

Efficient Random-Walk Methods for Approximating Polytope Volume

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The volume computation problem

Input: Polytope $P := \{x \in \mathbb{R}^d \mid Ax \leq b\}$ $A \in \mathbb{R}^{m \times d}$, $b \in \mathbb{R}^m$

Output: Volume of P

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- ▶ no deterministic poly-time algorithm can compute the volume with less than exponential relative error [Elekes'86]
- ▶ randomized poly-time algorithm approximates the volume of a convex body with high probability and arbitrarily small relative error [DyerFriezeKannan'91] $O^*(d^{23}) \rightarrow O^*(d^4)$ [LovVemp'04]

Implementations

Exact: VINCI [Bueler et al'00], Latte [deLoera et al], Qhull [Barber et al], LRS [Avis], Normaliz [Bruns et al]

- ▶ triangulation, sign decomposition methods
- ▶ cannot compute in high dimensions (e.g. > 20)

Randomized:

- ▶ [LovàszDeàk'12] cannot compute in > 10 dimensions
- ▶ Matlab code by Cousins & Vempala based on [LovVemp'04]
- ▶ **Ours:** based on [DyerFriezeKannan'91], . . . , [KannanLovàszSimon.'97]

How do we compute a random point in a polytope P ?

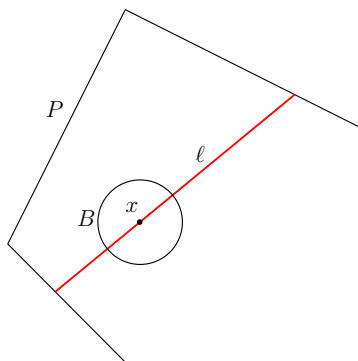
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- ▶ BUT for arbitrary polytopes we need *random walks*
e.g. grid walk, ball walk, **hit-and-run**

Random Directions Hit-and-Run (RDHR)



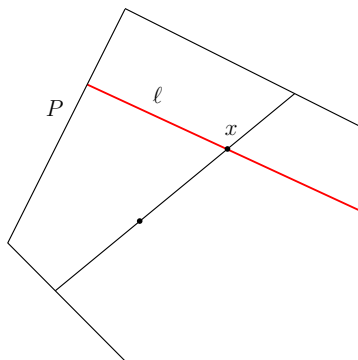
Input: point $x \in P$ and polytope $P \subset \mathbb{R}^d$

Output: a new point in P

1. line ℓ through x , uniform on $B(x, 1)$
2. set x to be a uniform distributed point on $P \cap \ell$

Iterate this for W steps and return x .

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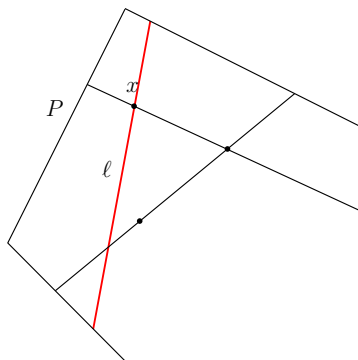
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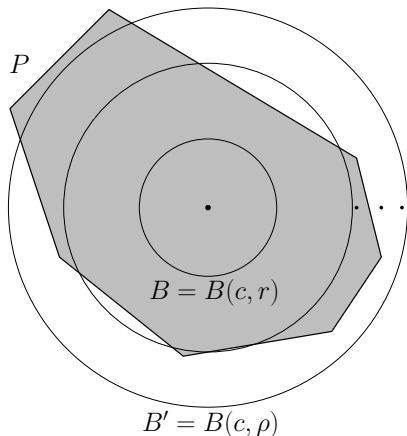
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- ▶ x is unif. random distrib. in P after $W = O^*(d^3)$ steps, where $O^*(\cdot)$ hides log factors [[LovaszVempala'06](#)]
- ▶ to generate **many** random points iterate this procedure

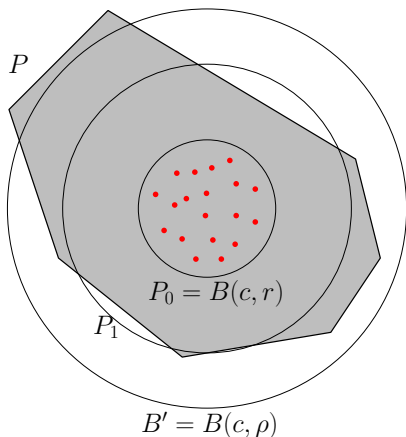
Multiphase Monte Carlo (Sequence of balls)



▶ $B(c, 2^{i/d})$, $i = \alpha, \alpha + 1, \dots, \beta$,
 $\alpha = \lfloor d \log r \rfloor$, $\beta = \lceil d \log \rho \rceil$

▶ $P_i := P \cap B(c, 2^{i/d})$, $i = \alpha, \alpha + 1, \dots, \beta$,
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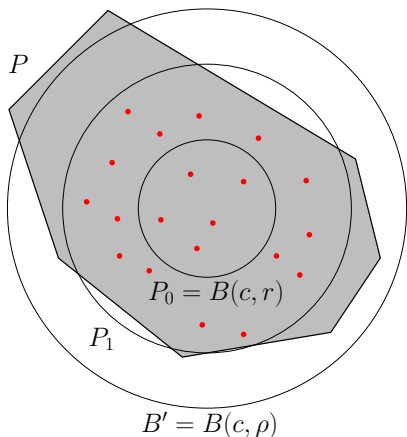


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1. Generate rand. points in P_i
2. Count how many rand. points in P_i fall in P_{i-1}

$$\text{vol}(P) = \text{vol}(P_\alpha) \prod_{i=\alpha+1}^{\beta} \frac{\text{vol}(P_i)}{\text{vol}(P_{i-1})}$$

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Complexity [KannanLS'97]

Assuming $B(c, 1) \subseteq P \subseteq B(c, \rho)$, the volume algorithm returns an estimation of $\text{vol}(P)$, which lies between $(1 - \epsilon)\text{vol}(P)$ and $(1 + \epsilon)\text{vol}(P)$ with probability $\geq 3/4$, making

$$O^*(d^5)$$

oracle calls, where ρ is the radius of a bounding ball for P .

Isotropic sandwiching: $\rho = O^*(\sqrt{d})$ and **ball walk**.

Runtime

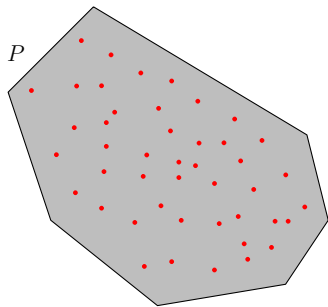
- ▶ generates $d \log(\rho)$ balls
- ▶ generate $N = 400\epsilon^{-2}d \log d$ random points in each ball $\cap P$
- ▶ each point is computed after $O^*(d^3)$ random walk steps

Modifications towards practicality

- ▶ $W = \lfloor 10 + d/10 \rfloor$ random walk steps (vs. $O^*(d^3)$ which hides constant 10^{11}) achieve $< 1\%$ error in up to 100 dim.
- ▶ sample **partial generations** of $\leq N$ points in each ball $\cap P$ (starting from the largest ball)
- ▶ coordinate (vs. random) directions hit-and-run (**CDHR**)
- ▶ implement **boundary oracles** with $O(m)$ runtime in CDHR

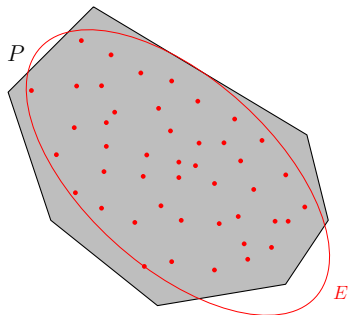
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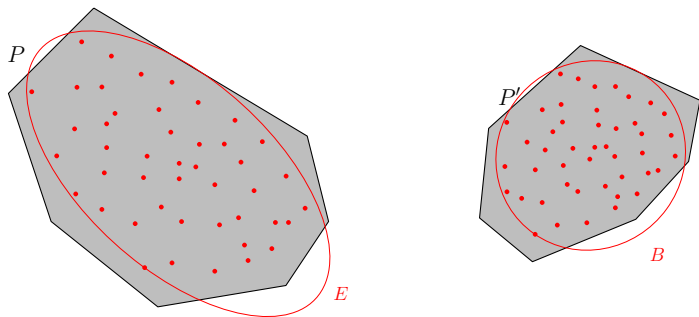
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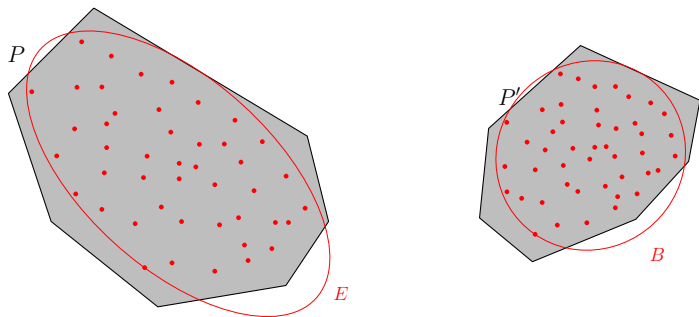
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Efficiently handle skinny polytopes in practice.

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- ▶ CDHR faster and more accurate than RDHR
- ▶ Compute the volume of Birkhoff polytopes B_{11}, \dots, B_{15} in few hrs whereas exact methods have only computed that of B_{10} by specialized software in ~ 1 year of parallel computation

Volumes of Birkhoff polytopes

n	d	estimation	asymptotic <small>[CanfieldMcKay09]</small>	<u>estimation</u> <u>asymptotic</u>	exact	<u>exact</u> <u>asymptotic</u>
3	4	1.12E+000	1.41E+000	0.7932847	1.13E+000	0.7973923
4	9	6.79E-002	7.61E-002	0.8919349	6.21E-002	0.8159296
5	16	1.41E-004	1.69E-004	0.8344350	1.41E-004	0.8341903
6	25	7.41E-009	8.62E-009	0.8598669	7.35E-009	0.8527922
7	36	5.67E-015	6.51E-015	0.8713891	5.64E-015	0.8665047
8	49	4.39E-023	5.03E-023	0.8729497	4.42E-023	0.8778632
9	64	2.62E-033	2.93E-033	0.8960767	2.60E-033	0.8874117
10	81	8.14E-046	9.81E-046	0.8305162	8.78E-046	0.8955491
11	100	1.40E-060	1.49E-060	0.9342584	???	???
12	121	7.85E-078	8.38E-078	0.9370513	???	???
13	144	1.33E-097	1.43E-097	0.9331517	???	???
14	169	5.96E-120	6.24E-120	0.9550089	???	???
15	196	5.70E-145	5.94E-145	0.9593786	???	???

Ongoing work

1. random walks for polytopes described by **optimization oracles**
e.g. resultant polytopes [Emiris,F,Konaxis,Penaranda SoCG'12]
2. use **approximate** oracles (utilizing approximate NN)
3. volume of more general convex bodies
e.g. **spectahedra**

Conclusion

- ▶ Practical volume estimation in high dimensions (e.g. 100)
- ▶ Software framework for testing theoretical ideas (e.g. new geometric random walks)

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