# Approximating a Convex Body by a Polytope Using the Epsilon-Net Theorem 

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Abstract We prove that roughly $\frac{d}{(1-\vartheta)^{d}} \ln \frac{1}{(1-\vartheta)^{d}}$ points chosen uniformly and independently from a centered convex body $K$ in $\mathbb{R}^{d}$ yield a polytope $P$ for which $\vartheta K \subseteq P \subseteq K$ holds with large probability. This gives a joint generalization of results of Brazitikos, Chasapis and Hioni and of Giannopoulos and Milman.

Keywords Approximation by polytopes • Convex body • Epsilon-net theorem • Grünbaum's theorem • VC-dimension

Mathematics Subject Classification 52A27 • 52A20

## 1 Introduction

A convex body (i.e., a compact convex set with non-empty interior) in $\mathbb{R}^{d}$ is called centered, if its center of mass is the origin.

We study the following problem. Given a centered convex body $K$ in $\mathbb{R}^{d}$, a positive integer $t \geq d+1$, and $\delta, \vartheta \in(0,1)$. We want to show that under some assumptions on the parameters $d, t, \delta, \vartheta$ (and without assumptions on $K$ ), the convex hull $P$ of $t$ randomly, uniformly and independently chosen points of $K$ contains $\vartheta K$ with probability at least $1-\delta$.

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[4, Thm. 1.1] concerns the case of very rough approximation, that is, where the number $t$ of chosen points is linear in the dimension $d$. It states that the convex hull of $t=\alpha d$ random points in a centered convex body $K$ is a convex polytope $P$ which satisfies $\frac{c_{1}}{d} K \subseteq P$, with probability $1-\delta=1-e^{-c_{2} d}$, where $c_{1}, c_{2}>0$ and $\alpha>1$ are absolute constants. In our first result, we obtain explicit constants.

Theorem 1.1 Let $K$ be a centered convex body in $\mathbb{R}^{d}$. Choose $t=60(d+1)$ points $X_{1}, \ldots, X_{t}$ of $K$ randomly, independently and uniformly. Then

$$
\frac{1}{d} K \subseteq \operatorname{conv}\left\{X_{1}, \ldots, X_{t}\right\} \subseteq K
$$

with probability at least $1-4 e^{-d-1}$.
Another instance of our general problem is [7, Thm. 5.2], which concerns fine approximation, that is, where the number $t$ of chosen points is exponential in the dimension $d$. It states that for any $\delta, \gamma \in(0,1)$, if we choose $t=e^{\gamma d}$ random points in any centered convex body $K$ in $\mathbb{R}^{d}$, then the convex polytope $P$ thus obtained satisfies $c(\delta) \gamma K \subseteq P$, with probability $1-\delta$. We note that it is not included explicitly in the statement of of [7, Thm. 5.2] that it only holds for sufficiently large $d$, that is, when $d>d_{0}$, where $d_{0}$ depends on $\delta$ and $\gamma$. This condition is clearly necessary, as for any $\gamma$ and any $K$, with some positive probability, the origin is not in the convex hull of $t=e^{\gamma d}$ random points in $K$.
[7, Prop. 5.3] follows from the same argument as Theorem 5.2 therein. It states that for any $\delta, \vartheta \in(0,1)$, if we choose $t=c(\delta)\left(\frac{c}{1-\vartheta}\right)^{d}$ random points in any centered convex body $K$ in $\mathbb{R}^{d}$, then the convex polytope $P$ thus obtained satisfies $\vartheta K \subseteq P$, with probability $1-\delta$.

Our main result is the following.
Theorem 1.2 Let $\vartheta \in(0,1), C \geq 2$. Set

$$
t:=\left\lceil C \frac{(d+1) e}{(1-\vartheta)^{d}} \ln \frac{e}{(1-\vartheta)^{d}}\right\rceil .
$$

Then for any centered convex body $K$ in $\mathbb{R}^{d}$, if t points $X_{1}, \ldots, X_{t}$ of $K$ are chosen randomly, independently and uniformly, then

$$
\vartheta K \subseteq \operatorname{conv}\left\{X_{1}, \ldots, X_{t}\right\} \subseteq K
$$

with probability at least $1-\delta$, where

$$
\delta:=4\left[11 C^{2}\left(\frac{(1-\vartheta)^{d}}{e}\right)^{C-2}\right]^{d+1}
$$

By substituting $\vartheta=\frac{1}{d}, C=6$, we obtain Theorem 1.1.

In order to recover [7, Thm. 5.2], substitute $C=3$ and $\vartheta=c(\delta) \gamma$ in our Theorem 1.2. Then $t \leq e^{3 c(\delta) \gamma d}$, when $d$ is large, and $\delta$ is roughly $e^{-c(\delta) \gamma d^{2}}$. Fixing $c(\delta)=1 / 3$ independently of $\delta$ yields the result.

We recover [7, Prop. 5.3] in a form which is slightly weaker if $\vartheta$ is close to 1 , as follows. In our Theorem 1.2, $t \leq \frac{10 C d^{2}}{(1-\vartheta)^{d+1}}$ (note the exponent $d+1$ instead of $d$ ) and $\delta \leq 11 C^{2} / e^{C-2}$. By setting $C$ sufficiently large (depending on the desired $\delta$ only), we can make the latter as small as required.

We compare our Theorem 1.2 with the main result, [4, Thm. 1.2], which states the following. Let $\beta \in(0,1)$. There exist a constant $\alpha=\alpha(\beta)>1$ depending only on $\beta$ and an absolute constant $c>0$ with the following property. Let $K$ be a centered convex body in $\mathbb{R}^{d}, \alpha d \leq t \leq e^{d}$, and choose $t$ points uniformly distributed in $K$. Then the convex polytope thus obtained contains $\vartheta K$, where $\vartheta=\frac{c \beta \ln (t / d)}{d}$ with probability $1-\delta$, where $\delta \leq \exp \left(-t^{1-\beta} d^{\beta}\right)$.

When $\vartheta$ is of order $1 / d$, the two results are the same up the constants involved, see our Theorem 1.1 and the discussion preceding it. For fine approximation, that is, when $\vartheta$ is a constant, by setting $C=\frac{1}{(1-\vartheta)^{d / 2}}$, we obtain roughly $t \approx \exp (\vartheta d / 2)$ and $\delta \approx \exp \left[-\vartheta d^{2} \exp (\vartheta d / 2)\right]$. In the mean time, [4, Thm. 1.2] gives roughly $t \approx$ $\exp (\vartheta d /(c \beta))$ and $\delta \approx \exp \left[-\exp ((1-\beta) \vartheta d /(c \beta)) d^{\beta}\right]$.

In Sect. 2, we present a generalization of a classical result of Grünbaum [10], according to which any half-space containing the center of mass of a convex body contains at least a $1 / e$ fraction of its volume. In Sect. 3, we state a specific form of the $\varepsilon$-net theorem, a result from combinatorics obtained by Haussler and Welzl [11] building on ideas of Vapnik and Chervonenkis [21], and then refined by Komlós et al. [12]. In Sect. 4, we combine these two to obtain Theorem 1.2. Finally, in Sect. 5, using a recent result of Fradelizi et al. [6], we extend our main result to approximating a linear section of a centered convex body.

For surveys on the topic of approximation of convex bodies by polytopes, cf. [2, 5, 9], and for some further recent results on approximation in the Banach-Mazur distance (or, geometric distance) when the vertices are not necessarily picked randomly and uniformly from the body, see [3,16].

We note that, in a similar vein, Gordon, Litvak, Pajor and Tomczak-Jaegermann [8, Thm. 3.1] showed that if $K$ is an origin-symmetric convex body in $\mathbb{R}^{d}$ and $t=(4 / \varepsilon)^{2 d}$ random points $X_{1}, \ldots, X_{t}$ are chosen from it uniformly and independently, then, with probability larger than $1-\exp \left(-(8 / \varepsilon)^{d} / 2\right)$, these $t$ points form a metric $\varepsilon$-net of $K$ with respect to $K$, that is, $K \subseteq \bigcup_{i=1}^{t}\left(X_{i}+\varepsilon K\right)$. We will use the term ' $\varepsilon$-net' in a different, combinatorial sense, to be defined in Sect. 3.

## 2 Convexity: A stability Version of a Theorem of Grünbaum

Grünbaum's theorem [10] states that for any centered convex body $K$ in $\mathbb{R}^{d}$, and any half-space $F_{0}$ that contains the origin we have

$$
\begin{equation*}
\operatorname{vol}(K) / e \leq \operatorname{vol}\left(K \cap F_{0}\right), \tag{1}
\end{equation*}
$$

where vol (•) denotes volume.
We say that a half-space $F$ supports $K$ from outside if the boundary of the half-space intersects bd $K$, but $F$ does not intersect the interior of $K$. Lemma 2.1, is a stability version of Grünbaum's theorem.

Lemma 2.1 Let $K$ be a convex body in $\mathbb{R}^{d}$ with centroid at the origin. Let $0<\vartheta<1$, and $F$ be a half-space that supports $\vartheta K$ from outside. Then

$$
\begin{equation*}
\operatorname{vol}(K) \frac{(1-\vartheta)^{d}}{e} \leq \operatorname{vol}(K \cap F) \tag{2}
\end{equation*}
$$

Proof Let $F_{0}$ be a translate of $F$ containing $o$ on its boundary, and let $F_{1}$ be a translate of $F$ that supports $K$ from outside. Finally, let $p \in \operatorname{bd} F_{1} \cap K$. Then $\vartheta p+(1-\vartheta)\left(K \cap F_{0}\right)$ (that is, the homothetic copy of $K \cap F_{0}$ with homothety center $p$ and ratio $1-\vartheta$ ) is in $K \cap F$. Its volume is $(1-\vartheta)^{d} \operatorname{vol}\left(K \cap F_{0}\right)$, which by $(1)$, is at least $(1-\vartheta)^{d} \operatorname{vol}(K) / e$, finishing the proof.

## 3 Combinatorics: The $\varepsilon$-Net Theorem of Haussler and Welzl

Definition 3.1 Let $\mathcal{F}$ be a family of subsets of some set $U$. The Vapnik-Chervonenkis dimension ( $V C$-dimension, in short) of $\mathcal{F}$ is the maximal cardinality of a subset $V$ of $U$ such that $V$ is shattered by $\mathcal{F}$, that is, $\{F \cap V: F \in \mathcal{F}\}=2^{V}$.

A transversal of the set family $\mathcal{F}$ is a subset $Q$ of $U$ that intersects each member of $\mathcal{F}$.

Let $\varepsilon \in(0,1)$ be given. When $U$ is equipped with a probability measure for which each member of $\mathcal{F}$ is measurable, then a transversal of those members of $\mathcal{F}$ that are of measure at least $\varepsilon$ is called an $\varepsilon$-net.

It follows from Radon's lemma (cf. [13, Thm. 1.3.1], or [19, Thm. 1.1.5]) that if $U$ is any subset of $\mathbb{R}^{d}$, and $\mathcal{F}$ is a family of half-spaces of $\mathbb{R}^{d}$, then the VC-dimension of $\mathcal{F}$ is at most $d+1$.

The $\varepsilon$-Net Theorem was first proved by Haussler and Welzl [11], and then improved by Komlós et al. [12]. We state a slightly weaker form of Theorem 3.1 of [12] than the original, in order to have an explicit bound on the probability $\delta$ of failure.

Lemma 3.2 ( $\varepsilon$-Net Theorem). Let $0<\varepsilon<1 / e, C \geq 2$, and let $D$ be a positive integer. Let $\mathcal{F}$ be a family of some measurable subsets of a probability space ( $U, \mu$ ), where the probability of each member $F$ of $\mathcal{F}$ is $\mu(F) \geq \varepsilon$. Assume that the $V C$ dimension of $\mathcal{F}$ is at most $D$. Set

$$
t:=\left\lceil C \frac{D}{\varepsilon} \ln \frac{1}{\varepsilon}\right\rceil
$$

Choose telements $X_{1}, \ldots, X_{t}$ of $V$ randomly, independently according to $\mu$. Then $\left\{X_{1}, \ldots, X_{t}\right\}$ is a transversal of $\mathcal{F}$ with probability at least $1-\delta$, where

$$
\delta:=4\left[11 C^{2} \varepsilon^{C-2}\right]^{D}
$$

Proof We provide an outline of the first, conceptual part of the proof closely following [17, Thm. 15.5]. Then, we continue with a detailed computation to obtain the bound on the probability stated in Lemma 3.2.

Let $T>t$ be an integer, to be set later. We select (with repetition) independently $t$ random elements of $U$ with respect to $\mu$, call it the first sample, and denote it by $x$. Then, we choose another $T-t$ elements, call it the second sample, and denote it by $y$. For any $F \in \mathcal{F}$, and any finite sequence $w$ of elements of $U$, let $I(F, w)$ denote the number of elements of $w$ in $F$ with multiplicity. Let $m_{F}$ denote the median of $I(F, y)$.

Note that $I(F, y)$ is a binomial variable, and hence, its mean and median are close to each other. More precisely,

$$
\begin{equation*}
m_{F} \geq(T-t) \varepsilon-1 \tag{3}
\end{equation*}
$$

It is not hard to see that

$$
\begin{equation*}
\mu(\exists F \in \mathcal{F}: I(F, x)=0) \leq 2 \mu\left(\exists F \in \mathcal{F}: I(F, x)=0 \text { and } I(F, y) \geq m_{F}\right) \tag{4}
\end{equation*}
$$

Denote the concatenation of the two sequences $x$ and $y$ by $\overline{x y}$. Fix any length $T$ sequence $z$ of elements of $U$.

It is simple to obtain a bound on the following conditional probability:

$$
\begin{align*}
\mu(\exists F \in \mathcal{F}: I(F, x)= & \left.0 \operatorname{and} I(F, y) \geq m_{F} \mid \overline{x y}=z\right) \\
& \leq \chi\left[I(F, z) \geq m_{F}\right]\left(1-\frac{t}{T}\right)^{m_{F}} \tag{5}
\end{align*}
$$

where $\chi$ denotes the indicator function of an event, that is, it is one if the event holds, and zero otherwise.

The key idea follows. Consider $z$ as a set. Then, by the Shatter function lemma (cf. of [17, Thm. 15.4] or [13, Lem. 10.2.5]) proved independently by Shelah [20], Sauer [18] and Vapnik and Chervonenkis [21], $z$ has at most

$$
\sum_{i=0}^{D}\binom{T}{i}
$$

distinct intersections with members of $\mathcal{F}$. Thus by (3) and (5), we have

$$
\begin{equation*}
\mu\left(\exists F \in \mathcal{F}: I(F, x)=0 \text { and } I(F, y) \geq m_{F} \mid \overline{x y}=z\right) \leq \sum_{i=0}^{D}\binom{T}{i}\left(1-\frac{t}{T}\right)^{(T-t) \varepsilon-1} \tag{6}
\end{equation*}
$$

Let $E$ be the 'bad' event, that is, when $\left\{X_{1}, \ldots, X_{t}\right\}$ is not a transversal of $\mathcal{F}$. So far, by (4) and (6), we obtained that for any integer $T>t$, we have that the probability of the event $E$ is

$$
\mu(E)<2 \sum_{i=0}^{D}\binom{T}{i}\left(1-\frac{t}{T}\right)^{(T-t) \varepsilon-1}
$$

From this point on, we describe the computations in detail, in order to obtain the bound on the probability stated in Lemma 3.2.

We set $T=\left\lfloor\frac{\varepsilon t^{2}}{D}\right\rfloor$ and use $\sum_{i=0}^{D}\binom{T}{i} \leq\left(\frac{e T}{D}\right)^{D}$, to obtain that

$$
\begin{aligned}
\mu(E) & <2\left(\frac{e T}{D}\right)^{D}\left(1-\frac{t}{T}\right)^{(T-t) \varepsilon-1}<2\left(\frac{e \varepsilon t^{2}}{D^{2}}\right)^{D}\left(1-\frac{D}{\varepsilon t}\right)^{\left(\varepsilon t^{2} / D-t-1\right) \varepsilon-1} \\
& <2\left(\frac{e \varepsilon t^{2}}{D^{2}}\right)^{D} e^{-\varepsilon t+D+D / t+D /(\varepsilon t)},
\end{aligned}
$$

which, after substituting the expression for $t$ in some places and using $\varepsilon<1 / e$, is at most

$$
\left(2 e^{1 / C+1 /(e C)}\right)\left(\frac{e^{2} \varepsilon t^{2}}{D^{2}}\right)^{D} \varepsilon^{C D}
$$

which, using $C \geq 2$ is at most

$$
\left(2 e^{1 / C+1 /(e C)}\right)\left(\frac{e^{2}(1+1 /(2 e))^{2} C^{2} \ln ^{2}(1 / \varepsilon)}{\varepsilon}\right)^{D} \varepsilon^{C D}<4\left(11 C^{2} \varepsilon^{C-2}\right)^{D}
$$

completing the proof of Lemma 3.2.
For more on the theory of $\varepsilon$-nets, see $[1,13,15,17]$.

## 4 Proof of Theorem 1.2

Proof of Theorem 1.2 We consider the following set system on the base set $K$ :

$$
\mathcal{F}:=\{K \cap F: F \text { is a half-space that supports } \vartheta K \text { from outside }\} .
$$

Clearly, the VC-dimension of $\mathcal{F}$ is at most $D:=d+1$. Let $\mu$ be the Lebesgue measure restricted to $K$, and assume that $\operatorname{vol}(K)=1$, that is, that $\mu$ is a probability measure. By (2), we have that each set in $\mathcal{F}$ is of measure at least $\varepsilon:=\frac{(1-\vartheta)^{d}}{e}$. Lemma 3.2 yields that if we choose $t$ points of $K$ independently with respect to $\mu$ (that is, uniformly), then with probability at least $1-\delta$, we obtain a set $Q \subseteq K$ that intersects every member of $\mathcal{F}$. The latter is equivalent to $\vartheta K \subseteq$ conv $Q$, completing the proof.

## 5 Approximating a Section of a Convex Body

Let $K$ be a centered convex body in $\mathbb{R}^{d}$, and $V$ a linear subspace of $\mathbb{R}^{d}$. Now, $K \cap V$ may not be centered however, we may still want to approximate $K \cap V$ with a polytope $P \subset K \cap V$ such that $\vartheta(K \cap V) \subset P$ for some not too small $\vartheta$.

The main result of [6] (for further results, see also [14]) states that there is an absolute constant $c>0$ such that for every centered convex body $K$ in $\mathbb{R}^{d}$, every
( $d-k$ )-dimensional linear subspace $V$ of $\mathbb{R}^{d}, 0 \leq k \leq d-1$, and any $u \in V$ unit vector, we have

$$
\begin{equation*}
\operatorname{vol}_{d-k}\left(K \cap V \cap u^{+}\right) \geq \frac{c}{(k+1)^{2}}\left(1+\frac{k+1}{d-k}\right)^{-(d-k-2)} \operatorname{vol}_{d-k}(K \cap V) \tag{7}
\end{equation*}
$$

where $u^{+}=\left\{x \in \mathbb{R}^{d}:\langle u, x\rangle \geq 0\right\}$ is the half-space with inner normal vector $u$.
Using this result, our proof of Theorem 1.2 immediately yields the following.
Theorem 5.1 Let $\vartheta \in(0,1), C \geq 2$. Let $K$ be a centered convex body in $\mathbb{R}^{d}$ and $V$ be $(d-k)$-dimensional linear subspace of $\mathbb{R}^{d}$ with $0 \leq k \leq d-1$. Set

$$
t:=\left\lceil C \frac{(d-k+1)(k+1)^{2}}{c\left(1+\frac{k+1}{d-k}\right)^{d-k-2}(1-\vartheta)^{d-k}} \ln \frac{(k+1)^{2}}{c\left(1+\frac{k+1}{d-k}\right)^{d-k-2}(1-\vartheta)^{d-k}}\right\rceil,
$$

where $c$ is the universal constant from (7). Choose $t$ points $X_{1}, \ldots, X_{t}$ of $K \cap V$ randomly, independently and uniformly with respect to the $(d-k)$-dimensional Lebesgue measure on $V$. Then

$$
\vartheta(K \cap V) \subseteq \operatorname{conv}\left\{X_{1}, \ldots, X_{t}\right\} \subseteq K \cap V,
$$

with probability at least $1-\delta$, where

$$
\delta:=4\left[11 C^{2}\left(\frac{c\left(1+\frac{k+1}{d-k}\right)^{d-k-2}(1-\vartheta)^{d-k}}{(k+1)^{2}}\right)^{C-2}\right]^{d-k+1} .
$$

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