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ANGLES OF THE GAUSSIAN SIMPLEX

ABSTRACT. Consider a d -dimensional simplex whose vertices are random points chosen independently according to the standard Gaussian distribution on \mathbb{R}^d . We prove that the expected angle sum of this random simplex equals the angle sum of the regular simplex of the same dimension d .

§1. MAIN RESULT

The sum of measures of angles in any triangle in the Euclidean plane is constant. However, a similar statement is not true in higher dimensions. The sum of solid angles of a d -dimensional simplex, where $d \geq 3$, can take any value between 0 and $1/2$ of the full solid angle, as will be shown in Proposition 5.4. Thus, it is natural to ask about the “average value” of the sum of solid angles of the simplex. Of course, the notion of “average” depends on the probability measure we put on the set of all simplices. In the present paper, we consider the *Gaussian simplex*, i.e., a random simplex in \mathbb{R}^d whose vertices X_0, \dots, X_d are chosen independently according to the standard Gaussian distribution on \mathbb{R}^d . Our main result is the following

Theorem 1.1. *The expected sum of the solid angles of the Gaussian simplex coincides with the sum of the solid angles of the regular simplex of the same dimension.*

Let us mention some related results. Feldman and Klain [2] showed that in every tetrahedron, the sum of solid angles, measured in steradians and divided by 2π , gives the probability that a random projection of the tetrahedron onto a uniformly chosen two-dimensional plane is a triangle. They also obtained a generalization of this result to simplices of arbitrary dimension. The probability that a random Gaussian tetrahedron is acute, as well as the distribution of its solid angles, is discussed in the papers of Finch [3] and Bosetto [1]. It seems that the angles of the Gaussian simplex in dimension $d \geq 4$ were not studied so far. An explicit formula for the

Key words and phrases: convex hull, Gaussian simplex, regular simplex, solid angle, random polytope, convex cone.

The work of the second author is supported by the grant RFBR 16-01-00367.

solid angles of the regular d -dimensional simplex is known and can be found in [7, 10, 6]. We shall not rely on this formula. Expected angles of the so-called beta simplices (which contain Gaussian simplices as a limiting case) were used in [5] to compute the expected f -vectors of beta polytopes, but no formula for the expected angles was given there.

The paper is organized as follows. After recalling some necessary facts from convex and stochastic geometry in Section 2, we shall present two different proofs of Theorem 1.1 in Sections 3 and 4. Section 5 contains some auxiliary (and probably known) results.

§2. FACTS FROM CONVEX AND STOCHASTIC GEOMETRY

For vectors $v_1, \dots, v_n \in \mathbb{R}^d$, we define their *positive* or *conic hull* as

$$\text{pos}(v_1, \dots, v_n) := \left\{ \sum_{i=1}^n \lambda_i v_i : \lambda_1, \dots, \lambda_n \geq 0 \right\}.$$

A set $C \subset \mathbb{R}^d$ is said to be a *polyhedral cone* (or just a *cone*) if it can be represented as a positive hull of finitely many vectors. The *solid angle* of the cone C is defined as

$$\alpha_d(C) := \mathbb{P}[Z \in C], \quad (1)$$

where Z is uniformly distributed on the unit sphere in \mathbb{R}^d . The maximal possible value of the solid angle in this normalization is $\alpha_d(\mathbb{R}^d) = 1$. If $C \neq \mathbb{R}^d$, then $\mathbb{P}[Z \in C, -Z \in C] = 0$ and (1) is equivalent to

$$\alpha_d(C) = \frac{1}{2} \mathbb{P}[W_1 \cap C \neq \{0\}], \quad (2)$$

where W_1 denotes the line passing through Z and $-Z$. Equivalently, W_1 is a random 1-dimensional linear subspace in \mathbb{R}^d uniformly chosen with respect to the Haar measure.

Let $\text{lin}(C)$ be the *linear hull* of C , i.e. the minimal linear subspace containing C . The dimension of the cone C , denoted by $\dim C$, is defined as the dimension of $\text{lin}(C)$. If $\dim C = k < d$, then, by definition, $\alpha_d(C) = 0$. However, similarly to (1), we can define $\alpha_k(C)$ as the solid angle of C measured with respect to the linear hull of C , which is isomorphic to \mathbb{R}^k . Namely, we define $\alpha_k(C) := \mathbb{P}[Z' \in C]$, where Z' is uniformly distributed on the unit sphere in the linear hull of C .

If $\dim C = k$ and C is not a k -dimensional linear subspace, the conic Crofton formula (see, e.g., [9, Eq. (6.63)]) implies the following generalization of (2):

$$\alpha_k(C) := \frac{1}{2} \mathbb{P}[W_{d-k+1} \cap C \neq \{0\}], \quad (3)$$

where W_{d-k+1} denotes a random $(d-k+1)$ -dimensional linear subspace in \mathbb{R}^d uniformly chosen with respect to the Haar measure. Alternatively, we can observe that $W_{d-k+1} \cap \text{lin}(C)$ is a random one-dimensional linear subspace of $\text{lin}(C)$ distributed uniformly on the set of all such subspaces, so that (3) follows from (2) applied to $\text{lin}(C)$ as the ambient space.

Let x_0, \dots, x_d be $d+1$ points in \mathbb{R}^n , where $n \geq d$, such that the affine subspace spanned by these points has dimension d . A *simplex* S with vertices at x_0, \dots, x_d is defined as the convex hull of these points, that is,

$$S := \text{conv}(x_0, \dots, x_d) := \left\{ \sum_{i=0}^d \lambda_i x_i : \lambda_0, \dots, \lambda_d \geq 0, \sum_{i=0}^d \lambda_i = 1 \right\}.$$

We say that the dimension of S is d . Define the *solid angle* of S at x_i as

$$\alpha_d(S, x_i) := \alpha_d(\text{pos}(x_0 - x_i, x_1 - x_i, \dots, x_d - x_i)).$$

The sum of the solid angles of S is denoted by

$$\gamma_d(S) := \sum_{i=0}^d \alpha_d(S, x_i). \quad (4)$$

A simplex is called *regular* if the pairwise distances between its vertices are all equal. We shall use the following convenient form of the regular d -dimensional simplex in \mathbb{R}^{d+1} :

$$T^d := \text{conv}(e_0, \dots, e_d),$$

where e_0, \dots, e_d is the standard orthonormal basis in \mathbb{R}^{d+1} .

We shall be interested in random simplices defined as follows. Let X_0, \dots, X_d be independent random points with standard Gaussian distribution on \mathbb{R}^d . The Lebesgue density of any of the X_i 's is thus given by

$$f(x) = (2\pi)^{-d/2} e^{-|x|^2/2},$$

where $|x|$ is the Euclidean norm of $x \in \mathbb{R}^d$. The d -dimensional *Gaussian simplex* is defined as the convex hull of X_0, \dots, X_d :

$$\mathcal{P}_d := \text{conv}(X_0, \dots, X_d).$$

With this notation, we can restate our main result as follows:

Theorem 2.1. *We have $\mathbb{E} \gamma_d(\mathcal{P}_d) = \gamma_d(T^d)$.*

Since the family (X_0, \dots, X_d) is exchangeable and all solid angles of the regular simplex are equal, an equivalent formulation of the theorem is as follows:

$$\mathbb{E} \alpha_d(\mathcal{P}_d, X_0) = \alpha_d(T^d, e_0). \quad (5)$$

In the next two sections, we give two different proofs of (5).

§3. PROOF I: LIFTING THE DIMENSION

The main idea is to represent the d -dimensional Gaussian simplex in \mathbb{R}^d as a projection of a d -dimensional Gaussian simplex in \mathbb{R}^n and then let n go to infinity. We shall show that the expected solid angles of both simplices are equal and there is a “freezing phenomenon”: In the large n limit, the d -dimensional Gaussian simplex in \mathbb{R}^n converges to the regular one.

Consider $d+1$ independent sequences of independent standard Gaussian variables (constructed on the same probability space):

$$\begin{aligned} &N_{01}, N_{02}, \dots, N_{0n}, \dots, \\ &N_{11}, N_{12}, \dots, N_{1n}, \dots, \\ &\quad \dots, \\ &N_{d1}, N_{d2}, \dots, N_{dn}, \dots \end{aligned}$$

For all $n \in \mathbb{N}$ and $k = 0, \dots, d$, let $X_k^{(n)}$ be a standard Gaussian vector in \mathbb{R}^n formed by the first n variables of the k th sequence:

$$X_k^{(n)} := (N_{k1}, \dots, N_{kn})^\top,$$

where $^\top$ stands for transpose. For $n \geq d$, the convex hull

$$\mathcal{P}_d^{(n)} := \text{conv}(X_0^{(n)}, \dots, X_d^{(n)})$$

is a d -dimensional simplex in \mathbb{R}^n , with probability one. In particular, $\mathcal{P}_d^{(d)}$ is equidistributed with \mathcal{P}_d . We now show that the expected solid angles of $\mathcal{P}_d^{(n)}$ and \mathcal{P}_d are equal.

Lemma 3.1. *For all $n \geq d$, we have*

$$\mathbb{E} \alpha_d(\mathcal{P}_d^{(n)}, X_0^{(n)}) = \mathbb{E} \alpha_d(\mathcal{P}_d, X_0).$$

Proof. By (3),

$$\begin{aligned}\mathbb{E} \alpha_d(\mathcal{P}_d^{(n)}, X_0^{(n)}) &= \mathbb{E} \alpha_d(\text{pos}(X_1^{(n)} - X_0^{(n)}, \dots, X_d^{(n)} - X_0^{(n)})) \\ &= \frac{1}{2} \mathbb{P}[W_{n-d+1} \cap \text{pos}(X_1^{(n)} - X_0^{(n)}, \dots, X_d^{(n)} - X_0^{(n)}) \neq \{0\}],\end{aligned}$$

where W_{n-d+1} is the random $(n-d+1)$ -dimensional linear subspace of \mathbb{R}^n distributed uniformly on the set of all such subspaces and independent of everything else. Let e_1, \dots, e_n denote the standard orthonormal basis in \mathbb{R}^n . Since the standard Gaussian distribution is rotationally invariant, we can replace W_{n-d+1} by $\text{lin}(e_d, \dots, e_n)$, the linear hull of e_d, \dots, e_n :

$$\begin{aligned}\mathbb{E} \alpha_d(\mathcal{P}_d^{(n)}, X_0^{(n)}) &= \frac{1}{2} \mathbb{P}[\text{lin}(e_d, \dots, e_n) \cap \text{pos}(X_1^{(n)} - X_0^{(n)}, \dots, X_d^{(n)} - X_0^{(n)}) \neq \{0\}].\end{aligned}$$

The next observation is that

$$\text{lin}(e_d, \dots, e_n) \cap \text{pos}(X_1^{(n)} - X_0^{(n)}, \dots, X_d^{(n)} - X_0^{(n)}) \neq \{0\}$$

if and only if the convex hull of the orthogonal projection of $X_1^{(n)} - X_0^{(n)}, \dots, X_d^{(n)} - X_0^{(n)}$ on $\text{lin}(e_d, \dots, e_n)^\perp = \text{lin}(e_1, \dots, e_{d-1})$ contains the origin. By definition, the orthogonal projection of $X_k^{(n)}$ on $\text{lin}(e_1, \dots, e_{d-1})$ is $X_k^{(d-1)}$. Therefore,

$$\mathbb{E} \alpha_d(\mathcal{P}_d^{(n)}, X_0^{(n)}) = \frac{1}{2} \mathbb{P}[0 \in \text{conv}(X_1^{(d-1)} - X_0^{(d-1)}, \dots, X_d^{(d-1)} - X_0^{(d-1)})]. \quad (6)$$

This relation holds for all $n \geq d$ and the right-hand side does not depend on n . Thus we have

$$\mathbb{E} \alpha_d(\mathcal{P}_d^{(n)}, X_0^{(n)}) = \mathbb{E} \alpha_d(\mathcal{P}_d^{(d)}, X_0^{(d)}) = \mathbb{E} \alpha_d(\mathcal{P}_d, X_0),$$

which proves the lemma. \square

To complete the proof of (5), we let n tend to infinity. The strong law of large numbers implies that we have

$$\lim_{n \rightarrow \infty} \frac{\langle X_i^{(n)}, X_j^{(n)} \rangle}{n} = 0 \quad \text{and} \quad \lim_{n \rightarrow \infty} \frac{\langle X_i^{(n)}, X_i^{(n)} \rangle}{n} = 1 \quad \text{almost surely (a.s.)}$$

whenever $0 \leq i < j \leq d$, whence it follows that

$$\lim_{n \rightarrow \infty} \frac{\langle X_i^{(n)} - X_0^{(n)}, X_j^{(n)} - X_0^{(n)} \rangle}{|X_i^{(n)} - X_0^{(n)}| |X_j^{(n)} - X_0^{(n)}|} = 1/2 \quad \text{a.s.}$$

whenever $1 \leq i < j \leq d$. On the other hand, for the regular simplex $T^d = \text{conv}(e_0, \dots, e_d)$ we have

$$\frac{\langle e_i - e_0, e_j - e_0 \rangle}{|e_i - e_0||e_j - e_0|} = 1/2.$$

By Corollary 5.2 and Remark 5.3 stated below, this yields the convergence of the corresponding solid angles:

$$\lim_{n \rightarrow \infty} \alpha_d(\mathcal{P}_d^{(n)}, X_0^{(n)}) = \alpha_d(T^d, e_0) \quad \text{a.s.}$$

Since the solid angle is bounded by 1, the dominated convergence theorem implies that

$$\lim_{n \rightarrow \infty} \mathbb{E} \alpha_d(\mathcal{P}_d^{(n)}, X_0^{(n)}) = \alpha_d(T^d, e_0).$$

Applying Lemma 3.1 completes the proof.

§4. PROOF II: PROJECTION

The starting point of our second proof of Theorem 1.1 is the identity

$$\mathbb{E} \alpha_d(\mathcal{P}_d, X_0) = \frac{1}{2} \mathbb{P}[0 \in \text{conv}(Y_1 - Y_0, \dots, Y_d - Y_0)], \quad (7)$$

where Y_0, \dots, Y_d are independent standard Gaussian vectors in \mathbb{R}^{d-1} . Even though this identity follows from (6), we provide an independent argument. With probability one, the cone $\text{pos}(X_1 - X_0, \dots, X_d - X_0)$ is of full dimension d and does not coincide with \mathbb{R}^d . Therefore, by (2),

$$\mathbb{E} \alpha_d(\mathcal{P}_d, X_0) = \frac{1}{2} \mathbb{P}[W_1 \cap \text{pos}(X_1 - X_0, \dots, X_d - X_0) \neq \{0\}],$$

where W_1 is a uniformly distributed one-dimensional linear subspace of \mathbb{R}^d which is independent of X_0, \dots, X_d . By rotational invariance, we can replace W_1 by the line $\text{lin}(e)$, where $e \in \mathbb{R}^d$ is any unit vector. Let Y_0, \dots, Y_d be the projections of X_0, \dots, X_d on the orthogonal complement of e (which we identify with \mathbb{R}^{d-1}). The key observation is that

$$\begin{aligned} \text{lin}(e) \cap \text{pos}(X_1 - X_0, \dots, X_d - X_0) \neq \{0\} & \quad \text{if and only if} \\ & \quad 0 \in \text{conv}(Y_1 - Y_0, \dots, Y_d - Y_0). \end{aligned}$$

The proof of (7) is complete.

Let us now look at the right-hand side of (7). Observe that $\text{conv}(Y_1 - Y_0, \dots, Y_d - Y_0)$ contains 0 if and only if there exist $\lambda_1, \dots, \lambda_d \geq 0$ with $\lambda_1 + \dots + \lambda_d > 0$ such that

$$\lambda_1(Y_1 - Y_0) + \dots + \lambda_d(Y_d - Y_0) = 0,$$

or equivalently,

$$(-\lambda_1 - \dots - \lambda_d)Y_0 + \lambda_1 Y_1 + \dots + \lambda_d Y_d = 0. \quad (8)$$

We denote by Y the $(d-1) \times (d+1)$ -matrix whose columns are Y_0, \dots, Y_d :

$$Y := (Y_0, \dots, Y_d).$$

Then Condition (8) is equivalent to the following one:

$$Y \begin{pmatrix} -\lambda_1 - \dots - \lambda_d \\ \lambda_1 \\ \dots \\ \lambda_d \end{pmatrix} = 0 \quad \text{or} \quad \begin{pmatrix} -\lambda_1 - \dots - \lambda_d \\ \lambda_1 \\ \dots \\ \lambda_d \end{pmatrix} \in \ker Y. \quad (9)$$

Now, let C be the cone in \mathbb{R}^{d+1} defined as

$$C := \text{pos}(e_1 - e_0, \dots, e_d - e_0),$$

where e_0, \dots, e_d is the standard orthonormal basis in \mathbb{R}^{d+1} . By definition, we have

$$\alpha_d(C) = \alpha_d(T^d, e_0). \quad (10)$$

On the other hand, it is obvious that

$$C = \{(-\lambda_1 - \dots - \lambda_d, \lambda_1, \dots, \lambda_d) \in \mathbb{R}^{d+1} : \lambda_1, \dots, \lambda_d \geq 0\}.$$

Therefore, the condition that there exist $\lambda_1, \dots, \lambda_d \geq 0$ with $\lambda_1 + \dots + \lambda_d > 0$ such that (9) holds is equivalent to the condition

$$C \cap \ker Y \neq \{0\}.$$

This implies that

$$\mathbb{E} \alpha_d(\mathcal{P}_d, X_0) = \frac{1}{2} \mathbb{P}[C \cap \ker Y \neq \{0\}].$$

By definition, Y is a $(d-1) \times (d+1)$ matrix whose entries are independent standard Gaussian variables. Thus, with probability one, $\ker Y$ is a 2-dimensional linear subspace in \mathbb{R}^{d+1} and it is uniformly distributed with respect to the Haar measure on the set of all 2-dimensional subspaces in \mathbb{R}^{d+1} . Recall that $\text{lin}(C)$ denotes the minimal linear subspace containing C . Since $\dim C = d$, we have that $W'_1 := \ker Y \cap \text{lin}(C)$ is uniformly

distributed on the set of all 1-dimensional linear subspaces in $\text{lin}(C)$ with respect to the Haar measure. Therefore, we have

$$\begin{aligned}\mathbb{E} \alpha_d(\mathcal{P}_d, X_0) &= \frac{1}{2} \mathbb{P}[C \cap \ker Y \neq \{0\}] = \frac{1}{2} \mathbb{P}[C \cap (\text{lin}(C) \cap \ker Y) \neq \{0\}] \\ &= \frac{1}{2} \mathbb{P}[C \cap W'_1 \neq \{0\}] = \alpha_d(C),\end{aligned}$$

see (2) for the last equality. Together with (10), this completes the proof of (5).

§5. APPENDIX

5.1. Formula for the solid angle of a simplicial cone. Since we could not find a proper reference for the following statement, we present its proof. This proof was obtained jointly with Anna Gusakova.

Proposition 5.1. *Let v_1, \dots, v_d be linearly independent vectors in \mathbb{R}^d . Then the solid angle of the cone $C := \text{pos}(v_1, \dots, v_d)$ is given by*

$$\alpha_d(C) = \frac{\sqrt{\det \Gamma}}{(2\pi)^{d/2}} \int_{\mathbb{R}_+^d} \exp\left(-\frac{1}{2}\langle x, \Gamma x \rangle\right) dx,$$

where Γ is the Gram matrix of v_1, \dots, v_d .

Proof. Let V be the $d \times d$ -matrix whose columns are v_1, \dots, v_d . Let V_{ij} denote the (i, j) -minor of V obtained by eliminating the i th row and the j th column. For $k = 1, \dots, d$, let n_k be the vector defined by

$$n_k := \frac{1}{\det V} \sum_{i=1}^d (-1)^{k+i} (\det V_{ik}) e_i,$$

where e_1, \dots, e_d is the standard orthonormal basis in \mathbb{R}^d . We shall compute the Gram matrix of n_1, \dots, n_d and show that C has the following representation:

$$C = \{x \in \mathbb{R}^d : \langle n_k, x \rangle \geq 0 \text{ for all } k = 1, \dots, d\}. \quad (11)$$

By definition of n_k , we have

$$\langle n_k, n_l \rangle = \frac{1}{(\det V)^2} \sum_{i=1}^d (-1)^{k+i} \det V_{ik} \det V_{il}.$$

The well-known formula for the inverse of a matrix, namely

$$V^{-1} = \frac{1}{\det V} ((-1)^{i+j} \det V_{ji})_{i,j=1}^d,$$

yields the Gram matrix of n_1, \dots, n_d :

$$\Sigma := (\langle n_k, n_l \rangle)_{k,l=1}^d = (V^T V)^{-1} = \Gamma^{-1}. \quad (12)$$

Now, let us prove (11). For a vector $x \in \mathbb{R}^d$, let $V_k(x)$ be the matrix with columns $v_1, \dots, v_{k-1}, x, v_{k+1}, \dots, v_d$. By the Laplace formula for the determinant, we have

$$\langle n_k, x \rangle = \frac{\det V_k(x)}{\det V}.$$

Taking $x = v_i$ gives

$$\langle n_k, v_i \rangle = \begin{cases} 1, & \text{if } k = i, \\ 0, & \text{if } k \neq i. \end{cases}$$

Therefore, the cones spanned by v_1, \dots, v_d and $-n_1, \dots, -n_d$ are polar to each other, so that we have

$$C = \{x \in \mathbb{R}^d : \langle n_k, x \rangle \geq 0 \text{ for all } k = 1, \dots, d\}.$$

Since the standard Gaussian distribution is rotationally invariant, Definition (1) is equivalent to

$$\alpha_d(C) = \mathbb{P}[X \in C],$$

where X is a standard Gaussian vector in \mathbb{R}^d . The last two equations imply that

$$\alpha_d(C) = \mathbb{P}[\langle n_k, X \rangle \geq 0 \text{ for all } k = 1, \dots, d].$$

The random vector $(\langle n_1, X \rangle, \dots, \langle n_d, X \rangle)$ is centered Gaussian with covariance matrix $\Sigma = \Gamma^{-1}$ given by (12) because

$$\mathbb{E}[\langle n_k, X \rangle \langle n_l, X \rangle] = \langle n_k, n_l \rangle.$$

Using the formula for its density function completes the proof. \square

Corollary 5.2. *Let $(v_{nk})_{n=0,1,\dots; k=1,\dots,d}$ be a family of vectors in \mathbb{R}^d such that for each n , the vectors v_{n1}, \dots, v_{nd} are linearly independent. If $n \in \{0, 1, \dots\}$, let C_n denote the cone $\text{pos}(v_{n1}, \dots, v_{nd})$. If for all i, j with $1 \leq i < j \leq d$ we have*

$$\lim_{n \rightarrow \infty} \frac{\langle v_{ni}, v_{nj} \rangle}{|v_{ni}| |v_{nj}|} = \frac{\langle v_{0i}, v_{0j} \rangle}{|v_{0i}| |v_{0j}|},$$

then

$$\lim_{n \rightarrow \infty} \alpha_d(C_n) = \alpha_d(C_0).$$

Proof. Since the replacement of v_{ni} by $v_{ni}/|v_{ni}|$ does not change the solid angles, the statement readily follows from Proposition 5.1 and the dominated convergence theorem. \square

Remark 5.3. Although we stated Corollary 5.2 for cones of full dimension, the result also holds for d -dimensional cones of the form $C_n = \text{pos}(v_{n1}, \dots, v_{nd})$ in $\mathbb{R}^{m(n)}$ with $m(n) \geq d$. Indeed, the solid angles $\alpha_d(C_n)$ depend on the Gram matrix only and do not depend on the ambient space.

5.2. Bounds on the sum of the solid angles of a simplex. A simplex is called nondegenerate if its interior is nonempty.

Proposition 5.4. *For each nondegenerate simplex $S \subset \mathbb{R}^d$ with $d \geq 3$, we have*

$$0 < \gamma_d(S) < \frac{1}{2}. \quad (13)$$

Moreover, for each $h \in (0, 1/2)$ there exists a nondegenerate simplex S such that $\gamma_d(S) = h$.

This fact must be well-known, but we could not find a proper reference. For the reader's convenience, we present a proof here. The idea of the proof is due to Sergei Ivanov [4].

Proof. First, we show that (13) holds. The lower bound on $\gamma_d(S)$ is trivial. Let us prove the upper one.

Any nondegenerate d -dimensional simplex is the intersection of $d + 1$ closed half-spaces in \mathbb{R}^d . In our case, this means that there exist vectors $y_0, \dots, y_d \in \mathbb{R}^d$ and closed half-spaces H_0^+, \dots, H_d^+ with boundaries H_0, \dots, H_d passing through the origin such that

$$S = \bigcap_{i=0}^d (y_i + H_i^+).$$

For $k = 0, \dots, d$, we denote by H_k^- the half-space complementary to H_k^+ , that is, the closure of $\mathbb{R}^d \setminus H_k^+$.

Since S is nondegenerate, it follows that the linear hyperplanes H_0, \dots, H_d are in general position, that is, any d of them have linearly independent

normal vectors. By Schläfli's formula [8], the hyperplanes divide \mathbb{R}^d into m polyhedral cones D_1, \dots, D_m , where

$$m = 2^{d+1} - 2. \quad (14)$$

By construction, we have

$$\text{Int } D_l \cap \text{Int } D_{l'} = \emptyset \quad \text{for } l \neq l', \quad (15)$$

where $\text{Int } D_l$ denotes the interior of D_l . Therefore,

$$\sum_{l=1}^m \alpha_d(D_l) = 1. \quad (16)$$

We introduce the following notation:

$$D^\epsilon := \bigcap_{i=0}^d H_i^{\epsilon_i}, \quad \text{where } \epsilon = (\epsilon_0, \dots, \epsilon_d) \in \{+, -\}^{d+1}.$$

For each $l \in \{1, \dots, m\}$, we have $D_l = D^\epsilon$ for some $\epsilon \in \{+, -\}^{d+1}$. Besides, if $D^\epsilon \neq \{0\}$ for some $\epsilon \in \{+, -\}^{d+1}$, then $D^\epsilon = D_l$ for some l .

For $k = 0, \dots, d$, we denote by x_k the vertex of S opposite to the face contained in $y_k + H_k$, and let C_k be the internal cone at x_k :

$$C_k := \text{pos}(x_0 - x_k, \dots, x_d - x_k).$$

In terms of the half-spaces, C_k is represented as follows:

$$C_k = \bigcap_{i:i \neq k} H_i^+.$$

Since $C_k \subset H_k^-$, we have

$$C_k = D^{(+, \dots, +, -, +, \dots, +)},$$

where the k th entry of the upper index is “−” and all of its other entries are “+”. Similarly, we have

$$-C_k = D^{(-, \dots, -, +, -, \dots, -)}.$$

Thus it follows from (4) and (16) that

$$\gamma_d(S) = \sum_{i=0}^d \alpha_d(C_i) = \frac{1}{2} \sum_{i=0}^d (\alpha_d(C_i) + \alpha_d(-C_i)) < \frac{1}{2} \sum_{l=1}^m \alpha_d(D_l) = \frac{1}{2}.$$

The inequality here is strict because (14) implies that $m > 2d+2$ for $d \geq 3$, which means that there exists D_l (with $\alpha_d(D_l) > 0$) such that $D_l \neq C_i$ and $D_l \neq -C_i$ for all $i = 0, \dots, d$.

Now, let us prove the second part of Proposition 5.4. Let e_1, \dots, e_d be the standard orthonormal basis in \mathbb{R}^d . We consider the simplex

$$S_0 := \text{conv}(0, e_1, \dots, e_d)$$

together with the following two families of simplices indexed by $t \in [0, 1)$:

$$S_1(t) := \text{conv}\left(0, e_1, \dots, e_{d-1}, (1-t)e_d + t(e_1 + \dots + e_{d-1})\right)$$

and

$$S_2(t) := \text{conv}\left(0, e_1 - t \cdot \frac{e_1 + \dots + e_d}{d}, \dots, e_d - t \cdot \frac{e_1 + \dots + e_d}{d}\right).$$

We have $S_1(0) = S_2(0) = S_0$ and

$$\lim_{t \rightarrow 1^-} \gamma_d(S_1(t)) = 0, \quad \lim_{t \rightarrow 1^-} \gamma_d(S_2(t)) = \frac{1}{2}.$$

Pasting these families together, we obtain a continuous family of simplices whose angle sums change from 0 to $1/2$. By continuity (see Section 5.1), this completes the proof. \square

ACKNOWLEDGEMENT

The authors are grateful to Anna Gusakova for her collaboration in proving Proposition 5.1 and to Sergei Ivanov for communicating to us the idea of proof of Proposition 5.4.

REFERENCES

1. E. Bosetto, *Systems of stochastically independent and normally distributed random points in the Euclidean space E_3* . — *Beiträge Algebra Geom.* **40**, No. 2 (1999), 291–301.
2. D. V. Feldman, D. A. Klain, *Angles as probabilities*. — *Amer. Math. Monthly* **116**, No. 8 (2009), 732–735.
3. S. Finch, *Random Gaussian tetrahedra*. [arXiv:1005.1033](https://arxiv.org/abs/1005.1033) (2010).
4. S. Ivanov, *Private communication*, 2018.
5. Z. Kabluchko, C. Thäle, D. Zaporozhets, *Beta polytopes and Poisson polyhedra: f -vectors and angles*. [arXiv:1805.01338](https://arxiv.org/abs/1805.01338) (2018).
6. Z. Kabluchko, D. Zaporozhets, *Absorption probabilities for Gaussian polytopes, and regular spherical simplices*. [arXiv:1704.04968](https://arxiv.org/abs/1704.04968) (2017).
7. C. A. Rogers, *The packing of equal spheres*. — *Proc. London Math. Soc.* (3) **8** (1958), 609–620.
8. L. Schläfli, *Theorie der vielfachen Kontinuität*. — In: *Gesammelte mathematische Abhandlungen* (1950), pp. 167–387.
9. R. Schneider, W. Weil, *Stochastic and Integral Geometry*, Springer-Verlag, Berlin, 2008.

10. A. M. Vershik, P. V. Sporyshev, *Asymptotic behavior of the number of faces of random polyhedra and the neighborliness problem*. — *Selecta Math. Soviet.* **11**, No. 2 (1992), 181–201.

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Поступило 28 ноября 2018 г.

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