

**SLICING CONVEX BODIES—BOUNDS FOR SLICE AREA  
 IN TERMS OF THE BODY'S COVARIANCE**

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**ABSTRACT.** Let  $Q$  be a zero-symmetric convex set in  $\mathbf{R}^N$  with volume 1 and covariance matrix  $V^2 \text{Id}_{N \times N}$ . Let  $P$  be a  $K$ -dimensional vector subspace of  $\mathbf{R}^n$ ,  $K < N$ , and let  $J = N - K$ . Then there exist constants  $C_1(J)$  and  $C_2(J)$  such that

$$V^{-J}C_1(J) < \text{vol}_K(P \cap Q) < V^{-J}C_2(J).$$

The lower bound has applications to Diophantine equations.

**1. Introduction.** The restriction to bodies of covariance a constant multiple of  $\text{Id}_{N \times N}$  and volume 1 made in the abstract is not essential, as any centro-symmetric convex body can be brought to that form by a suitable linear transformation. Yet such bodies comprise the most important special cases. The unit cube, the  $L^1$  ball  $\sum_1^N |x_i| \leq r$  of volume 1 and the "complex cube"  $|z_i| \leq \pi^{-1/2}$ ,  $1 \leq i \leq N$ , are examples.

For the case of the cube, real or complex, only the upper bound is of interest as there is a sharp lower bound of 1, independent of  $K$  and  $N$ , due to Vaaler [10].<sup>2</sup> For the real cube in the case  $K = N - 1$  an upper bound of 5 was given in [6], which we improve here to  $\sqrt{6}$ .

Examples with a cube show that  $C_1(J) \not\prec 12^{-J/2}$  and  $C_2(J) \not\prec 6^{-J/2}$ . We take

$$C_1(J) = (J + 2)^{-J/2} \pi^{J/2} \Gamma(\frac{1}{2}J + 1),$$

$$C_2(J) = 2(8(\log 2)^{-J-3}(J + 2)!)^{J/2} \pi^{J/2} \Gamma(\frac{1}{2}J + 1) \quad \text{for } J \geq 2,$$

and  $C_2(1) = 1/\sqrt{2}$ . Then we have

**THEOREM 1.** Let  $Q$  be a centro-symmetric convex body in  $\mathbf{R}^N$  with volume  $U$  and diagonal covariance matrix  $(V_i^2 \delta_{ij})$ ,  $1 \leq i, j \leq N$ . Let  $P$  be a  $K$ -dimensional vector subspace of  $\mathbf{R}^N$  with  $K < N$ ,  $K + J = N$ . Let

$$V = \left( U^{-N-2} \prod_1^N V_i^2 \right)^{1/2N} \quad \text{and} \quad c_i = V_i V^{-1} U^{-1/2}.$$

Let  $W_K$  be the product of the  $K$  smallest  $c_i$  and  $W'_K$  the product of the  $K$  largest  $c_i$ .

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<sup>2</sup>The lower bound 1 for the case  $K = N - 1$  is implicit in H. Hadwiger's *Gitterperiodische Punktmen-gen und Isoperimetrie*, Monatsh. Math. 76 (1972), 410-418.

Then

$$C_1(J)W_K V^{-J} \leq \text{vol}_K(P \cap Q) \leq C_2(J)W'_K V^{-J}.$$

If all the  $V_i$  are equal to some  $V$ , and  $U = 1$  then all  $c_i$ ,  $W_K$ , and  $W'_K$  are 1 and Theorem 1 reduces to

**THEOREM 1'.** *Let  $Q$  be a centro-symmetric convex body in  $\mathbb{R}^N$  with volume 1 and covariance matrix  $V^2 \text{Id}_{N \times N}$ . Let  $P$  be a  $K$ -dimensional vector subspace of  $\mathbb{R}^N$ ,  $K < N$  and  $J = N - K$ . Then*

$$V^{-J}C_1(J) \leq \text{vol}_K(P \cap Q) \leq V^{-J}C_2(J).$$

Theorem 1 follows from 1' by an elementary lemma whose proof we omit. (But see Example 12, [8].)

**LEMMA 1.** *Let  $E$  be a convex set in  $\mathbb{R}^N$  of dimension  $K < N$ . Let  $T$  be a linear transformation which maps each unit coordinate vector  $\bar{e}_i$ ,  $1 \leq i \leq N$ , in  $\mathbb{R}^N$  to  $c_i \bar{e}_i$ , with  $c_i \geq 0$ . Let  $E' = TE$ . Let  $W_K$  be the product of the  $K$  smallest  $c_i$  and  $W'_K$  the product of the  $K$  largest  $c_i$ . Then*

$$W_K \text{vol}_K E \leq \text{vol}_K E' \leq W'_K \text{vol}_K E.$$

One may adapt our lower bound to Vaaler [12] by using a variant of his Lemma 6. Suppose  $Q \subseteq \mathbb{R}^N$  has volume 1, is centro-symmetric and has covariance matrix  $\text{Cov}(Q) = V^2 \text{Id}_{N \times N}$ . Let  $L_j(\bar{x})$ ,  $1 \leq j \leq N$ , be  $N$  real linear forms in  $K$  variables,  $L_j(\bar{x}) = \sum_{k=1}^K a_{jk} x_k$ , so that  $A = (a_{jk})$  is an  $N \times K$  matrix, with  $N > K$ ,  $N = K + J$ .

**LEMMA 6'.** *Let  $M$  be a positive integer and suppose that*

$$M |\det A^T A|^{1/2} < V^{-J} C_1(J).$$

*Then there exist at least  $M$  distinct pairs of nonzero lattice points  $\pm \bar{v}_m$ ,  $1 \leq m \leq M$ , such that for each  $m$ ,*

$$\bar{l}_m = (L_j(\bar{v}_m)) \in 2Q.$$

With Lemma 6' in place of Lemma 6 of [12] and following [12] otherwise, we have a generalization of that paper's main result. Let  $\Lambda_j(\bar{x}) = \sum_{k=1}^K a'_{jk} x_k$  for  $1 \leq j \leq J$  be  $J$  real linear forms in  $K$  real variables  $x_1, \dots, x_K$ . Let  $N = J + K$ . Let  $Q$  be centro-symmetric and convex with  $\text{vol}_N Q = 1$  and  $\text{Cov}(Q) = V^2 \text{Id}_{N \times N}$ . For  $\bar{y} \in \mathbb{R}^K$  let  $l(\bar{y}) = (y_1, y_2, \dots, y_K, \Lambda_1(\bar{y}), \dots, \Lambda_J(\bar{y})) \in \mathbb{R}^N$ .

**THEOREM 2.** *Let  $M$  be a positive integer and suppose that*

$$M^2 \prod_1^J \left( 1 + \sum_1^K |a'_{jk}|^2 \right) < C_1(J) V^{-J}.$$

*Then there exist  $M$  distinct pairs of nonzero lattice points  $\pm \bar{v}_m$ ,  $1 \leq m \leq M$ , in  $\mathbb{Z}^K$  such that  $l(\pm \bar{v}_m) \in 2Q$ .*

REMARK. When  $Q$  is taken to be the unit cube we get a weaker version of Theorem 1 in [12] with all  $\alpha_k$  and  $\beta_j$  there being 1; our more general lower bound is not as sharp for the cube as that in [11].

2. **Proofs.** Lemma 6' follows from the proof in [11] of Theorem 2 (which appears in [12] as Lemma 6), but we use our Theorem 1' in place of Theorem 1 of [11].

For the proof of Theorem 1 from Theorem 1' we take the  $c_i$  in the statement of Theorem 1 as defining  $T$ , and let  $Q' = T^{-1}Q$ ,  $P' = T^{-1}P$ . Applying Theorem 1' to  $Q'$  and then Lemma 1 to the resulting bounds on  $\text{vol}_K(P \cap Q)$  gives Theorem 1.

It remains to prove Theorem 1'. For the proof we shall need some lemmas about centro-symmetric log concave functions. (A function  $f: \mathbf{R}^J \rightarrow \mathbf{R}^+ \cup \{0\}$  is log concave if  $\log f$  is concave.)

LEMMA 2. Suppose  $f: \mathbf{R}^J \rightarrow \mathbf{R}^+ \cup \{0\}$  satisfies  $f(\bar{x}) = f(-\bar{x})$ ,  $\{\bar{x}: f(\bar{x}) > t\}$  is convex and open for each  $t$ , and  $\int_{\mathbf{R}^J} f(\bar{x}) d^J(\bar{x}) = 1$ . Suppose further that for all unit  $\bar{\theta} \in \mathbf{R}^J$ ,

$$\int_{\mathbf{R}^J} (\bar{x} \cdot \bar{\theta})^2 f(\bar{x}) d^J(\bar{x}) < V^2.$$

Then

$$f(\bar{0}) \geq V^{-J}(J + 2)^{-J/2} \pi^{J/2} \Gamma(\frac{1}{2}J + 1) = V^{-J} C_1(J).$$

PROOF. Let  $h: \mathbf{R}^J \rightarrow \mathbf{R}^+ \cup \{0\}$  be constant at  $V^{-J} C_1(J)$  for  $\|\bar{x}\| < R = V(J + 2)^{1/2}$ , and 0 for  $\|\bar{x}\| \geq R$ . Then  $\int_{\mathbf{R}^J} h = 1$  and  $\int_{\mathbf{R}^J} x_i^2 h = V^2$ , so  $h$  satisfies the hypotheses of Lemma 2. Thus to prove the lemma it suffices to prove that if  $f \neq h$ ,  $f(\bar{0}) > h(\bar{0})$ .

We first show that  $f$  can be replaced by an  $f_1$  such that  $f_1$  is circular symmetric, that is,  $f_1(\bar{x}) = f_1(\bar{y})$  if  $\|\bar{x}\| = \|\bar{y}\|$ , and such that  $f_1(\bar{0}) = f(\bar{0})$  and  $f_1$  satisfies the hypotheses of the lemma.

For let  $E_t = \{\bar{x}: f(\bar{x}) > t\}$ . By hypothesis  $E_t$  is convex, and if  $t < s$ ,  $E_s \subseteq E_t$ . Thus

$$\int_{\mathbf{R}^J} f(\bar{x}) = \int_0^{f(0)} \int_{E_t} 1 d^J(\bar{x}) dt, \tag{3}$$

while

$$\int_{\mathbf{R}^J} x_i^2 f(\bar{x}) = \int_0^{f(0)} \int_{E_t} x_i^2 d^J(\bar{x}) dt. \tag{4}$$

Let  $E'_t$  be the ball about  $\bar{0}$  of the same  $\text{vol}_J$  as  $E_t$ , and let  $f_1(\bar{x}) = \sup\{t: \bar{x} \in E'_t\}$ . Then

$$\int_{\mathbf{R}^J} f_1(\bar{x}) = \int_0^{f(0)} \int_{E'_t} 1 d^J(\bar{x}) dt = \int_0^{f(0)} \int_{E_t} 1 d^J(\bar{x}) dt = \int_{\mathbf{R}^J} f(\bar{x}) = 1. \tag{5}$$

Further,

$$\int_{\mathbf{R}^J} x_i^2 f_1(\bar{x}) = J^{-1} \int_{\mathbf{R}^J} \|\bar{x}\|^2 f_1(\bar{x}) = J^{-1} \int_0^{f(0)} \int_{E'_t} \|\bar{x}\|^2 d^J(\bar{x}) dt. \tag{6}$$

Now consider a particular value of  $t$ . If we show that  $\int_{E'_t} \|\bar{x}\|^2 \leq \int_{E_t} \|\bar{x}\|^2$ , the claim about  $f_1$  is proved. So let  $\text{INT} = E'_t \cap E_t$ ,  $\text{EX} = E_t \setminus E'_t$  and  $\text{EX}' = E'_t \setminus E_t$ . Then

$$\int_{E_t} \|\bar{x}\|^2 = \int_{\text{INT}} \|\bar{x}\|^2 + \int_{\text{EX}} \|\bar{x}\|^2 \geq \int_{\text{INT}} \|\bar{x}\|^2 + \int_{\text{EX}'} \|\bar{x}\|^2 = \int_{E'_t} \|\bar{x}\|^2,$$

because  $\|\bar{x}\|^2$  is everywhere in  $\text{EX}$  at least as large as anywhere in  $\text{EX}'$  and  $\text{vol}_{J-1} \text{EX} = \text{vol}_{J-1} \text{EX}'$ . Thus  $\int_{\mathbf{R}^J} \|\bar{x}\|^2 f_1(\bar{x}) d^J(\bar{x}) \leq \int_{\mathbf{R}^J} \|\bar{x}\|^2 f(\bar{x}) d^J(\bar{x})$ , and because of the circular symmetry of  $f_1$ ,  $\int_{\mathbf{R}^J} (\bar{x} \cdot \bar{\theta})^2 f_1(\bar{x}) d^J(\bar{x}) \leq V^2$  for all unit vectors  $\bar{\theta}$ .

Now suppose the lemma is false and  $h(\bar{0}) > f(\bar{0}) = f_1(\bar{0})$ . Since  $h$  and  $f_1$  depend only on  $\|\bar{x}\| = r$  we shall by an abuse of language write  $h(r), f_1(r)$  for  $h(\bar{x}), f_1(\bar{x})$  when  $\|\bar{x}\| = r$ . (For  $r < R$ ,  $h(r) = h(0)$ , as well.) Now

$$\begin{aligned} \int_{\mathbf{R}^J} f_1(\bar{x}) d^J(\bar{x}) &= \int_0^\infty f_1(r) J\sigma_J r^{J-1} dr = 1 = \int_{\mathbf{R}^J} h(\bar{x}) d^J(\bar{x}) \\ &= V^{-J} C_1(J) \cdot J\sigma_J \int_0^R r^{J-1} dr. \end{aligned}$$

Let  $F_1(r) = \int_0^r f_1(u) J\sigma_J u^{J-1} du$  and  $H(r) = \int_0^r h(u) J\sigma_J u^{J-1} du$ . Since for  $r < R$ ,  $h(0) = h(r) > f_1(r)$  while for  $r > R$ ,  $0 = h(r) < f_1(r)$ , we have  $H(r) > F_1(r)$  for  $0 < r < R$ , and  $H(r) \geq F_1(r)$  in any case. Thus

$$\int_0^\infty 2rF_1(r) dr < \int_0^\infty 2rH(r) dr$$

and

$$\int_0^\infty r^2 dF_1(r) = \int_0^\infty r^2 f_1(r) J\sigma_J r^{J-1} dr > \int_0^\infty r^2 dH(r) = \int_0^R r^2 h(0) J\sigma_J r^{J-1} dr.$$

In other words,

$$\int_{\mathbf{R}^J} \|\bar{x}\|^2 f_1(\bar{x}) d^J(\bar{x}) > \int_{\mathbf{R}^J} \|\bar{x}\|^2 h(\bar{x}) d^J(\bar{x}),$$

a contradiction.  $\square$

**REMARK.** Log concave centro-symmetric functions which satisfy the covariance hypothesis satisfy the other hypotheses of Lemma 2, as is proved in the preliminary lemma of [2]. Lemma 2 gives a sharp lower bound since an extremal function,  $h$ , is found. Our next lemma is not as sharp.

**LEMMA 3.** Suppose  $f: \mathbf{R}^J \rightarrow \mathbf{R}^+ \cup \{0\}$  satisfies  $f(\bar{x}) = f(-\bar{x})$ , is log concave, and  $\int_{\mathbf{R}^J} f(\bar{x}) d^J(\bar{x}) = 1$ . Suppose further that for all unit  $\bar{\theta} \in \mathbf{R}^J$ ,

$$\int_{\mathbf{R}^J} (\bar{x} \cdot \bar{\theta})^2 f(\bar{x}) d^J(\bar{x}) \geq V^2.$$

Then  $f(\bar{0}) \leq V^{-J} C_2(J)$ , where  $C_2(J)$  is the same as in Theorem 1.

**PROOF.** Let  $E_1 = \{\bar{x} \in \mathbf{R}^J: f(\bar{x}) > \frac{1}{2}f(\bar{0})\}$ .  $E_1$  is convex and centro-symmetric. Let  $E_i = iE_1 \setminus (i-1)E_1$  for  $i > 1$ , so that  $\mathbf{R}^J$  is the disjoint  $\cup_{i=1}^\infty E_i$ . Let  $r$  be the minimal radius of  $E_1$ , and let  $\bar{\theta}$  be a unit vector in the direction of a point on  $\partial E_1$  of norm  $r$ .

Since  $\int_{\mathbf{R}^J} (\bar{x} \cdot \bar{\theta})^2 f(\bar{x}) d^J(\bar{x}) \geq V^2$ ,

$$V^2 \leq \sum_{i=1}^{\infty} \int_{E_i} (\bar{x} \cdot \bar{\theta})^2 f(\bar{x}) d^J(\bar{x}) < 4 \sum_{i=1}^{\infty} i^{J+2} 2^{-i} r^2$$

because  $\text{vol}_J E_1 \leq 2/f(\bar{0})$  and  $\int_{E_1} (\bar{x} \cdot \bar{\theta})^2 f(\bar{x}) d^J(\bar{x}) \leq (2/f(\bar{0})) r^2 f(\bar{0}) = 2r^2$ . Now

$$4 \sum_1^{\infty} i^{J+2} 2^{-i} r^2 \leq 8r^2 \int_0^{\infty} s^{J+2} 2^{-s} ds = Dr^2 = 8r^2 (\log 2)^{-J-3} (J+2)!$$

by the integral comparison test. Thus  $r \geq D^{-1/2} V$ . On the other hand  $2 > f(\bar{0}) \text{vol}_J(E_1) \geq f(\bar{0}) \sigma_J r^J$  so

$$\begin{aligned} f(\bar{0}) &\leq 2r^{-J} \sigma_J^{-1} \leq 2V^{-J} D^{J/2} \sigma_J^{-1} \\ &= V^{-J} \cdot 2(8(\log 2)^{-J-3} (J+2)!)^{J/2} \pi^{-J/2} \Gamma(\frac{1}{2}J+1) = V^{-J} C_2(J). \quad \square \end{aligned}$$

In case  $J = 1$  we can find the extremal function and obtain a sharper upper bound.

**LEMMA 4.** *Suppose  $f: \mathbf{R} \rightarrow \mathbf{R}^+ \cup \{0\}$  satisfies  $f(x) = f(-x)$ , is log concave, and  $\int_{\mathbf{R}} f(x) dx = 1, \int_{\mathbf{R}} x^2 f(x) dx = V^2$ . Then  $f(0) \geq 2^{-1/2} V^{-1}$ .*

**PROOF.** Let  $\alpha = f(0)$ . Since  $\int_{-\infty}^{\infty} e^{-2\alpha|x|} dx = 1$ ,  $f(x)$  cannot be everywhere  $> \alpha e^{-2\alpha|x|}$ . They are equal at 0, so either  $f(x) \equiv \alpha e^{-2\alpha|x|}$  or there exists  $\beta > 0$  such that for  $0 < x < \beta, f(x) > \alpha e^{-2\alpha|x|}$  while for  $x > \beta, f(x) < \alpha e^{-2\alpha|x|}$ . (Both are log concave.)

Now let  $F(x) = \int_0^x f(t) dt$  and  $G(x) = \int_0^x \alpha e^{-2\alpha|t|} dt$ . Then

$$\begin{aligned} \int_{-\infty}^{\infty} x^2(f(x) - \alpha e^{-2\alpha|x|}) dx &= 2 \int_0^{\infty} x^2(f(x) - \alpha e^{-2\alpha x}) dx \\ &= 4 \int_0^{\infty} x(G(x) - F(x)) dx \end{aligned}$$

since  $x^2(F - G)$  declines exponentially to zero at  $\pm \infty$ . Since  $F(0) = G(0) = 0$  and  $F(\infty) = G(\infty) = \frac{1}{2}$ , and since  $d(F - G)/dx > 0$  only for  $0 < x < \beta$ ,  $x(G(x) - F(x))$  is always negative and  $\int_{-\infty}^{\infty} x^2(f(x) - \alpha e^{-2\alpha|x|}) dx < 0$ .  $\square$

What does all this have to do with  $\text{vol}_K(P \cap Q)$ ? Let  $K_J$  denote the unit cube  $|x_i| \leq \frac{1}{2}, K < i \leq N$ . Let  $B$  denote the null matrix of  $P$ , so that  $B$  has  $N$  columns and  $J$  rows, and  $B\bar{x} = \bar{0}$  if  $\bar{x} \in P$ . Without loss of generality we may assume  $\text{rank}(B) = J$ . Let  $B_\epsilon = [B|\epsilon \text{Id}_{J \times J}]$ , and let  $P_\epsilon$  be the null space of  $B_\epsilon$ . Then

$$\lim_{\epsilon \rightarrow 0} \text{vol}_N(P_\epsilon \cap (Q \times K_J)) = \text{vol}_K(P \cap Q), \tag{7}$$

and

$$\text{vol}_N(P_\epsilon \cap (Q \times K_J)) = \text{vol}_N\{\bar{z}: B_\epsilon \bar{z} = 0 \text{ and } \bar{z} \in Q \times K_J\}. \tag{8}$$

We wish to compare the volume in (8) to its projection onto the first  $N$  coordinates,  $\text{vol}_N(\{\bar{x} \in Q: B\bar{x} \in \epsilon K_J\})$ . This ratio is  $\epsilon^{-J} (\det B_\epsilon B_\epsilon^T)^{1/2}$ , which we now prove.

We shall need a lemma about how areas are affected by projection. If  $\bar{a}_1 \dots \bar{a}_N$  are column vectors in  $\mathbf{R}^{N+J}$ , if  $A = [\bar{a}_1 \bar{a}_2 \dots \bar{a}_N]^T$  is a matrix of  $N + J$  columns

and  $N$  rows, and  $\text{Box}_A = \{\bar{a}: \bar{a} = \sum_1^N \lambda_i \bar{a}_i, 0 < \lambda_i < 1\}$  then  $\text{vol}_N \text{Box}_A = (\det AA^T)^{1/2}$  (Eves [3]).

Now suppose  $A$  has the form  $[\text{Id}_{N \times N} | -D]$  and let  $A' = [D^T | \text{Id}_{J \times J}]$ .

LEMMA 5.  $\text{Det}(AA^T) = \text{Det}(A'A'^T)$ .

PROOF. By the Cauchy-Binet theorem,  $\text{Det}(AA^T) = \sum_\alpha (\det \alpha)^2$  where the summation is over all  $N \times N$  minors  $\alpha$  of  $A$  [5]. Now a typical  $\alpha$  consists of  $l$  columns of  $\text{Id}_{N \times N}$  and  $m$  columns of  $D$ ,  $l < N$ ,  $m < J$ ,  $l + m = N$ .

$$\alpha = (\bar{e}_{\alpha_1} \bar{e}_{\alpha_2} \cdots \bar{e}_{\alpha_l} - \bar{D}_{\alpha_{l+1}} - \bar{D}_{\alpha_{l+2}} - \cdots - \bar{D}_{\alpha_{m+l}}).$$

Expanding  $\det \alpha$  about the 1st through  $l$ th columns, each of which contain 1 once and otherwise zeros, we have  $\text{Det}(\alpha) = \pm \text{Det}(\beta)$  where  $\beta$  is the square matrix consisting of the intersection of the  $N - l$  rows not indexed by  $\alpha_1, \alpha_2 \dots \alpha_l$  and the  $m$  columns of  $-D$  belonging to  $\alpha$ . Thus  $\det(AA^T) = 1 + \sum_\beta (\det \beta)^2$ , the sum taken over all  $m \times m$  minors of  $D$ ,  $m < J$ . But  $\text{Det}(A'A'^T) = \sum_\gamma (\det \gamma)^2$ , sum over  $J \times J$  minors  $\gamma$  of  $A'$ , and these may be expanded about the columns of  $\text{Id}_{J \times J}$  to obtain again  $1 + \sum_\beta (\det \beta)^2$ .  $\square$

Now let  $\bar{b}_i, 1 < i < N$ , denote the column vectors of  $B$ . Let

$$\bar{a}_i = \left[ \begin{array}{c} \bar{e}_i \\ -\varepsilon^{-1} \bar{b}_i \end{array} \right] \quad \text{and} \quad A^T = [\bar{a}_1 \bar{a}_2 \dots \bar{a}_N].$$

Then  $B_\varepsilon A^T = (0_{J \times N})$  so that the  $\bar{a}_i$  are vectors in  $\text{Null}(B_\varepsilon)$ . When these  $\bar{a}_i$  are projected onto their first  $N$  coordinates they are  $\bar{e}_1, \bar{e}_2 \dots \bar{e}_N$ . Therefore they are linearly independent and form a basis of  $\text{Null}(B_\varepsilon) = P_\varepsilon$ .

Now  $\text{vol}_N \text{Box}_A = (\det AA^T)^{1/2}$  while  $\text{vol}_N \text{Proj}(\text{Box}_A) = 1$  as  $\bar{e}_1 \dots \bar{e}_N$  are orthonormal. The ratio of volumes,  $(\det AA^T)^{1/2}$ , is independent of which measurable set is projected.

Now  $A$  has the form  $A = [\text{Id}_{N \times N} | -D]$  with  $D = \varepsilon^{-1} B^T$ , so that  $A' = [\varepsilon^{-1} B | \text{Id}_{J \times J}]$ . Thus

$$\begin{aligned} (\det AA^T)^{1/2} &= (\det A'A'^T)^{1/2} \quad \text{by Lemma 5} \\ &= \varepsilon^{-J} (\det B_\varepsilon B_\varepsilon^T)^{1/2}, \end{aligned}$$

and this is the ratio of a volume in  $P_\varepsilon$  to the volume of its projection onto the first  $N$  coordinates, as claimed.

Since we can find  $\text{vol}_N(P_\varepsilon \cap (Q \times K_J))$  from  $\text{vol}_N\{\bar{x} \in Q: B\bar{x} \in \varepsilon K_J\}$ , we turn our attention to this last. It may be regarded as the probability that an  $\bar{x}$  taken "at random" from  $Q$  (the probability measure being Lebesgue measure restricted to  $Q$ ) will satisfy  $B\bar{x} \in \varepsilon K_J$ . Let  $f(\bar{x})$  denote the probability density function of  $B\bar{x}$ . Since  $Q$  is convex,  $f(\bar{x})$  is log concave. (This is the key observation.)

For let  $\mu$  denote the probability measure associated with  $f$ , and  $\nu$  Lebesgue measure restricted to  $Q$ . Let  $s = 1 - t, 0 < t < 1$ , and let  $C, D$  be open convex sets in  $\mathbf{R}^J$ . Let  $B^{-1}C, B^{-1}D$  be the inverse images in  $\mathbf{R}^N$  under  $B$  of  $C$  and  $D$  respectively. From Prekopa [9], since  $\chi_Q$ , the characteristic function of  $Q$ , is log concave, the measure  $\nu$  is log concave, that is,  $\nu(sC' + tD') \geq (\nu(C'))^s (\nu(D'))^t$ . Let

$C' = B^{-1}D$ ,  $D' = B^{-1}D$ . Then

$$\begin{aligned}\mu(sC + tD) &= \nu(B^{-1}(sC + tD)) = \nu(sB^{-1}C + tB^{-1}D) \quad (\text{since } B \text{ is linear}) \\ &= \nu(sC' + tD') > (\nu(C'))^s (\nu(D'))^t = (\mu(C))^s (\mu(D))^t.\end{aligned}$$

So  $\mu$  is a log concave measure and, again by Prekópa [9],  $f$  is a log concave function. Now as  $\varepsilon \rightarrow 0$ ,

$$\text{Prob}(B\bar{x} \in \varepsilon K_J) \sim \varepsilon^J f(\bar{0}). \quad (9)$$

Also,

$$\text{Cov}(f)_{ij} = \int_{\mathbb{R}^J} x_i x_j f(\bar{x}) d^J(\bar{x}) = (V^2 BB^T)_{ij}. \quad (10)$$

Let  $\bar{X}$  be the random vector uniformly distributed on  $Q$ . Since  $BB^T$  is selfadjoint and positive definite ( $\text{rank } B = J$ ), there is a square matrix  $S$  such that  $SS^T = BB^T$  [1].

Let  $Y$  be the random vector  $\bar{Y} = S^{-1}B\bar{X}$ . Then  $\text{Cov}(B\bar{X}) = V^2 BB^T$ , so  $\text{Cov}(\bar{Y}) = V^2 S^{-1} BB^T S^{-1T} = V^2 \text{Id}_{J \times J}$ , since for any random vector  $\bar{Z}$ , any square matrix  $U$ ,  $\text{Cov}(U\bar{Z}) = E(U\bar{Z}(U\bar{Z})^T) = E(U\bar{Z}\bar{Z}^T U^T) = U E(\bar{Z}\bar{Z}^T) U^T = U \text{Cov}(\bar{Z}) U^T$  because expectation ( $E$ ) is linear.

Let  $h(\bar{x})$  be the probability density function associated with  $\bar{Y}$ . Then

$$h(\bar{0}) = (\det BB^T)^{1/2} f(\bar{0}), \quad (11)$$

$$\text{Cov}(h) = \text{Cov}(\bar{Y}) = V^2 \text{Id}_{J \times J}, \quad (12)$$

and since log concavity is preserved under the linear transformation  $S^{-1}$ ,  $h$  is log concave.

Applying Lemmas 2 and 3 to  $h$ , and Lemma 4 in case  $J = 1$ , we have from (12) that

$$C_1(J) V^{-J} < h(\bar{0}) < C_2(J) V^{-J} \quad (13)$$

and from (7) through (11) that

$$C_1(J) V^{-J} < \text{vol}_K(P \cap Q) < C_2(J) V^{-J}. \quad \square$$

#### BIBLIOGRAPHY

1. L. Breiman, *Probability*, Addison-Wesley, Reading, Mass., 1968, p. 239.
2. Ju. S. Davidovic, B. I. Korenbljum and B. I. Hacet, *A property of logarithmically concave functions*, Dokl. Akad. Nauk SSSR **85** (1969) = Soviet Math. Dokl. **10** (1969), 477–480.
3. H. Eves, *Elementary matrix theory*, Allyn and Bacon, Boston, Mass., 1966, p. 176.
4. W. Fleming, *Functions of several variables*, Springer-Verlag, New York, 1977.
5. F. Gantmacher, *Matrizenrechnung*, Vol. 1, Deutsche Verlag der Wissenschaften, Berlin, 1970.
6. D. Hensley, *Slicing the cube in  $\mathbb{R}^n$  and probability*, Proc. Amer. Math. Soc. **73** (1979), 95–100.
7. M. Kanter, *Unimodality and dominance for symmetric random vectors*, Trans. Amer. Math. Soc. **229** (1977), 65–85.
8. M. Marcus and H. Minc, *Introduction to linear algebra*, Macmillan, New York, 1965, p. 209.
9. A. Prekópa, *On logarithmic concave measures and functions*, Acta Sci. Math. (Szeged) **34** (1973), 335–343.
10. Y. Rinnot, *On convexity of measures*, Ann. Probability **4** (1976), 1020–1026.
11. J. Vaaler, *A geometric inequality with applications to geometry of numbers*, Pacific J. Math. (to appear).
12. ———, *On linear forms and Diophantine approximation*, Pacific J. Math. (to appear).