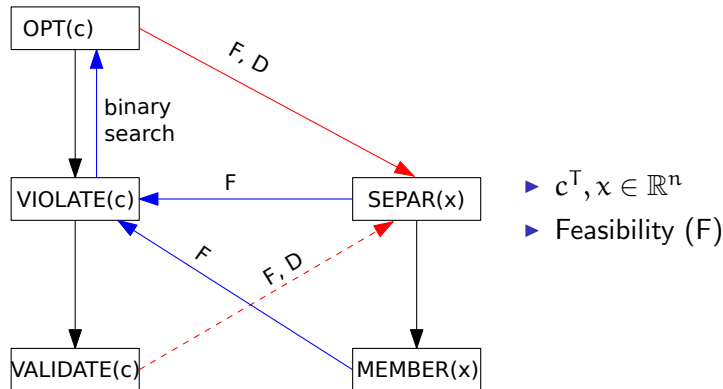
Explicit: Vertex-, Halfspace - representation (V_P, H_P),
Edge-skeleton (ES_P), Triangulation (T_P), Face lattice

Implicit: Oracles (OPT_P, SEP_P, MEM_P)

Motivation-Applications

- ▶ Resultant, Discriminant, Secondary polytopes
- ▶ (Generalized) Minkowski sums

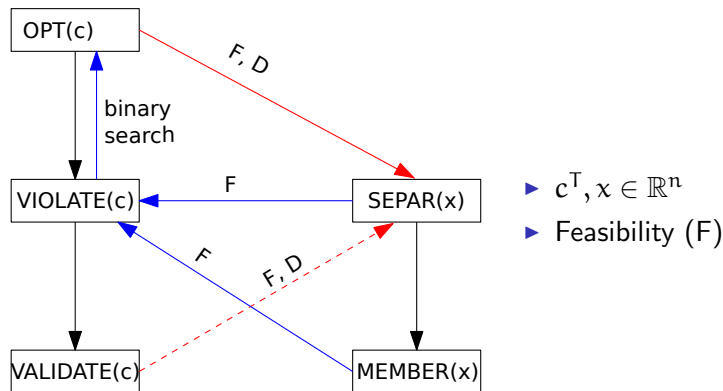
Oracles and duality [Grötschel et al.'93]



(Polar) Duality (D):

$$\mathbf{0} \in \text{int}(P), \quad P^* := \{c \in \mathbb{R}^n : c^T x \leq 1, \text{ for all } x \in P\} \subseteq (\mathbb{R}^n)^*$$

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Given OPTIMIZATION compute SEPARATION.

Polytope change of representation

Problem	Algorithm	Complexity
$V_P \rightarrow H_P$	Convex hull	EXP
Feasibility	Ellipsoid [Kha'79], Las Vegas [BV'04]	P_{bit} , ZPP
$\text{OPT}_P + \{\text{edge dir.}\} \rightarrow \text{ES}_P$	Incremental [EFG'12]	$P_{\text{bit}}(\text{in}, \text{out})$
$\text{MEM}_P \rightarrow \epsilon\text{-approx vol}(P)$	Monte-Carlo [Dyer et.al'91, LV'04]	BPP

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Our contribution: Theory & Implementation

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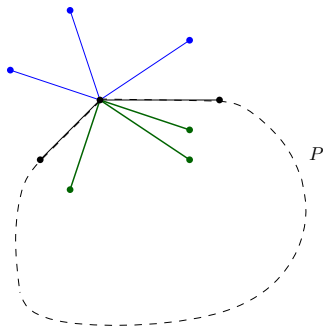
Edge skeleton computation

Input:

- ▶ OPT_P
- ▶ Edge directions of P : D

Output:

- ▶ Edge-skeleton of P



Sketch of **Algorithm**:

- ▶ Compute a vertex of P ($x = \text{OPT}_P(c)$ for arbitrary $c^T \in \mathbb{R}^n$)
- ▶ Compute segments $S = \{(x, x + d), \text{ for all } d \in D\}$
- ▶ Remove from S all segments (x, y) s.t. $y \notin P$ ($\text{OPT}_P \rightarrow \text{SEP}_P$)
- ▶ Remove from S the segments that are not extreme

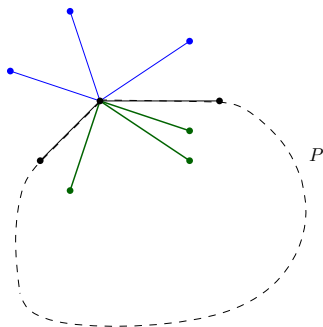
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Open problem: Do not use $\text{OPT}_P \rightarrow \text{SEP}_P$.

Edge skeleton computation

Proposition

[RothblumOnn07] Let $P \subseteq \mathbb{R}^n$ given by OPT_P , and $E \supseteq D(P)$. All vertices of P can be computed in

$O(|E|^{n-1})$ calls to $\text{OPT}_P + O(|E|^{n-1})$ arithmetic operations.

Theorem

The edge skeleton of P can be computed in

$O^*(m^3 n)$ calls to $\text{OPT}_P + O^*(m^3 n^{3.38} + m^4 n)$ arithmetic operations,

m : the number of vertices of P .

Corollary

For resultant polytopes $R \subset \mathbb{Z}^n$ this becomes (d is a constant)

$$O^*(m^3 n^{\lfloor (d/2)+1 \rfloor} + m^4 n).$$

Outline

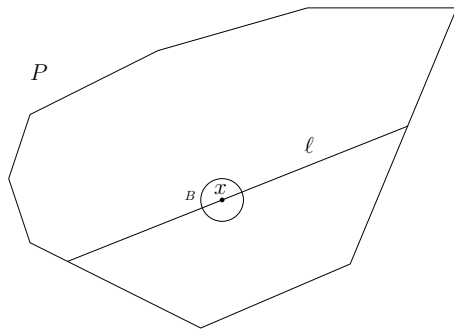
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Random points in polytopes with SEP_P

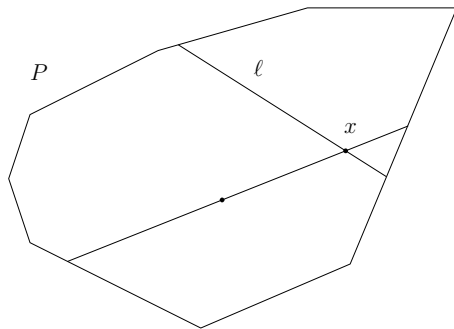


Hit-and-Run walk

- ▶ line ℓ through x , uniform on $B_x(1)$
- ▶ move x to a uniform distributed point on $P \cap \ell$

x will become “random” after $O(n^3)$ hit-and-run steps [Lovász98]

Random points in polytopes with SEP_P

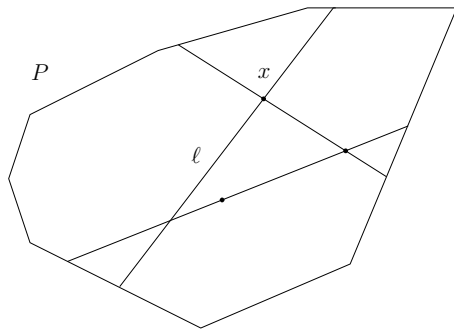


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Random points in polytopes with OPT_P

1. Hit-and-Run walk
with $\text{OPT} \rightarrow \text{SEP}$ in every step
2. Vertex walk
 - ▶ for uniform c compute $\text{OPT}_P(c)$
 - ▶ segment ℓ , connect x , $\text{OPT}_P(c)$
 - ▶ move x to a uniform distributed point on ℓ

Open problem: Analyse Vertex walk (or a similar walk).

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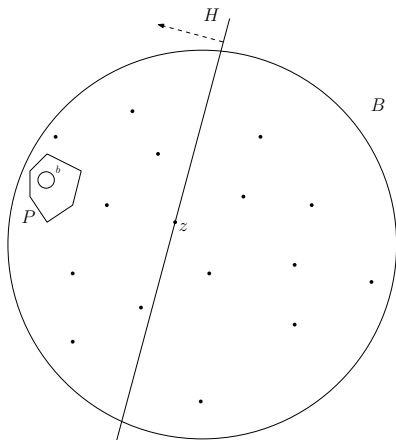
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Optimization using random walks [BV'04]

Optimization reduces to Feasibility:

Input: SEP_P , B , $L = \lg \frac{\text{radius}(B)}{\text{radius}(b)}$

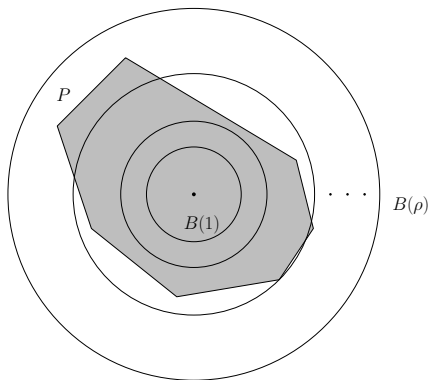
Output: $z \in P \subseteq \mathbb{R}^n$ or P is empty



1. Compute N random points y_1, \dots, y_N uniform in B ;
2. Let $z \leftarrow \frac{1}{N} \sum_{i=1}^N y_i$; $H \leftarrow \text{SEP}_P(z)$;
3. If $z \in P$ return z , else $B \leftarrow B \cap H$;
4. Repeat steps 1-3, $2nL$ times;
Report P is empty;

Complexity: $O^*(n)$ oracle calls + $O^*(n^7)$ arithm. oper.

Volume computation using random walks [Dyer et.al'91]



Input: MEM_P, ρ :

$$B(1) \subseteq P \subseteq B(\rho) \subseteq \mathbb{R}^n$$

Output: ϵ -approximation $\text{vol}(P)$

1. $P_i := P \cap B(2^{i/n})$, $i = 0 : \lceil n \lg \rho \rceil$;
 $P_0 = B(1)$, $P_{n \lg \rho} = P$
2. Generate rand. point in P_0
3. Generate rand. points in P_i and count how many fall in P_{i-1}

$$\text{vol}(P) = \text{vol}(P_0) \prod_{i=1}^m \frac{\text{vol}(P_i)}{\text{vol}(P_{i-1})}$$

Complexity [Lovász et al.'04]: $O^*(n^4)$ oracle calls

Volume of polytopes given by OPT_P

Input: OPT_P , ρ : $B(1) \subseteq P \subseteq B(\rho)$

Output: ϵ -approximation $\text{vol}(P)$

- ▶ Call volume algorithm
- ▶ Each MEM_P oracle calls feasibility/optimization algorithm

Corollary

An approximation of the volume of resultant and Minkowski sum polytopes given by OPT oracles can be computed in $O^(n^{\lfloor (d/2)+5 \rfloor})$ and $O^*(n^{7.38})$ respectively, where d is a constant.*

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Experiments Optimization

- ▶ n-cubes (table), n-crosspolytopes, skinny crosspolytopes
- ▶ M: multipoint walk, H: Hit-and-Run walk

n	# rand. points	# walk steps	Alg. O1		Alg. O2		Alg. O3	
			M(sec)	H(sec)	M(sec)	H(sec)	M(sec)	H(sec)
2	4	0	0.02	0.05	0.01	0.01	0.01	0.006
4	38	0	0.59	1.53	0.10	0.08	0.10	0.119
6	96	1	5.54	13.23	0.47	0.84	0.99	0.727
8	172	4	61.40	73.94	4.33	5.34	9.82	4.527
10	265	10	306.20	357.88	26.64	17.22	74.86	16.44
11	316	14	559.97	853.04	54.71	36.95	112.57	55.60

- ▶ Efficient computation ($< 1\text{min}$) up to dimension 11 using Hit-and-Run

Experiments Volume given Membership oracle

- n-cubes (table), n-crosspolytopes, σ =average absolute deviation, μ =average over 20 experiments

n	exact vol	exact sec	# rand. points	# walk steps	vol min	vol max	vol μ	vol σ	approx sec
2	4	0.06	2218	8	3.84	4.12	3.97	0.05	0.23
4	16	0.06	2738	7	14.99	16.25	15.59	0.32	1.77
6	64	0.09	5308	38	60.85	67.17	64.31	1.12	39.66
8	256	2.62	8215	16	242.08	262.95	252.71	5.09	46.83
10	1024	388.25	11370	40	964.58	1068.22	1019.02	30.72	228.58
12	4096	–	14725	82	3820.94	4247.96	4034.39	80.08	863.72

- (the only known) implementation of [Lovász et al.'12] tested only for cubes up to $n = 8$
- volume up to dimension 12 within mins with $< 2\%$ error
- no hope for exact methods in much higher than 10 dim
- the minimum and maximum values bounds the exact volume

Experiments Volume of Minkowski sum

- ▶ Mink. sum of n -cube and n -crosspolytope, σ =average absolute deviation, μ =average over 10 experiments

n	exact vol	exact sec	# rand. points	# walk steps	vol min	vol max	vol μ	vol σ	approx sec
2	14.00	0.01	216	11	12.60	19.16	15.16	1.34	119.00
3	45.33	0.01	200	7	42.92	57.87	49.13	3.92	462.65
4	139.33	0.03	100	7	100.78	203.64	130.79	21.57	721.42
5	412.26	0.23	100	7	194.17	488.14	304.80	59.66	1707.97

- ▶ slower than volume with MEM
- ▶ improvements in optimization and volume implementation improve also this

Future work - Open problems

1. describe an *efficient* random walk procedure for P given by OPT instead of MEM
2. P of special case (e.g. Minkowski sum, resultant, secondary polytope)
3. volume computation in the polar dual and *Mahler volume*
4. describe *all* edge directions of a resultant polytope

Last slide !

The code

- ▶ <http://sourceforge.net/projects/randgeom>

E Y X A P I Σ T □