



# The variance conjecture on projections of the cube <sup>☆</sup>



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## ABSTRACT

We prove that the uniform probability measure  $\mu$  on every  $(n - k)$ -dimensional projection of the  $n$ -dimensional unit cube verifies the variance conjecture with an absolute constant  $C$

$$\text{Var}_\mu |x|^2 \leq C \sup_{\theta \in S^{n-1}} \mathbb{E}_\mu \langle x, \theta \rangle^2 \mathbb{E}_\mu |x|^2,$$

provided that  $1 \leq k \leq \sqrt{n}$ . We also prove that if  $1 \leq k \leq n^{\frac{2}{3}}(\log n)^{-\frac{1}{3}}$ , the conjecture is true for the family of uniform probabilities on its projections on random  $(n - k)$ -dimensional subspaces.

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## 1. Introduction and notation

The (generalized) variance conjecture states that there exists an absolute constant  $C$  such that for every centered log-concave probability  $\mu$  on  $\mathbb{R}^n$  (i.e. of the form  $d\mu = e^{-v(x)}dx$  for some convex function  $v : \mathbb{R}^n \rightarrow (-\infty, \infty]$ )

$$\text{Var}_\mu |x|^2 \leq C \lambda_\mu^2 \mathbb{E}_\mu |x|^2,$$

where  $\mathbb{E}_\mu$  and  $\text{Var}_\mu$  denote the expectation and the variance with respect to  $\mu$  and  $\lambda_\mu$  is the largest eigenvalue of the covariance matrix, i.e.  $\lambda_\mu^2 = \max_{\theta \in S^{n-1}} \mathbb{E}_\mu \langle x, \theta \rangle^2$  where  $S^{n-1}$  denotes the unit Euclidean sphere in  $\mathbb{R}^n$ .

This conjecture was first considered in the context of the so called Central Limit Problem for isotropic convex bodies in [7] and it is a particular case of a more general statement, known as the Kannan, Lovász, and Simonovits or KLS-conjecture, see [9], which conjectures the existence of an absolute constant  $C$  such

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that for any centered log-concave probability in  $\mathbb{R}^n$  and any locally Lipschitz function  $g : \mathbb{R}^n \rightarrow \mathbb{R}$  such that  $\text{Var}_\mu g(x)$  is finite

$$\text{Var}_\mu g(x) \leq C\lambda_\mu^2 \mathbb{E}_\mu |\nabla g(x)|^2.$$

In recent years a number of families of measures have been proved to verify these conjectures (see [2] for a recent review on the subject). For instance, the KLS-conjecture is known to be true for the Gaussian probability and the uniform probability measures on the  $\ell_p^n$ -balls, some revolution bodies, the simplex and, with an extra  $\log n$  factor, on unconditional bodies and log-concave probabilities with many symmetries (see [4–6,8,10,11,14]). The best general known result for the KLS-conjecture adds a factor  $\sqrt{n}$  and is due to Lee and Vempala (see [12]). Besides, the variance conjecture is known to be true for uniform probabilities on unconditional bodies and on hyperplane projections of the cross-polytope and the cube (see [10] and [1]). The best general estimate for the variance conjecture is the one given by Lee and Vempala for the KLS-conjecture.

We would like to remark that, while in the case of the KLS-conjecture one can assume without loss of generality that  $\mu$  is isotropic (since then every linear transformation of the measure verifies it) this is not the case when we restrict to the variance conjecture, as we are considering only the function  $g(x) = |x|^2$ .

Before stating our results let us introduce some more notation. Let

$$B_\infty^n := \{x \in \mathbb{R}^n : |x_i| \leq 1, \forall 1 \leq i \leq n\}$$

denote the  $n$ -dimensional unit cube and, for any  $1 \leq k \leq n$ , let  $G_{n,n-k}$  be the set of all  $(n-k)$ -dimensional subspaces of  $\mathbb{R}^n$ . For any  $E \in G_{n,n-k}$  we will denote by  $K := P_E B_\infty^n$  the orthogonal projection of  $B_\infty^n$  onto  $E$  and by  $\mu$  the uniform probability on  $K$ .  $\{e_i\}_{i=1}^n$  will denote the standard canonical basis in  $\mathbb{R}^n$ . As mentioned before, it was proved in [1] that the family of uniform probabilities on any  $(n-1)$ -dimensional projection of  $B_\infty^n$  verifies the variance conjecture.

In this paper we will prove the following

**Theorem 1.1.** *There exists an absolute constant  $C$  such that for any  $1 \leq k \leq \sqrt{n}$  and any  $E \in G_{n,n-k}$ , if  $\mu$  denotes the uniform probability measure on  $K = P_E B_\infty^n$ , then*

$$\text{Var}_\mu |x|^2 \leq C\lambda_\mu^2 \mathbb{E}_\mu |x|^2.$$

We will also prove the following theorem, which shows that for  $k$  in a larger range, the variance conjecture is true for the family of uniform probabilities on the projections of  $B_\infty^n$  on a random  $(n-k)$ -dimensional subspace. For that matter, we denote by  $\mu_{n,n-k}$  the Haar probability measure on  $G_{n,n-k}$ .

**Theorem 1.2.** *There exist absolute constants  $C, c_1, c_2$  such that for any  $1 \leq k \leq \frac{n^{\frac{2}{3}}}{(\log n)^{\frac{1}{3}}}$ , if  $\mu$  denotes the uniform probability measure on  $K = P_E(B_\infty^n)$ , the measure  $\mu_{n,n-k}$  of the subspaces  $E \in G_{n,n-k}$  for which*

$$\text{Var}_\mu |x|^2 \leq C\lambda_\mu^2 \mathbb{E}_\mu |x|^2$$

*is greater than  $1 - c_1 e^{-c_2 n^{\frac{2}{3}} (\log n)^{\frac{2}{3}}}$ .*

The main tool to prove both theorems will be to decompose an integral on  $K$  as the sum of the integrals on the projections of some  $(n-k)$ -dimensional faces. It was proved in [3] that for any  $E \in G_{n,n-k}$  there exist  $F_1, \dots, F_l$  a set of  $(n-k)$ -dimensional faces of  $B_\infty^n$  such that for any integrable function  $f$  on  $K$

$$\mathbb{E}_\mu f := \frac{1}{|K|} \int_K f(x) dx = \sum_{i=1}^l \frac{|P_E(F_i)|}{|K|} \mathbb{E}_{P_E(F_i)} f(x)$$

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