## A PROOF OF THE GROTHENDIECK INEQUALITY

ΒY

RONALD E. RIETZ<sup>+</sup>

## ABSTRACT

The fundamental constant of Grothendieck's inequality, defined below, was shown by Grothendieck to be less than  $\sin \pi/2 = 2.301 + .$  We improve the bound slightly, and show that for the positive definite case  $\pi/2$  suffices.

In 1956, A. Grothendieck proved that the supremum of the injective tensor norms of the identity map on  $\mathbb{R}^n$  is uniformly bounded for all *n*, with least upper bound  $\gamma$  between  $\pi/2$  and sinh  $\pi/2$ . This remarkable result leads directly to a number of important theorems involving  $\mathcal{L}_p$ -spaces. These are given in a paper of Lindenstrauss and Pe $\mathcal{L}$ czyński [2], together with the following matrix formulation of  $\gamma$ : let  $A = (a_{ij})$  be a real  $n \times n$  matrix, *n* any finite positive integer, and  $\mathcal{H}$  an arbitrary real Hilbert space with inner product  $(\cdot, \cdot)$ . Define  $|A|_{\mathbb{X}}$  and |A| by

(1) 
$$|A|_{\mathcal{X}} = \sup \{ |\Sigma \Sigma a_{ij}(x_i, y_j)| : x_i, y_j \text{ in } \mathcal{H} \text{ of norm} \leq 1 \}$$

and

(2) 
$$|A| = \sup \{|\Sigma \Sigma a_{ij}t_iu_j| : -1 \le t_i, u_j \le 1\}.$$

Then  $|A|_{\mathbf{x}} \leq \gamma |A|$ .

The proof given by Grothendieck [1, p. 62] and reformulated by Lindenstrauss and Pe $\lambda'$ czyński [2, p. 279] obtains  $\gamma \leq \sinh \pi/2 = 2.301 + by$  averaging (2), with  $t_i = \operatorname{sgn}(x_i, \omega)$  and  $u_j = \operatorname{sgn}(y_j, \omega)$ , over the unit sphere  $\Omega$  in  $\mathbb{R}^n$  with normalized surface measure  $d\omega$ . We obtain a slightly smaller bound,  $\gamma < 2.261$ ,

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by averaging over  $\mathbb{R}^n$  with normalized Gaussian measure and using a variational argument to determine an optimal (in this context) scalar map corresponding to the signum function. Our argument also shows that for positive definite matrices A, we have  $|A|_{\mathbb{X}} \leq \pi/2|A|$ .

Sten Kaijser (private correspondence) has pointed out that our proof generalizes directly to the complex case, and yields an upper bound of 1.607 for the complex constant.

For a fixed positive integer *n*, let  $dG(x) = (2\pi)^{-n/2} \exp(-|x|^2/2) dx$  be normalized Gaussian measure of mean zero and unit variance on  $\mathbb{R}^n$ . Write  $L^2$  for  $L^2(\mathbb{R}^n, dG)$  and let  $\|\cdot\|$  and  $(\cdot, \cdot)$  be the Hilbert norm and inner product in  $L^2$ .

If  $x = (x_i)$  and  $y = (y_i)$  are vectors in  $\mathbb{R}^n$ , let |x| be the Euclidean norm of x, and  $x \cdot y = \sum x_i y_i$  be the inner product of x and y. For  $t \ge 0$ , write  $dm(t) = (2/\pi)^{\frac{1}{2}} \exp(-t^2/2)dt$ , where dt is Lebesgue measure. All unmarked sums  $\sum$  are taken with i and j ranging independently over the integers from 1 through n.

Let  $\mathcal{U}$  be the set of measurable functions  $f(t) \ge 0$ , defined on  $t \ge 0$ , with ess sup  $f(t) \le 1$ . Given  $f \in \mathcal{U}$ , let  $\psi$  be the odd extension of f to domain  $\mathbf{R}$ , i.e.  $\psi(t) = f(t)$  if  $t \ge 0$  and  $\psi(t) = -f(-t)$  if t < 0. Then given  $x \in \mathbf{R}^n$ , we define the real valued functions  $\varphi_x$  and  $\psi_x$  on  $\mathbf{R}^n$  by

(3) 
$$\varphi_x(z) = x \cdot z \text{ and } \psi_x(z) = \psi(x \cdot z) \quad (z \in \mathbf{R}^n).$$

Clearly  $\varphi_x$  and  $\psi_x$  are in  $L^2$  for each  $x \in \mathbb{R}^n$ .

LEMMA 1. Let  $f \in \mathcal{U}$  and  $\psi$  the odd extension of f. If |x| = |y| = 1, then (i)  $(\varphi_x, \varphi_y) = x \cdot y$ (ii)  $(\varphi_x, \psi_y) = K(x \cdot y)$  where

$$K=K_f=\int_0^\infty tf(t)dm(t)$$

(iii)  $\|\varphi_x - \psi_x\|^2 = 1 - 2K + L$ , where

$$L=L_f=\int_0^\infty f^2(t)dm(t).$$

**PROOF.** Since x and y have norm 1, we have  $x = (x \cdot y)y + y'$ , where y' is orthogonal to y. Then

$$\int (\mathbf{x} \cdot \mathbf{z})(\mathbf{y} \cdot \mathbf{z}) dG(\mathbf{z}) = (\mathbf{x} \cdot \mathbf{y}) \cdot \int |\mathbf{y} \cdot \mathbf{z}|^2 dG(\mathbf{z}) + \int (\mathbf{y}' \cdot \mathbf{z})(\mathbf{y} \cdot \mathbf{z}) dG(\mathbf{z}).$$

The measure dG is characterized by assigning the function  $z \rightarrow y \cdot z$  a Gaussian distribution of mean zero and variance  $|y|^2$ , so  $\int |y \cdot z|^2 dG(z) = |y|^2 =$ 

1. Since the Gaussian distributions are basis-invariant, we may choose coordinates so that y is one of the basis vectors, to get  $\int (y' \cdot z)(y \cdot z)dG(z) = 0$ . This completes (i).

For (ii), with 
$$x = (x \cdot y)y + y'$$
 as before,  

$$\int (x \cdot z)\psi(y \cdot z)dG(z) = (x \cdot y) \cdot \int (y \cdot z)\psi(y \cdot z)dG(z) + \int (y' \cdot z)\psi(y \cdot z)dG(z).$$

Again choose coordinates so that y is a basis vector, and integrate first along y. We obtain

$$\int (y \cdot z)\psi(y \cdot z)dG(z) = \int_0^\infty tf(t)dm(t),$$

and

$$\int (y' \cdot z)\psi(y \cdot z)dG(z) = 0.$$

For (iii), expand  $\|\varphi_x - \psi_x\|^2$ , note that

$$(\psi_x,\psi_x)=\int\psi^2(x\cdot z)dG(z)=\int_0^\infty f^2(t)dm(t),$$

and apply (i) and (ii).

THEOREM 1 (Grothendieck). There is  $\blacktriangle$  finite number  $\gamma$ , independent of n, such that  $|A|_{\mathbf{x}} \leq \gamma |A|$  for all  $n \times n$  matrices A.

**PROOF.** Let  $f \in \mathcal{U}$ . Write  $K = K_f$ ,  $L = L_f$ , and let  $\psi$  be the odd extension of f. For any fixed n, let  $A = (a_{ij})$  and  $\mathcal{H}$  be given, with  $|A| \leq 1$ .

Note that  $|A|_{\mathcal{X}}$  is the supremum of the sums  $\sum_{i} ||\sum_{j} a_{ij} x_{j}||$ , taken over  $x_{i}$  in  $\mathcal{X}$  of norm  $\leq 1$ . Since any such set  $\{x_{i}\}$  of *n* vectors spans a subspace of  $\mathcal{X}$  of dimension at most *n*, we may take  $\mathcal{H} = \mathbb{R}^{n}$ . Further, the convexity of the unit sphere in Euclidean space  $\mathbb{R}^{n}$  allows us to choose vectors  $x_{i}$ ,  $y_{i}$  of norm 1 in  $\mathbb{R}^{n}$  such that  $|A|_{\mathcal{X}} = \sum a_{ij}(x_{i} \cdot y_{j})$ .

By Lemma 1, the identity

$$(\psi_x,\psi_y)=(\varphi_x,\psi_y)+(\psi_x,\varphi_y)-(\varphi_x,\varphi_y)-(\varphi_x-\psi_x,\psi_y-\varphi_y)$$

yields:

(4) 
$$\sum a_{ij}(\psi_{x_i}, \psi_{y_j}) = (2K - 1) \sum a_{ij}(x_i \cdot y_j) - \sum a_{ij}(\varphi_{x_i} - \psi_{x_i}, \psi_{y_j} - \varphi_{y_j}).$$

From (iii), the last sum above is bounded in absolute value by  $|A|_{\mathbf{x}}(1-2K+L)$ , and since  $|A| \leq 1$ , we also have  $|\sum a_{ij}(\psi_{x_i}, \psi_{y_j})| \leq 1$ . Hence the triangle inequality applied to (4) yields:

(5) 
$$1 \ge \left| \sum a_{ij}(\psi_{x_i}, \psi_{y_j}) \right| \ge |A|_{\mathbf{x}}(|2K-1|+2K-L-1).$$

This inequality holds for any function f in  $\mathcal{U}$ . In particular, let f be identically 1. Then  $K = (2/\pi)^{\frac{1}{2}}$  and L = 1, so (5) reads

$$1 \ge |A|_{\mathcal{H}}(|2(2/\pi)^{\frac{1}{2}} - 1| + 2(2/\pi)^{\frac{1}{2}} - 2).$$

The coefficient of  $|A|_{\pi}$  is easily seen to be greater than .19, hence  $|A|_{\pi} < (.19)^{-1} = 5.2 + .$ 

COROLLARY 1. If  $f \in \mathcal{U}$  is such that  $2K_f^2 > L_f$ , then  $\gamma \leq (2K_f^2 - L_f)^{-1}$ .

**PROOF.** From the proof of Theorem 1, for any  $f \in \mathcal{U}$  we have:

(5) 
$$1 \ge |A|_{\mathcal{X}}(|2K-1|+2K-L-1))$$

When f is replaced by cf, c > 0, we get  $K_{cf} = cK_f$  and  $L_{cf} = c^2 L_f$ , so (5) becomes:

(6) 
$$c^2 \ge |A|_{\mathbf{x}}(|2cK-1|+2cK-c^2L-1).$$

If c is such that  $2cK - 1 \le 0$ , then (6) is trivial, so we only consider constants c > 0 for which  $4cK - c^2L - 2 > 0$ . By (6),  $|A|_{\mathbf{x}} \le c^2(4cK - c^2L - 2)^{-1}$  for all such c. For fixed K and L, the minimum value of the right side of this inequality occurs when c = 1/K. Since  $2K^2 - L > 0$ , this value of c yields  $|A|_{\mathbf{x}} \le (2K^2 - L)^{-1}$ . This completes the proof.

We observe that the real-valued mapping  $f \rightarrow (2K_f^2 - L_f)$  is lower, semicontinuous in the weak\*-topology on  $L^{\infty}(\mathbb{R}^+, dm(t))$  and  $\mathcal{U}$  is a weak\*-closed subset of the unit sphere of this  $L^{\infty}$  space. Hence  $2K_f^2 - L_f$  attains its maximum on  $\mathcal{U}$ , say at  $\mu(t)$ . We now determine  $\mu(t)$  explicitly.

Let *h* be a bounded measurable function on  $[0,\infty)$  such that  $h(t) \ge 0$  on  $[\mu = 0]$ ,  $h(t) \le 1$  on  $[\mu = 1]$  and  $\operatorname{ess\,sup}_{t\ge 0} |\mu(t) + \epsilon h(t)| = 1$  for  $\epsilon > 0$  sufficiently small. Fix such an  $\epsilon$ , write K' and L' for the K and L corresponding to the function  $\mu + \epsilon h$ , and delete the  $\mu$ -subscript on  $K_{\mu}$  and  $L_{\mu}$ . Then

$$2K'^{2} - L' = 2K^{2} - L + 2\epsilon \int_{0}^{\infty} [2Kh(t)t - \mu(t)h(t)]dm(t) + O(\epsilon^{2}).$$

Since  $2K^2 - L \ge 2K'^2 - L'$ , and  $\epsilon$  can be taken arbitrarily small,

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(7) 
$$0 \ge \int_0^\infty [2Kh(t)t - \mu(t)h(t)]dm(t).$$

Three cases need to be considered.

Case 1. Let  $h(t) \ge 0$  and supported on  $[\mu = 0]$ . By (7),  $0 \ge \int_{0}^{\infty} 2tKh(t)dm(t)$ , which is impossible unless  $[\mu = 0]$  has measure zero. Hence  $\mu(t) > 0$  a.e.

Case 2. Let  $h(t) \leq 0$  and supported on  $[\mu = 1]$ . Then (7) implies  $0 \geq \int_0^{\infty} h(t)(2Kt-1)dm(t)$ , so  $2Kt-1 \geq 0$  for all t such that  $\mu(t) = 1$ .

Case 3. Let h(t) be supported on  $[0 < \mu < 1]$ . Note that, for any  $\epsilon \le 1$ , there are functions h, satisfying the given properties, which are strictly positive or strictly negative on  $[0 < \mu < 1]$ . For any such functions  $h, 0 \ge \int_0^{\infty} h(t)[2Kt - \mu(t)]dm(t)$ , and this is possible only if  $2Kt = \mu(t)$  a.e. on  $[0 < \mu < 1]$ .

The information obtained from these cases determines a.e. the maximal function  $\mu(t) = 2Kt$  if  $0 \le t \le 1/2K$  and  $\mu(t) = 1$  if  $t \ge 1/2K$ .

Calculating  $2K^2 - L$  for this  $\mu$ , we immediately obtain the equations  $1 = 2 \cdot \int_0^{1/2k} dm(t)$  and  $2K^2 - L = 2K(2/\pi)^{\frac{1}{2}} \exp(-1/8K^2) - \frac{1}{2}$ . From tables in [3],  $2K^2 - L > .4423$ , so by Theorem 2,  $\gamma < 2.261$ .

If the given matrix A is positive definite, the corresponding value of  $\gamma$  is at most  $\pi/2$ . Grothendieck showed [1, p. 51] that  $\pi/2$  is a lower bound for  $\gamma$  (over all matrices) by calculating, for each *n*, the exact value of a particular tensor norm of the identity on  $\mathbb{R}^n$ , showing that this norm is dominated by the supremum of the injective tensor norms, and letting  $n \to \infty$ . We show the sufficiency of  $\pi/2$  for positive definite A by exploring the fact that the last sum in inequality (4) is non-positive for such matrices.

Let A be a positive definite  $n \times n$  matrix with  $|A| \leq 1$ . Consider the vector space of *n*-tuples of vectors in  $\mathbb{R}^n$ . This is a Hilbert space when endowed with the inner product  $(x, y) = \sum x_i \cdot y_i$ . Here  $\mathbf{x} = (x_1, \dots, x_n)$  and  $\mathbf{y} = (y_1, \dots, y_n)$ , each  $x_i$  and  $y_i$  in  $\mathbb{R}^n$ . This inner product induces the norm  $||\mathbf{x}||^2 = \sum |x_i|^2$ . Now  $\sum a_{ij}(x_i \cdot y_j) = (A\mathbf{x}, \mathbf{y}) = (A^{\frac{1}{2}}\mathbf{x}, A^{\frac{1}{2}}\mathbf{y})$ , since A is positive definite: hence  $|\sum a_{ij}(x_i \cdot y_j)| \leq ||A^{\frac{1}{2}}\mathbf{x}|| \cdot ||A^{\frac{1}{2}}\mathbf{y}||$ , with equality, in particular, whenever  $\mathbf{x} = \mathbf{y}$ . This establishes:

LEMMA 3. If A is positive definite, there exist vectors  $x_1, \dots, x_n$ , each of unit norm in  $\mathbb{R}^n$ , such that  $|A| = \sum a_{ij}(x_i \cdot x_j)$ .

THEOREM 4. For any positive definite  $n \times n$  matrix A,  $|A|_{x} \leq \pi/2$ .

**PROOF.** We follow the proof of Theorem 2. Again  $f \in \mathcal{U}$  and  $\psi$  is the odd extension of f.

Assume  $|A| \leq 1$  and choose  $x_1, \dots, x_n$  of norm 1 in  $\mathbb{R}^n$  such that  $|A|_{\mathfrak{X}} = \sum a_{ij}(x_i \cdot x_j)$ . Then inequality (4) of the proof of Theorem 2 holds with  $y_j$  replaced by  $x_j$ . But  $\sum a_{ij}(\varphi_{x_i} - \psi_{x_i}, \varphi_{x_j} - \psi_{x_j})$  is non-negative, so  $\sum a_{ij}(\psi_{x_i}, \psi_{x_j}) \geq |A|_{\mathfrak{X}}(2K-1)$  and  $1 \leq |A|_{\mathfrak{X}}(2K-1)$ . Replace f by cf, c > 0. Then  $c^2 \geq |A|_{\mathfrak{X}}(2cK-1)$ , so  $|A|_{\mathfrak{X}} \leq c^2(2cK-1)^{-1}$  whenever  $c \geq 1/2K$ . The minimum value of this bound for  $|A|_{\mathfrak{X}}$ , over all c, occurs when c = 1/K, so  $|A|_{\mathfrak{X}} \leq 1/K^2$ . K is clearly maximized when f is identically 1, and then  $K = (2/\pi)^{\frac{1}{2}}$ . Thus  $|A|_{\mathfrak{X}} \leq \pi/2$ .

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DEPARTMENT OF MATHEMATICS.

**GUSTAVUS ADOLPHUS COLLEGE.** 

ST. PETER, MINNESOTA 56082 USA.