

On the isotropy constant of projections of polytopes [☆]

David Alonso-Gutiérrez ^{a,*}, Jesús Bastero ^a, Julio Bernués ^a, Paweł Wolff ^b

^a *Departamento de Matemáticas, Facultad de Ciencias, Universidad de Zaragoza, 50009 Zaragoza, Spain*

^b *Department of Mathematics, Case Western Reserve University, Cleveland, OH 44106-7058, USA*

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Abstract

The isotropy constant of any d -dimensional polytope with n vertices is bounded by $C\sqrt{n/d}$ where $C > 0$ is a numerical constant.

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1. Introduction

The boundedness of the isotropy constant (see definition below) is a major conjecture in Asymptotic Geometric Analysis. The answer is known to be positive for many families of convex bodies, see for instance [15] or [12] and the references therein. In this paper we focus our attention on the isotropy constant of polytopes or, equivalently, of projections of the unit ball of ℓ_1^n space (in the symmetric case) and of the regular n -dimensional simplex S_n (in the non-symmetric case).

M. Junge [9] proved that the isotropy constant of all orthogonal projections of B_p^n , the unit ball of the ℓ_p^n space, $1 < p \leq \infty$, is bounded by Cp' an estimate improved to $C\sqrt{p'}$ in [13] (p' is the

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* Corresponding author.

E-mail addresses: daalonso@unizar.es (D. Alonso-Gutiérrez), bastero@unizar.es (J. Bastero), bernues@unizar.es (J. Bernués), pawel.wolff@case.edu (P. Wolff).

conjugate exponent of p and C a numerical constant). Later [10, Theorem 4] M. Junge showed that the isotropy constant of any symmetric polytope with $2n$ vertices is bounded by $C \log n$, see also [14].

In a recent paper [12], B. Klartag and G. Kozma show the boundedness of the isotropy constant of random Gaussian polytopes. The integral over a polytope, which defines its isotropy constant, is computed by passing to an integral over its surface (faces). A consequence of their results is that “most” (see precise meaning below) d -dimensional projections of B_1^n as well as of S_n have bounded isotropy constant. When reading this statement one should have in mind the well-known fact that every symmetric convex body in \mathbb{R}^d is “almost” a projection of a B_1^n -ball, with possibly large n . In the same spirit, positive answers for other random d -dimensional polytopes with $n (\geq Cd)$ vertices were given in [1,6].

Our main theorem (Corollary 3.5) states that for any d -dimensional polytope K with n vertices its isotropy constant L_K verifies

$$L_K \leq C \sqrt{\frac{n}{d}},$$

where $C > 0$ is a numerical constant.

We now pass to describe the contents of the paper. The second section introduces the geometric tool (Proposition 2.1) necessary to deal with integration on d -dimensional projections of polytopes (Corollary 2.8). Some time ago, one of the authors learned about this tool from Prof. Franck Barthe. The ideas originate from a paper by U. Betke [3], where a general result was presented, namely a related formula for mixed volumes of two polytopes. However, for the sake of completeness, we provide the proof of the particular result we need. It also seems that the content of the proof is more geometric.

In the third section we use these tools to prove our aforementioned main result (Corollary 3.5) by easily reducing it to the cases $K = P_E B_1^n$ or $K = P_E S_n$ (Theorem 3.4) where $E \subset \mathbb{R}^n$ is any d -dimensional subspace and P_E is the orthogonal projection onto E . Also in this section we give a proof of the observation that for “most” subspaces, that is, for a subset A of the Grassmann space $G_{n,d}$ of Haar probability measure $\geq 1 - c_1 e^{-c_2 \max\{\log n, d\}}$, one has $L_{P_E B_1^n} < C$ and $L_{P_E S_n} < C$ for every $E \in A$ with numerical constants C, c_1, c_2 (Proposition 3.3).

The next section studies the isotropy constant of projections of random polytopes with vertices on the sphere S^{n-1} . Using the techniques from Section 2 and [1] we show that, with high probability, the isotropy constants of all d -dimensional projections of random polytopes are bounded by $C \sqrt{\frac{n}{d}}$ (Proposition 4.1).

In the last section we show that for every isotropic convex body, the isotropy constants of its hyperplane projections are comparable to the isotropy constant of the body itself (Corollary 5.1). Recall that the analogous result for hyperplane sections was already proved in [15]. The proof uses Steiner symmetrization in a similar way as it appears in [5], with better numerical constants. In particular, we have $L_{P_H B_p^n} \leq C$ for any hyperplane H and $1 < p < \infty$ improving Junge’s estimate [9] for the case of hyperplanes. In [2] a different proof of this fact is given with the hope it might be extended to lower-dimensional projections.

We recall that a convex body $K \subset \mathbb{R}^n$ is *isotropic* if it has volume $\text{Vol}_n(K) = 1$, the barycenter of K is at the origin and its inertia matrix is a multiple of the identity. Equivalently, there exists a constant $L_K > 0$ called isotropy constant of K such that $L_K^2 = \int_K \langle x, \theta \rangle^2 dx, \forall \theta \in S^{n-1}$.

It is well known [15], that every convex body $K \subset \mathbb{R}^n$ has an affine transformation K_1 isotropic, so we can write $L_K := L_{K_1}$. This is well defined and moreover,

$$nL_K^2 = \inf \left\{ \frac{1}{\text{Vol}_n(K)^{\frac{2}{n+1}}} \int_{a+TK} |x|^2 dx; a \in \mathbb{R}^n, T \in SL(n) \right\}. \quad (1.1)$$

For a convex body $K \subseteq \mathbb{R}^n$, $r(K) = \min\{|x|: x \in K\}$ is the inradius of K .

We will think of S_n as an n -dimensional regular simplex in \mathbb{R}^n with center of mass at the origin. We will write $\Delta_n = \text{conv}\{e_1, \dots, e_{n+1}\}$ for the natural position of an n -dimensional regular simplex in \mathbb{R}^{n+1} .

The Lebesgue measure on an affine subspace E will be denoted by λ_E . For a measurable set $A \subseteq E$, if d is a dimension of E , $\text{Vol}_d(A)$ will stand for $\lambda_E(A)$.

The notation $a \sim b$ means $a \cdot c_1 \leq b \leq a \cdot c_2$ for some numerical constants $c_1, c_2 > 0$.

2. Projections of polytopes

Throughout this section, $K \subseteq \mathbb{R}^n$ is a polytope (non-empty but possibly of empty interior), $E \subseteq \mathbb{R}^n$ is a linear subspace of dimension d ($1 \leq d \leq n-1$) and P_E is the orthogonal projection onto E .

Let us fix some notation and recall necessary definitions (we follow the book by Schneider [18, Chapters 1, 2]). For a subset $A \subseteq \mathbb{R}^n$, $\text{aff } A$ denotes the minimal affine subspace which contains A . The dimension of a convex set A is $\dim \text{aff } A$. When writing $\text{relint } A$ we mean the relative interior of A w.r.t. the topology of $\text{aff } A$. If $G \subseteq \mathbb{R}^n$ is an affine subspace then G_0 denotes the linear subspace parallel to G . A convex subset $F \subseteq K$ of a polytope K is called a *face* if for any $x, y \in K$, $(x+y)/2 \in F$ implies $x, y \in F$ (see also [18, Section 1.4, p. 18]). The set of j -dimensional faces (j -faces, in short) of K will be denoted as $\mathcal{F}_j(K)$ ($j = 0, 1, \dots, n$), and $\mathcal{F}(K) = \bigcup_{j=0}^n \mathcal{F}_j(K) \cup \{\emptyset\}$ is the set of all faces of K (\emptyset is also a face). K can be decomposed into a disjoint union of $\{\text{relint } F; F \in \mathcal{F}(K)\}$ (see [18, Theorem 2.1.2]). For that reason for any $x \in K$ the unique face $F \in \mathcal{F}(K)$ such that $x \in \text{relint } F$ will be denoted by $F(K, x)$.

For $x \in K$, a *normal cone* of K at x is

$$N(K, x) = \{u \in \mathbb{R}^n; \forall_{z \in K} \langle z - x, u \rangle \leq 0\}.$$

$N(K, x)$ is a closed convex cone. We shall also consider another closed convex cone, namely

$$S(K, x) = \bigcup_{\lambda > 0} \lambda(K - x).$$

(In general, i.e. when K is a convex body, $S(K, x)$ does not have to be closed.) By [18, (2.2.1)],

$$N(K, x)^* = S(K, x), \quad (2.2)$$

where the polarity used here is the polarity of convex cones, namely, if $C \subseteq \mathbb{R}^n$ is a convex cone,

$$C^* = \{y \in \mathbb{R}^n; \forall_{x \in C} \langle x, y \rangle \leq 0\}$$

(see also [18, Section 1.6, p. 34]). We shall also need to consider normal cones taken w.r.t. an affine subspace. If G is an affine subspace of \mathbb{R}^n and $L \subseteq G$ is a convex body, then for $x \in L$ we define a normal cone for L at x taken w.r.t. G :

$$N_G(L, x) = \{u \in G_0; \forall_{z \in L} \langle z - x, u \rangle \leq 0\}.$$

Note that $N_G(L, x) \subseteq G_0$. The similar duality relation to (2.2) holds:

$$N_G(L, x)^{*G_0} = S(L, x), \quad (2.3)$$

where the polarity is taken w.r.t. G_0 .

For any face $\emptyset \neq F \in \mathcal{F}(K)$, define

$$N(K, F) := N(K, x), \quad \text{where } x \in \text{relint } F.$$

This definition does not depend on the choice of x (see [18, Section 2.2, p. 72]). $N_G(L, F)$ is analogously defined.

For a given polytope $K \subseteq \mathbb{R}^n$ and a linear subspace $E \subseteq \mathbb{R}^n$ of dimension d , let us fix any $u \in E^\perp \setminus \{0\}$ which satisfy

$$u \notin \bigcup \{P_{E^\perp} N(K, F); F \in \mathcal{F}(K) \setminus \{\emptyset\}, \dim P_{E^\perp} N(K, F) \leq n - d - 1\}. \quad (2.4)$$

Clearly, such u exists, since (2.4) excludes only a finite union of sets of dimension $< n - d$ from E^\perp which is of dimension $n - d$.

Consider the following subsets of $\mathcal{F}(K)$:

$$\begin{aligned} \tilde{\mathcal{F}}(K, E, u) &:= \{F \in \mathcal{F}(K); u \in P_{E^\perp} N(K, F)\}, \\ \tilde{\mathcal{F}}_d(K, E, u) &:= \tilde{\mathcal{F}}(K, E, u) \cap \mathcal{F}_d(K). \end{aligned}$$

Proposition 2.1. *Let $K \subset \mathbb{R}^n$ be a polytope, E a d -dimensional subspace, $u \in E^\perp$ verifying (2.4) and $\tilde{\mathcal{F}}(K, E, u)$ as described above. Then*

- (a) $\{P_E(\text{relint } F); F \in \tilde{\mathcal{F}}(K, E, u)\}$ is a family of pair-wise disjoint sets,
- (b) $\bigcup \{P_E F; F \in \tilde{\mathcal{F}}(K, E, u)\} = P_E K$.

Moreover, $\tilde{\mathcal{F}}(K, E, u) \subseteq \bigcup_{0 \leq j \leq d} \mathcal{F}_j(K)$ and for each $F \in \tilde{\mathcal{F}}_d(K, E, u)$, $P_E|_F : F \rightarrow P_E F$ is an affine isomorphism.

In the proof of the proposition we shall use several lemmas.

Lemma 2.2. *Let L be a polytope in \mathbb{R}^n and $G \subseteq \mathbb{R}^n$ be an affine subspace. If $x \in L \cap G$ then*

$$P_{G_0} N(L, x) = N_G(L \cap G, x).$$

Proof. By taking polars w.r.t. G_0 we see that the assertion is equivalent to

$$N(L, x)^* \cap G_0 = N_G(L \cap G, x)^{*G_0} \quad (2.5)$$

(for the l.h.s. we used the fact that for a convex cone C , $(P_{G_0} C)^{*G_0} = C^* \cap G_0$). Since $G_0 = G - x$ by (2.2) we get

$$\begin{aligned} N(L, x)^* \cap G_0 &= S(L, x) \cap (G - x) = \bigcup_{\lambda > 0} \lambda(L - x) \cap (G - x) \\ &= S(L \cap G, x). \end{aligned}$$

Applying (2.3) we see that the r.h.s. of (2.5) is also equal to $S(L \cap G, x)$. \square

Lemma 2.3. (See [18, Section 2.2].) Let L be a polytope contained in an affine subspace $G \subseteq \mathbb{R}^n$. Then

$$\bigcup_{F \in \mathcal{F}_0(L)} N_G(L, F) = G_0.$$

Lemma 2.4. With the hypothesis as in the previous lemma, for $x, y \in L$,

$$N_G(L, x) \cap N_G(L, y) = N_G(L, (x + y)/2).$$

Proof. The inclusion \subseteq is immediate from the definition of a normal cone. For the converse inclusion take $u \in N_G(L, (x + y)/2)$. Then $\langle x - \frac{x+y}{2}, u \rangle \leq 0$, $\langle y - \frac{x+y}{2}, u \rangle \leq 0$, so $\langle x - y, u \rangle = 0$. Now, for all $z \in L$,

$$\langle z - x, u \rangle = \left\langle z - \frac{x + y}{2}, u \right\rangle + \left\langle \frac{y - x}{2}, u \right\rangle \leq 0,$$

so $u \in N_G(L, x)$. Similarly $u \in N_G(L, y)$. \square

Lemma 2.5. (See [18, Section 2.4].) With the hypothesis as in Lemma 2.3, for $\emptyset \neq F \in \mathcal{F}(L)$,

$$\dim N_G(L, F) = \dim G - \dim F.$$

Remark 2.6. Actually we shall use only the inequality $\dim N_G(L, F) \leq \dim G - \dim F$ which simply follows from the fact $N_G(L, F) \subseteq ((\text{aff } F)_0)^{\perp_{G_0}}$.

Lemma 2.7. Let $K \subseteq \mathbb{R}^n$ be a polytope, $E \subseteq \mathbb{R}^n$ be a linear subspace of dimension d and $u \in E^\perp$ satisfies (2.4). Let $y \in K$, $x = P_E y \in E$, $K_x = K \cap (x + E^\perp)$ (K_x is a polytope in $x + E^\perp$). If one of the equivalent condition holds:

- (i) $u \in P_{E^\perp} N(K, y)$,
- (ii) $u \in N_{x+E^\perp}(K_x, y)$,

then $\{y\} \in \mathcal{F}_0(K_x)$ and $\dim F(K, y) \leq d$.

Proof. The conditions (i) and (ii) are equivalent by Lemma 2.2. Consider $F = F(K, y)$ and $F' = F(K_x, y)$. By the condition (2.4) on u ,

$$\dim P_{E^\perp} N(K, F) \geq n - d,$$

so $\dim N_{x+E^\perp}(K_x, F') \geq n - d$ and also $\dim N(K, F) \geq n - d$. Therefore Lemma 2.5 applied to K_x and F' implies $\dim F' = 0$, so $\{y\} = F' \in \mathcal{F}_0(K_x)$. Eventually, applying the same lemma to K and F yields $\dim F \leq d$. \square

Proof of Proposition 2.1. (a) Take $F_1, F_2 \in \tilde{\mathcal{F}}(K, E, u)$ such that for some $x \in E$,

$$x \in P_E(\text{relint } F_1) \cap P_E(\text{relint } F_2)$$

which means that for $i = 1, 2$ one can find $y_i \in x + E^\perp$ that $y_i \in \text{relint } F_i$ and then

$$u \in P_{E^\perp}N(K, F_i) = P_{E^\perp}N(K, y_i).$$

Consider a convex polytope $K_x = K \cap (x + E^\perp)$. Lemma 2.7 implies that $\{y_1\}, \{y_2\} \in \mathcal{F}_0(K_x)$ and also

$$\begin{aligned} u &\in N_{x+E^\perp}(K_x, y_1) \cap N_{x+E^\perp}(K_x, y_2) \\ &= N_{x+E^\perp}(K_x, (y_1 + y_2)/2), \end{aligned}$$

where the last equality is due to Lemma 2.4. But again, Lemma 2.7 implies that also $\{(y_1 + y_2)/2\} \in \mathcal{F}_0(K_x)$, hence $y_1 = y_2$ (see definition of a face) and consequently, $F_1 = F(K, y_1) = F_2$.

(b) The inclusion “ \subseteq ” is obvious. For the inclusion “ \supseteq ” take arbitrary $x \in P_E K$. Put $K_x = K \cap (x + E^\perp)$. K_x is a non-empty polytope in $x + E^\perp$. By Lemma 2.3 one can find $y \in x + E^\perp$ such that $\{y\} \in \mathcal{F}_0(K_x)$ and $u \in N_{x+E^\perp}(K_x, y)$. Lemma 2.7 (or just Lemma 2.2) implies $u \in P_{E^\perp}N(K, F)$ where $F = F(K, y)$. Consequently, $F \in \tilde{\mathcal{F}}(K, E, u)$.

By the definition of $\tilde{\mathcal{F}}(K, E, u)$ and Lemma 2.7, any face $F \in \tilde{\mathcal{F}}(K, E, u)$ has dimension $\leq d$.

Finally we show that for $F \in \tilde{\mathcal{F}}_d(K, E, u)$, $P_E|F : F \rightarrow P_E F$ is an isomorphism. The condition (2.4) and Lemma 2.5 implies

$$n - d \leq \dim P_{E^\perp}N(K, F) \leq \dim N(K, F) \leq n - \dim F = n - d,$$

so $\dim P_{E^\perp}N(K, F) = \dim N(K, F) = n - d$. This means $(\text{span } N(K, F)) \cap E = \{0\}$ and $N(K, F)^\perp \cap E^\perp = \{0\}$. The definition of a normal cone yields $(\text{aff } F)_0 \subseteq N(K, F)^\perp$, which finally gives $(\text{aff } F)_0 \cap E^\perp = \{0\}$. \square

The following corollary is an immediate consequence of Proposition 2.1.

Corollary 2.8. Let $K \subseteq \mathbb{R}^n$ be a convex polytope, $E \subseteq \mathbb{R}^n$ a d -dimensional linear subspace ($1 \leq d \leq n - 1$). Then there exists a subset $\tilde{\mathcal{F}}$ of $\mathcal{F}_d(K)$ such that for any integrable function $f : E \rightarrow \mathbb{R}$,

$$\int_{P_E K} f(x) \lambda_E(dx) = \sum_{F \in \tilde{\mathcal{F}}} \frac{\text{Vol}_d(P_E F)}{\text{Vol}_d(F)} \int_F f(P_E y) \lambda_{\text{aff } F}(dy). \quad (2.6)$$

In particular (for $f \equiv 1$),

$$\text{Vol}_d(P_E K) = \sum_{F \in \tilde{\mathcal{F}}} \text{Vol}_d(P_E F).$$

Proof. Choose any $u \in E^\perp$ satisfying (2.4) for K and E and put $\tilde{\mathcal{F}} = \tilde{\mathcal{F}}_d(K, E, u)$. By Proposition 2.1,

$$\begin{aligned} \int_{P_E K} f(x) \lambda_E(dx) &= \sum_{F \in \tilde{\mathcal{F}}} \int_{P_E F} f(x) \lambda_E(dx) \\ &= \sum_{F \in \tilde{\mathcal{F}}} \frac{\text{Vol}_d(P_E F)}{\text{Vol}_d(F)} \int_F f(P_E y) \lambda_{\text{aff } F}(dy). \quad \square \end{aligned}$$

For our purposes we shall use above corollary with $f(x) = |x|^2$. In such case, the obvious inequality $|P_E y| \leq |y|$ and the identity $1 = \sum_{F \in \tilde{\mathcal{F}}} \frac{\text{Vol}_d(P_E F)}{\text{Vol}_d(P_E K)}$ lead to the following estimate: if $P_E K$ is a body of dimension d (i.e. is non-degenerated) then

$$\frac{1}{\text{Vol}_d(P_E K)} \int_{P_E K} |x|^2 \lambda_E(dx) \leq \max_{F \in \tilde{\mathcal{F}}} \frac{1}{\text{Vol}_d(F)} \int_F |y|^2 \lambda_{\text{aff } F}(dy). \quad (2.7)$$

3. Projections of the ℓ_1^n -ball and the regular simplex

First of all, we are going to see that “most” projections of B_1^n on d -dimensional subspaces ($d \leq n$) have the isotropy constant bounded. It is well known that any symmetric convex polytope in \mathbb{R}^d with $2n$ vertices is linearly equivalent to $P_E B_1^n$ for some $E \in G_{n,d}$. Indeed, if $T : \mathbb{R}^n \rightarrow \mathbb{R}^d$ is a linear transformation of full rank, then taking the d -dimensional subspace $E = (\ker T)^\perp \subseteq \mathbb{R}^n$, T can be represented as $T|_E P_E$ where $T|_E : E \rightarrow \mathbb{R}^d$ is a linear isomorphism being a restriction of T to the subspace E . As an immediate consequence we obtain the following

Lemma 3.1. *Let $K = \text{conv}\{\pm v_1, \dots, \pm v_n\} \subseteq \mathbb{R}^d$ be a symmetric convex polytope with non-empty interior and let $T : \mathbb{R}^n \rightarrow \mathbb{R}^d$ be the linear map such that $T e_i = v_i$. Then for $E = (\ker T)^\perp \in G_{n,d}$, $P_E B_1^n$ and K are linearly equivalent.*

One may also prove a similar lemma in the non-symmetric case. Recall that $\Delta_n = \text{conv}\{e_1, \dots, e_{n+1}\} \subseteq H \subseteq \mathbb{R}^{n+1}$ where H , as in the whole of this section, denotes the hyperplane orthogonal to the vector $(1, \dots, 1) \in \mathbb{R}^{n+1}$.

Lemma 3.2. *Let $K = \text{conv}\{v_1, \dots, v_{n+1}\} \subseteq \mathbb{R}^d$ ($n \geq d$) be a convex polytope with non-empty interior. Let $T : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^d$ be the linear map that $T e_i = v_i - v_0$ where $v_0 = \frac{1}{n+1} \sum_{i=1}^{n+1} v_i$ and $E = (\ker T)^\perp \subseteq \mathbb{R}^{n+1}$. Then $E \subseteq H$ is a subspace of dimension d and $K - v_0$ is linearly equivalent to $P_E \Delta_n$. Consequently, K is affinely equivalent to some orthogonal projection of the n -dimensional regular simplex S_n onto a d -dimensional subspace.*

Proof. Clearly $(1, \dots, 1) \in \ker T$, so $E \subseteq H$. Since K has non-empty interior, vectors $v_i - v_0$ span the whole of \mathbb{R}^d , so T is of full rank. Therefore $\dim E = d$ and the argument given above applies. \square

Now we can prove the following result concerning the isotropy constant of random projections of B_1^n and S_n .

Proposition 3.3. *There exist absolute constants $C, c_1, c_2 > 0$ such that the Haar probability measure of the set of subspaces $E \in G_{n,d}$ verifying*

$$L_{P_E B_1^n} < C \quad \text{and} \quad L_{P_E S_n} < C$$

is greater than $1 - c_1 e^{-c_2 \max\{\log n, d\}}$.

Proof. For small values of d , namely $d \leq c \log n$, the isotropy constant of a random projection is bounded by an absolute constant with probability greater than $1 - \frac{c_1}{n^{c_2}}$ as a consequence of Dvoretzky's theorem.

Let $G = (g_{ij})$ be a $d \times n$ Gaussian random matrix, i.e. the g_{ij} 's are i.i.d. $\mathcal{N}(0, 1)$ Gaussian random variables. Since $(\ker G)^\perp = \text{Im}(G^t) \subseteq \mathbb{R}^n$, G^t being the transpose matrix of G , and the columns of G^t are independent and rotationally invariant random vectors in \mathbb{R}^n , then a random subspace $E = (\ker G)^\perp$ has dimension d a.s. and is distributed according to the Haar probability measure μ on $G_{n,d}$. Therefore for any constant $C > 0$,

$$\mu\{E \in G_{n,d}; L_{P_E B_1^n} < C\} = \mathbb{P}\{L_{P_E B_1^n} < C\}.$$

Lemma 3.1 and the affine invariance of the isotropy constant imply $L_{P_E B_1^n} = L_{\text{conv}(\pm G e_1, \dots, \pm G e_n)}$ a.s. Klartag and Kozma proved in [12] that if C is a sufficiently large absolute constant,

$$\mathbb{P}\{L_{\text{conv}(\pm G e_1, \dots, \pm G e_n)} < C\} > 1 - c_1 e^{-c_2 d}$$

which completes the proof in the symmetric case.

For the non-symmetric case, we proceed analogously. For a $d \times (n+1)$ Gaussian random matrix $G = (g_{ij})$, take $\tilde{G} = (g_{ij} - \frac{1}{n+1} \sum_{k=1}^{n+1} g_{ik})_{i \leq d, j \leq n+1}$. Since the sum of the columns of \tilde{G} is zero, $(\ker \tilde{G})^\perp = \text{Im}(\tilde{G}^t) \subseteq H \subseteq \mathbb{R}^{n+1}$. Moreover, since rows of \tilde{G} (equivalently, columns of \tilde{G}^t) are independent canonical Gaussian random vectors in H , the random subspace $E = (\ker \tilde{G})^\perp \subseteq H$ is distributed according to the Haar probability measure on $G_{H,d}$ (Grassmann manifold of d -dimensional subspaces of H). Lemma 3.2 and the affine invariance of the isotropy constant imply $L_{P_E \Delta_n} = L_{\text{conv}(G e_1, \dots, G e_{n+1})}$ a.s.

Since $P_E \Delta_n = P_E(P_H \Delta_n)$ and $P_H \Delta_n$ is an n -dimensional regular simplex (in H), a non-symmetric counterpart of the result of Klartag and Kozma [12],

$$\mathbb{P}\{L_{\text{conv}(G e_1, \dots, G e_{n+1})} < C\} > 1 - c_1 e^{-c_2 d},$$

finishes the proof. \square

In the final part of the section we will use the tools from Section 2 to prove the main result. In particular, whenever $d \geq cn$ the boundedness of the isotropy constant holds not only for “most” projections of B_1^n and S_n but deterministically for all of them.

Theorem 3.4. Let $E \subseteq \mathbb{R}^n$ be a subspace of dimension $1 \leq d \leq n-1$ and $\mathcal{K} = P_E B_1^n$, $\mathcal{T} = P_E S_n$. Then

$$L_{\mathcal{K}}, L_{\mathcal{T}} \leq C\sqrt{n/d},$$

where $C > 0$ is a universal constant.

Proof. As an immediate consequence of (1.1),

$$L_{\mathcal{K}}^2 \leq \frac{1}{d} \frac{1}{\text{Vol}_d(\mathcal{K})^{2/d}} \frac{1}{\text{Vol}_d(\mathcal{K})} \int_{\mathcal{K}} |x|^2 \lambda_E(dx). \quad (3.8)$$

Applying (2.7), we obtain the bound

$$\frac{1}{\text{Vol}_d(\mathcal{K})} \int_{\mathcal{K}} |x|^2 \lambda_E(dx) \leq \frac{1}{\text{Vol}_d(\Delta_d)} \int_{\Delta_d} |x|^2 \lambda_{\text{aff } \Delta_d}(dy) = \frac{2}{d+2} \quad (3.9)$$

(for the last equality see e.g. [12, Lemma 2.3]). To estimate $\text{Vol}_d(\mathcal{K})$ note that $n^{-1/2} B_2^n \subseteq B_1^n$, so $n^{-1/2}(B_2^n \cap E) \subseteq P_E B_1^n$. Therefore

$$\text{Vol}_d(\mathcal{K})^{1/d} \geq \frac{c}{\sqrt{nd}}. \quad (3.10)$$

Combining these two, we get

$$L_{\mathcal{K}}^2 \leq \frac{1}{d} \frac{nd}{c^2} \frac{2}{d+2} \leq C' \frac{n}{d}.$$

In the case of the simplex it is convenient to embed E and S_n into H . More precisely, we take $S_n = \text{conv}\{P_H e_i: i = 1, \dots, n+1\} \subseteq H \subseteq \mathbb{R}^{n+1}$ and assume $E \subseteq H$. Now observe that $\mathcal{T} = P_E S_n = P_E \Delta_n$ so (2.7) again yields

$$\frac{1}{\text{Vol}_d(\mathcal{T})} \int_{\mathcal{T}} |x|^2 \lambda_E(dx) \leq \frac{2}{d+2}.$$

To bound the volume radius of \mathcal{T} from below, we use the Rogers–Shephard inequality [17]:

$$\left(\frac{2d}{d}\right)^{-1} \text{Vol}_d(\mathcal{T} - \mathcal{T}) \leq \text{Vol}_d(\mathcal{T}).$$

Note that

$$\begin{aligned} \mathcal{T} - \mathcal{T} &\supseteq \text{conv}(\mathcal{T} \cup -\mathcal{T}) = \text{conv}(P_E \Delta_n \cup -P_E \Delta_n) \\ &= P_E(\text{conv}(\Delta_n \cup -\Delta_n)) = P_E B_1^{n+1}. \end{aligned}$$

Combining with the estimate (3.10),

$$\begin{aligned}\text{Vol}_d(T)^{1/d} &\geq \left(\frac{2d}{d}\right)^{-1/d} \text{Vol}_d(P_E B_1^{n+1})^{1/d} \\ &\geq \left(\frac{2d}{d}\right)^{-1/d} \frac{c}{\sqrt{(n+1)d}} \geq \frac{c'}{\sqrt{nd}}. \quad \square\end{aligned}$$

Due to Lemma 3.2, we immediately get the following:

Corollary 3.5. *Let $K \subseteq \mathbb{R}^d$ be a non-degenerated ($\dim K = d$) convex polytope with n vertices. Then*

$$L_K \leq C \sqrt{\frac{n}{d}}.$$

4. Isotropy constant of projections of random polytopes

In this section we consider polytopes generated by the convex hull of vertices randomly chosen on the S^{n-1} . The main result is

Proposition 4.1. *There exist absolute constants C , c_1 and c_2 , such that if $m \geq n$, $\{P_i\}_{i=0}^m$ are independent random vectors on S^{n-1} and $K = \text{conv}\{\pm P_1, \dots, \pm P_m\}$ or $K = \text{conv}\{P_0, \dots, P_m\}$, then*

$$\mathbb{P}\left\{L_{P_E K} \leq C \sqrt{\frac{n}{d}}, \forall E \in G_{n,d}, \forall 1 \leq d \leq n-1\right\} \geq 1 - c_1 e^{-c_2 n}.$$

The proof follows [1]. We shall only sketch the main ideas as the technical computations can be found in that reference.

Sketch of the proof. Let $E \subseteq \mathbb{R}^n$ denotes a d -dimensional subspace. The ideas in what follows will give us the proof for $m \geq cn$ with an absolute constant c . If $m < cn$, Corollary 3.5 gives deterministically $L_{P_E K} \leq C \sqrt{\frac{m}{d}} \leq C' \sqrt{\frac{n}{d}}$.

Apply once again (1.1). Writing $r(K)$ the inradius of K and using the inequality (2.7), the main consequence of Proposition 2.1, we obtain that for any polytope $K \subset \mathbb{R}^n$ and any d -dimensional subspace E ,

$$L_{P_E K}^2 \leq \frac{C}{r(K)^2} \max_{F \in \mathcal{F}_d(K)} \frac{1}{\text{Vol}_d(F)} \int_F |x|^2 \lambda_{\text{aff } F}(dx).$$

When K is the symmetric convex hull of m independent random points in S^{n-1} , it was proved in [1, Lemma 3.1], that for some constant c such that $cn \leq m \leq ne^{\frac{n}{2}}$,

$$\mathbb{P}\left\{r(K) < \frac{1}{2\sqrt{2}} \sqrt{\frac{\log \frac{m}{n}}{n}}\right\} \leq e^{-n}.$$

The same proof gives the statement in the non-symmetric case.

On the other hand, with probability 1, each d -dimensional face of K is a simplex $F = \text{conv}\{Q_1, \dots, Q_{d+1}\}$ with $Q_i = \varepsilon_i P_{j_i}$ (or just $Q_i = P_{j_i}$ in the non-symmetric case) where $1 \leq j_1 < \dots < j_{d+1} \leq m$ and $\varepsilon_i \in \{-1, 1\}$. The same proof as in [1,12] shows that with probability 1 we have

$$\frac{1}{\text{Vol}_d(F)} \int_F |x|^2 \lambda_{\text{aff } F}(dx) = \frac{2}{d+2} + \frac{1}{(d+1)(d+2)} \sum_{i_1 \neq i_2}^{d+1} \langle Q_{i_1}, Q_{i_2} \rangle. \quad (4.11)$$

In order to give a bound for this quantity for a fixed $F \in \mathcal{F}_d(K)$ we proceed in the same way as in [1, Theorem 3.1], by using a version of Bernstein's inequality as stated in [4]. We thus obtain

$$\mathbb{P} \left\{ \sum_{i_1 \neq i_2}^{d+1} \langle Q_{i_1}, Q_{i_2} \rangle > \epsilon(d+1) \right\} \leq 2e^{-c\epsilon n} \quad (4.12)$$

for every $\epsilon > \epsilon_0$, where ϵ_0 is an absolute constant.

Now, for each $F \in \mathcal{F}_d(K)$ let Q_1^F, \dots, Q_{d+1}^F be vertices of F . Applying (4.12) and the union bound over $\mathcal{F}_d(K)$ (whose cardinality is clearly bounded by $\binom{2m}{d+1}$), we obtain for $\epsilon \log \frac{m}{n} > \epsilon_0$,

$$\begin{aligned} & \mathbb{P} \left\{ \max_{F \in \mathcal{F}_d(K)} \sum_{i_1 \neq i_2}^{d+1} \langle Q_{i_1}^F, Q_{i_2}^F \rangle > \epsilon(d+1) \log \frac{m}{n} \right\} \\ & \leq \binom{2m}{d+1} 2e^{-c\epsilon n \log \frac{m}{n}} \leq 2e^{-c\epsilon n \log \frac{m}{n} + (d+1) \log \frac{2em}{d+1}} \\ & \leq 2e^{-c\epsilon n \log \frac{m}{n} + n \log \frac{2em}{n}}, \end{aligned}$$

since the function $x \log \frac{C}{x}$ is increasing when $\frac{C}{x} > e$. Consequently, by the union bound over d ,

$$\begin{aligned} & \mathbb{P} \left\{ \exists 1 \leq d \leq n-1 \text{ s.t. } \max_{F \in \mathcal{F}_d(K)} \sum_{i_1 \neq i_2}^{d+1} \langle Q_{i_1}^F, Q_{i_2}^F \rangle > \epsilon(d+1) \log \frac{m}{n} \right\} \\ & \leq 2e^{-c\epsilon n \log \frac{m}{n} + n \log \frac{2em}{n} + \log n}. \end{aligned}$$

Since $m \geq cn$, considering the complement set and using (4.11), we can fix $\epsilon > 0$ a large enough numerical constant to obtain

$$\mathbb{P} \left\{ \forall 1 \leq d \leq n-1, \max_{F \in \mathcal{F}_d(K)} \frac{1}{\text{Vol}_d(F)} \int_F |x|^2 \lambda_{\text{aff } F}(dx) \leq \frac{C}{d} \log \frac{m}{n} \right\} \geq 1 - 2e^{-cn \log \frac{m}{n}}.$$

Thus, there exist constants $c, C > 0$ such that if $cn \leq m \leq ne^{\frac{n}{2}}$ then the set of points (P_1, \dots, P_m) for which the inequality $L_{P_E K} \leq C\sqrt{\frac{n}{d}}$ holds for every d -dimensional subspace E and for every $1 \leq d \leq n-1$ has probability greater than $1 - 2e^{-cn \log \frac{m}{n}} - e^{-n} > 1 - c_1 e^{-c_2 n}$.

In case $m > ne^{\frac{n}{2}}$, for n large enough, $r(K) \geq \frac{1}{4}$ with probability greater than $1 - e^{-n}$ so with this probability

$$dL_{P_E K}^2 \leq \frac{1}{\text{Vol}_d(P_E K)^{\frac{2}{d}}} \frac{1}{\text{Vol}_d(P_E K)} \int_{P_E K} |x|^2 dx \leq \frac{1}{\text{Vol}_d(\frac{1}{4}B_2^d)^{\frac{2}{d}}} \leq cd$$

and the proof is complete. \square

5. A general result

In this section we prove a general relation between the isotropy constant of the hyperplane projections of an isotropic convex body and of the body itself.

Corollary 5.1. *Let K be an isotropic convex body and let H be a hyperplane. Then*

$$L_{P_H K} \sim L_K.$$

Its proof relies on the next proposition which improves the numerical constants appearing in a more general statement in [5] for the case of projections onto hyperplanes.

Proposition 5.2. *Let $K \subset \mathbb{R}^n$ be an isotropic convex body and let $H = v^\perp$ be a hyperplane. If $S(K)$ is the Steiner symmetrization of K with respect to H then,*

$$\left(1 - c \frac{\log n}{n}\right) L_K \leq L_{S(K)} \leq L_K$$

for some numerical constant $c > 0$.

Proof. Without loss of generality, we may assume that $v = e_n = (0, \dots, 0, 1)$. Write $E = \langle e_n \rangle$ for the 1-dimensional subspace generated by e_n . The Steiner symmetrization of K is defined by

$$S(K) := \left\{ (y, t) \in \mathbb{R}^{n-1} \times \mathbb{R}; y \in P_H K, |t| \leq \frac{1}{2} \text{Vol}_1(K \cap (x + E)) \right\}.$$

Clearly $P_H K = P_H S(K) = S(K) \cap H$.

Now we study the inertia matrix of $S(K)$. First notice that for $x \in P_H K$, $\text{Vol}_1(K \cap (x + E)) = \text{Vol}_1(S(K) \cap (x + E))$. For every $\theta \in S^{n-1} \cap H$, Fubini's theorem yields

$$\begin{aligned} \int_{S(K)} \langle x, \theta \rangle^2 dx &= \int_{P_H K} \int_{S(K) \cap (x+E)} \langle y + te_n, \theta \rangle^2 dt dy \\ &= \int_{P_H K} \langle y, \theta \rangle^2 \text{Vol}_1(K \cap (x + E)) dy = L_K^2. \end{aligned}$$

Using the fact that $\int_{S(K) \cap (x+E)} t dt = 0$ for $x \in P_H K$, in the similar fashion we show that for every $\theta \in S^{n-1} \cap H$,

$$\int_{S(K)} \langle x, \theta \rangle \langle x, e_n \rangle dx = 0.$$

Also $\int_{S(K) \cap (x+E)} t^2 dt \leq \int_{K \cap (x+E)} t^2 dt$ for $x \in P_H K$, thus

$$\int_{S(K)} \langle x, e_n \rangle^2 dx \leq \int_K \langle x, e_n \rangle^2 dx = L_K^2.$$

Taking $\sigma > 0$ such that $\sigma^2 := \int_{S(K)} \langle x, e_n \rangle^2 dx / L_K^2$, we obtain that the inertia matrix of $S(K)$ is

$$M = L_K^2 \begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & 1 & \\ & & & \sigma^2 \end{pmatrix}.$$

The volume of $S(K)$ is 1, so $L_{S(K)} = (\det M)^{1/2n}$ (see [15]), which means

$$L_{S(K)} = \sigma^{1/n} L_K = \left(\frac{(\int_{S(K)} \langle x, e_n \rangle^2 dx)^{1/2}}{L_K} \right)^{1/n} L_K. \quad (5.13)$$

Since $\sigma \leq 1$, we obtain $L_{S(K)} \leq L_K$.

A well-known fact due to Hensley [8] states that $\text{Vol}_{n-1}(K_1 \cap H) \sim (\int_{K_1} \langle x, e_n \rangle^2 dx)^{-1/2}$ for any convex body K_1 with volume 1 and center of mass at the origin. Using this fact for $S(K)$ in (5.13) we obtain that for some absolute constant $c > 0$,

$$L_{S(K)} \geq \left(\frac{c}{\text{Vol}_{n-1}(S(K) \cap H) L_K} \right)^{1/n} L_K = \left(\frac{c}{\text{Vol}_{n-1}(P_H K) L_K} \right)^{1/n} L_K.$$

Now we use the following inequality:

$$\frac{1}{n} \text{Vol}_{n-1}(P_H K) \text{Vol}_1(K \cap E) \leq \text{Vol}_n(K) = 1$$

(for the proof, see for instance [16, Lemma 8.8] which works in the non-symmetric case). Hence

$$\text{Vol}_{n-1}(P_H K) \leq \frac{n}{\text{Vol}_1(K \cap E)} \leq \frac{n}{2r(K)},$$

where $r(K)$ is the inradius of K .

Since every isotropic convex body verifies $r(K) \geq L_K$ (see [7] or [11], for instance) we obtain

$$L_{S(K)} \geq \left(\frac{2c}{n} \right)^{1/n} L_K. \quad \square$$

Proof of Corollary 5.1. Since $P_H K = S(K) \cap H$ we have

$$L_{P_H K} = L_{S(K) \cap H} \sim L_{S(K)} \sim L_K,$$

where the first equivalence is the corresponding one for sections of convex bodies as proved in [15]. \square

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