The Combinatorial Complexity of Approximating Polytopes

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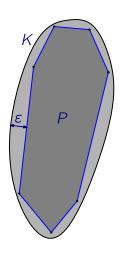
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SoCG 2016

Polytope Approximation

Problem description:

- Input: convex body K in d-dimensional space and parameter ε
- Output: succinct polytope P which ε-approximates K
- Question: how succinct can P be?
- ε -approximate: Hausdorff distance ε · diam(K)
- Succinct: Low combinatorial complexity (to be defined)
- Assume w.l.o.g. that diam(K) = 1
- Dimension d is a constant $(2^d/\varepsilon = O(1/\varepsilon))$



Uniform vs. Nonuniform Bounds

Nonuniform bounds:

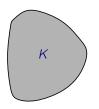
- Hold for $\varepsilon \leq \varepsilon_0$, where ε_0 depends on the input K
- Example: Gruber [Gru93] bounds the complexity n using the Gaussian curvature κ of the input

$$n = (1/\varepsilon)^{(d-1)/2} \int_{\partial K} \sqrt{\kappa(x)} dx$$

Uniform bounds: (our case)

- Hold for $\varepsilon < \varepsilon_0$, where ε_0 is a constant
- Example: Dudley [Dud74] and Bronshteyn and Ivanov [BI74] bound the maximum number of facets/vertices as a function of ε , d, and the diameter of the input

Bronshteyn and Ivanov's Approximation

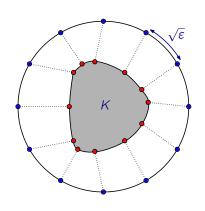


- Surround K by a sphere of radius 2
- 2 Distribute points on the sphere with distance $\sim \sqrt{\varepsilon}$
- Take the nearest neighbor on K for each point
- Make P the convex hull of the points

Bronshteyn and Ivanov, 1974:

A convex body K of diameter 1 can be ε -approximated by a polytope P with $O(1/\varepsilon^{(d-1)/2})$ vertices.

Bronshteyn and Ivanov's Approximation

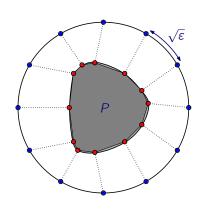


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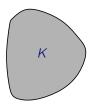


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Dudley's Approximation

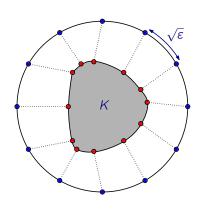


- Surround K by a sphere of radius 2
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Dudley's Approximation

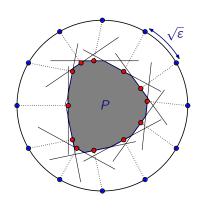


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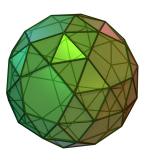
Combinatorial Complexity

• Faces have different dimensions:

```
0-face: vertex
1-face: edge
∴
(d − 1)-face: facet
```

• Combinatorial complexity:

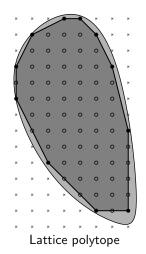
sum of number of k-faces $k = 0, \ldots, d-1$



Upper bound theorem

A polytope with n vertices (or facets) has combinatorial complexity $O(n^{\lfloor d/2 \rfloor})$.

Approximation of Low Combinatorial Complexity

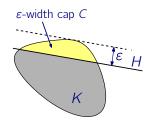


- Bronshteyn and Ivanov: optimal number of vertices $O(1/\varepsilon^{(d-1)/2})$
- Dudley: same optimal number of facets
- Upper bound theorem: their combinatorial complexity is $O(1/\varepsilon^{d^2/4})$ (Maybe it's much better, but we don't know how to prove)

Combinatorial complexity

- Best known bound: Roughly $O(1/\varepsilon^{d-2})$ using lattice polytopes [And63]
- Our bound^a: $\widetilde{O}(1/\varepsilon^{(d-1)/2})$
- $a\widetilde{O}$ hides $\log^{(d-1)/2} \frac{1}{\varepsilon}$

Caps

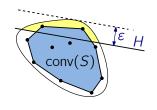


- Cap: intersection of K and a halfspace H
- Width: measured perpendicular to *H*
- A set S of points stabs all ε -width caps if for every ε -width cap C we have $C \cap S \neq \emptyset$

Approximation via Hitting Sets

If a point set $S \subset K$ stabs all ε -width caps, then $\operatorname{conv}(S)$ is an ε -approximation to K.

Caps



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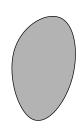
We identify two sets of regions:

- \bullet \mathcal{W} : witnesses
- \bullet C: collectors, one per witness

that satisfy

- (1) Each witness contributes one point to S
- (2) Any halfspace H either
 - Deep: contains a witness, or
 - Shallow: $H \cap K$ is contained within a collector
- (3) Each collector contains O(1) points of S

Witness-Collector Complexity Bound [Devillers et al. 2013

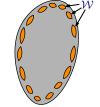


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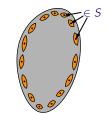


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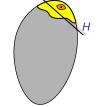
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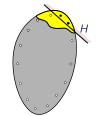
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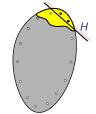
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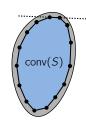
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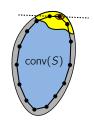
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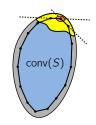
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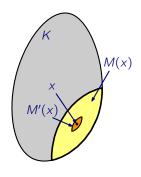
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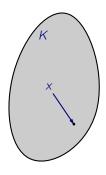


Macbeath Regions [Macbeath 52]

Given a convex body K, $x \in K$, and $\lambda > 0$:

- $M^{\lambda}(x) = x + \lambda((K x) \cap (x K))$
- $M(x) = M^1(x)$: intersection of K and K reflected around x
- $M'(x) = M^{1/5}(x)$

- M(x) resembles the minimum volume cap containing x
- $M'(x) \cap M'(y) \neq \emptyset \Rightarrow M'(x) \subseteq M(y)$

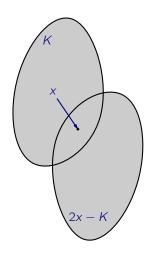


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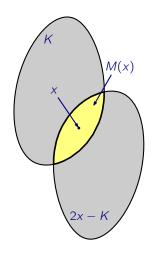


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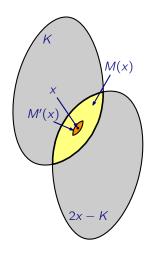


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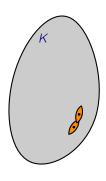


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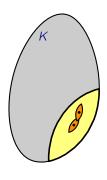


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Economical Cap Covering

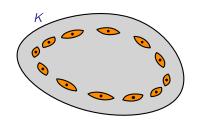
Economical Cap Covering

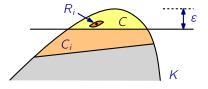
Given:

- K: convex body
- \bullet ε : small positive parameter

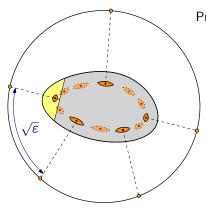
There exists [ELR70, Bar07]:

- Macbeath regions R_1, \ldots, R_k
- Caps C_1, \ldots, C_k of width $\Theta(\varepsilon)$
- $k = O(1/\varepsilon^{(d-1)/2})$ (new)
- For every cap C, there is i:
 - $R_i \subseteq C$ if width $> \varepsilon$
 - $C \subseteq C_i$ if width $\leq \varepsilon$





Cardinality Bound



Proof strategy:

- Prune Macbeath regions that are too close to each other (by increasing volume)
- A constant fraction of the regions are pruned
- Project the centers of the regions onto the Dudley ball (perpendicularly to the corresponding cap)
- Show that the pairwise distance in the Dudley ball is at least $\sqrt{\varepsilon}$

Approximating with ECC

Economical Cap Covering

Given:

- K: convex body
- ε : small positive parameter

There exists [ELR70, Bar07]:

- Macbeath regions R_1, \ldots, R_k
- Caps C_1, \ldots, C_k of width $\Theta(\varepsilon)$
- $k = O(1/\varepsilon^{(d-1)/2})$ (new)
- For every cap *C*, there is *i*:
 - $R_i \subseteq C$ if width $\geq \varepsilon$
 - $C \subseteq C_i$ if width $\leq \varepsilon$

Procedure:

- Create an ECC
- Place a point inside each Macbeath region
- Compute the convex hull

Approximation:

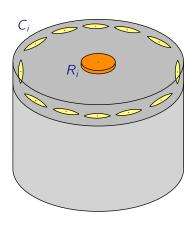
• Every ε -width cap is stabbed

Complexity:

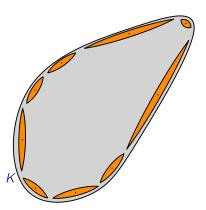
- $O(1/\varepsilon^{(d-1)/2})$ vertices
- Combinatorial complexity: $O(1/\varepsilon^{d^2/4})$ by the upper bound theorem

Better Bound on the Combinatorial Complexity

What if we use the Macbeath regions as witnesses and the caps as collectors?

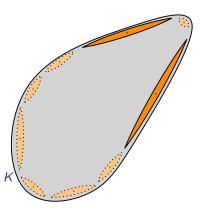


- Does not work for the traditional construction (for fixed width)
- We need one additional property:
 For all i, C_i intersects a constant number of bodies R_j



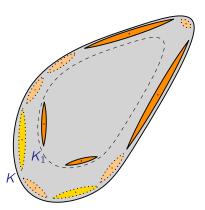
- Organize the Macbeath regions in O(log ½) layers by volume
- Larger volume Macbeath regions in outer layers
- Approximation error increases from ε to $O(\varepsilon \log \frac{1}{\varepsilon})$
- Scale ε accordingly: $\hat{\varepsilon} = \frac{\varepsilon}{\log \frac{1}{\varepsilon}}$
- Number of regions grows to

$$O\left(\left(\frac{1}{\hat{\varepsilon}}\right)^{\frac{d-1}{2}}\right) = O\left(\left(\frac{\log\frac{1}{\varepsilon}}{\varepsilon}\right)^{\frac{d-1}{2}}\right)$$



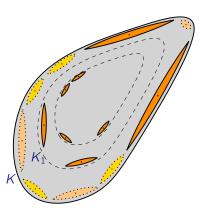
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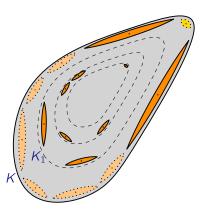
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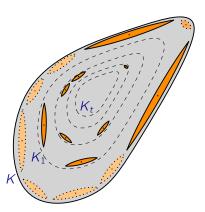
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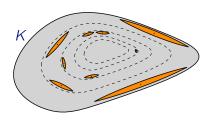
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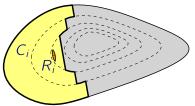


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Stratified Economical Cap Covering





Stratified ECC

Given:

- K: convex body
- ε : small positive parameter

There exists:

- Convex bodies R_1, \ldots, R_k
- Regions C_1, \ldots, C_k
- $k = \widetilde{O}(1/\varepsilon^{(d-1)/2})$
- Every ε -width cap C contains an R_i
- For every cap C, either $R_i \subseteq C$ or $C \subseteq C_i$ for some i
- For all i, C_i intersects a constant number of bodies R_i

Comparing Economical Cap Coverings

Economical Cap Covering

Given: K, ε

There exists:

- Convex bodies R_1, \ldots, R_k
- Caps C_1, \ldots, C_k of width $\Theta(\varepsilon)$
- $k = O(1/\varepsilon^{(d-1)/2})$
- For every cap *C*, there is *i*:
 - $R_i \subseteq C$ if width $\geq \varepsilon$
 - $C \subseteq C_i$ if width $\leq \varepsilon$

Stratified ECC

Given: K, ε

There exists:

- Convex bodies R_1, \ldots, R_k
- Non-convex regions C_1, \ldots, C_k
- $k = \widetilde{O}(1/\varepsilon^{(d-1)/2})$
- For every cap *C*, there is *i*:
 - $R_i \subseteq C$ if width $\geq \varepsilon$
 - $C \subseteq C_i$ if no $R_i \subseteq C$
- For all *i*, *C_i* intersects a constant number of bodies *R_i*

Succinct Approximation

Procedure:

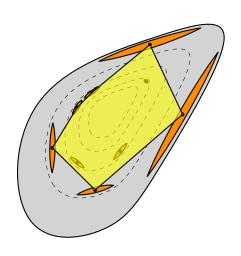
- Create a stratified ECC
- Place a point inside each R_i
- Compute the convex hull

Approximation:

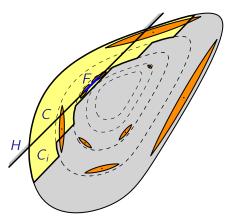
• Every ε -width cap is stabbed

Complexity:

- $\widetilde{O}(1/\varepsilon^{(d-1)/2})$ vertices
- Combinatorial complexity: $\widetilde{O}(1/\varepsilon^{(d-1)/2})$ Why?



Counting the Faces



- R_i are witnesses and C_i are collectors
- Charge each face F to a C_i and show each C_i receives O(1) charges
- H: supporting halfspace of F
- $C = K \cap H$ contains no R_i
- Some C_i contains C
- Charge F to C_i
- C_i intersects O(1) bodies R_i
 - $\rightarrow C_i$ contains O(1) points
 - $\rightarrow O(1)$ faces charge C_i

Conclusion and Open Problems

Our result:

First near-optimal bound on the combinatorial complexity of an ε -approximating polytope in \mathbb{R}^d :

$$O\left(\left(\frac{\log\frac{1}{\varepsilon}}{\varepsilon}\right)^{\frac{d-1}{2}}\right) = \widetilde{O}\left(\left(\frac{1}{\varepsilon}\right)^{\frac{d-1}{2}}\right)$$

Open problems:

- How long does it take to actually build the approximation?
- Can we analyze Dudley's construction?
- Can we get rid of the log factors?
- Can we compete against the optimal combinatorial complexity for the given K?

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Sculpture by Antony Gormley

Thank you!