

AN INEQUALITY INVOLVING THE ℓ_1 , ℓ_2 , AND ℓ_∞ NORMS

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For $\mathbf{x} \in [0, 1]^n$ with $\|\mathbf{x}\|_1 = 1$ and $\mathbf{y} \in [1, \infty)^n$, we prove that

$$\frac{\|\mathbf{xy}\|_\infty}{\|\mathbf{xy}\|_2} \leq \|\mathbf{x}\|_\infty \|\mathbf{y}\|_1 \frac{\|\mathbf{y}\|_\infty}{\|\mathbf{y}\|_2},$$

where $\mathbf{xy} \in \mathbb{R}^n$ is the vector with components $x_i y_i$. This bound does not seem to easily follow from known inequalities, and the proof technique may be of independent interest.

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1. Statement of Main Result

Our objective^a is to prove

Theorem 1.1. *For $n \in \mathbb{N}$, $\mathbf{x} \in [0, 1]^n$ with $\|\mathbf{x}\|_1 = 1$ and $\mathbf{y} \in [1, \infty)^n$, we have*

$$\frac{\|\mathbf{xy}\|_\infty}{\|\mathbf{xy}\|_2} \leq \|\mathbf{x}\|_\infty \|\mathbf{y}\|_1 \frac{\|\mathbf{y}\|_\infty}{\|\mathbf{y}\|_2}.$$

This bound does not seem to easily follow from known inequalities.

2. Notation

We use the conventions $[n] = \{1, \dots, n\}$ and $\mathbb{N} = \{1, 2, \dots\}$ and write $\mathbf{1}_{\{ \cdot \}}$ to denote the 0-1 truth value of the subscripted predicate. For $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, we denote

^aThis problem arose in the course of trying to efficiently learn random finite-state automata. Although that approach had to be abandoned [1], perhaps our result will still be of some interest — if only as a mathematical curiosity.

their componentwise product by $\mathbf{xy} \in \mathbb{R}^n$. The ℓ_p norm of a vector $\mathbf{x} \in \mathbb{R}^n$ is given by

$$\|\mathbf{x}\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{1/p}$$

for $1 \leq p < \infty$ and $\|\mathbf{x}\|_\infty = \max_{i \in [n]} |x_i|$. The $(n - 1)$ -dimensional simplex Δ_{n-1} is defined by

$$\Delta_{n-1} = \{\mathbf{z} \in [0, 1]^n : \|\mathbf{z}\|_1 = 1\}.$$

Denote the all 1s vector by $\bar{\mathbf{1}} = (1, 1, \dots, 1)$.

3. Proof of Theorem 1.1

As $\|\mathbf{xy}\|_\infty \leq \|\mathbf{x}\|_\infty \|\mathbf{y}\|_\infty$, we may break up our analysis into two cases: (I) $\|\mathbf{xy}\|_\infty = \|\mathbf{x}\|_\infty \|\mathbf{y}\|_\infty$ and (II) $\|\mathbf{xy}\|_\infty < \|\mathbf{x}\|_\infty \|\mathbf{y}\|_\infty$.

Case (I): $\|\mathbf{xy}\|_\infty = \|\mathbf{x}\|_\infty \|\mathbf{y}\|_\infty$.

Proving the theorem for this case amounts to showing that

$$\|\mathbf{y}\|_2 \leq \|\mathbf{y}\|_1 \|\mathbf{xy}\|_2 \tag{3.1}$$

whenever $\mathbf{x} \in \Delta_{n-1}, \mathbf{y} \in [1, \infty)^n$ and

$$\|\mathbf{xy}\|_\infty = \|\mathbf{x}\|_\infty \|\mathbf{y}\|_\infty. \tag{3.2}$$

Assume without loss of generality that $y_1 \geq y_2 \geq \dots \geq y_n$. Also, there is no loss of generality in assuming that $x_1 = \|\mathbf{x}\|_\infty$. We will show that (3.1) holds under the condition

$$x_1 \geq n^{-1}, \tag{3.3}$$

which certainly holds whenever (3.2) does. Let $M = x_1$ be a parameter and let us minimize the quadratic form $\sum_{i=2}^n x_i^2 y_i^2$ over x_2, \dots, x_n , subject to $x_i \geq 0, i = 2, \dots, n$ and $\sum_{i=2}^n x_i = 1 - x_1$. The following Lagrangian incorporates the objective function and the equality constraint (the inequality constraint will be satisfied automatically):

$$\mathcal{L} = \frac{1}{2} \sum_{i=2}^n x_i^2 y_i^2 - \lambda \sum_{i=2}^n x_i.$$

Setting $\partial \mathcal{L} / \partial x_i = 0$, we get

$$x_i = \lambda / y_i^2$$

for $i = 2, \dots, n$, while the normalization constraint yields

$$\lambda = \frac{1 - M}{\sum_{i=2}^n y_i^{-2}} = (1 - M)W,$$

where $W^{-1} = \sum_{i=2}^n y_i^{-2}$. Thus,

$$\min\{\|\mathbf{xy}\|_2^2 : \mathbf{x} \in \Delta_{n-1}, x_1 = M\} = M^2 y_1^2 + (1 - M)^2 W.$$

Proving (3.1) is now reduced to showing that

$$\|\mathbf{y}\|_2^2 \leq \|\mathbf{y}\|_1^2 (M^2 y_1^2 + (1 - M)^2 W).$$

Minimizing the quadratic polynomial $f(M) = M^2 y_1^2 + (1 - M)^2 W$ with respect to M and recalling condition (3.3), we may take

$$M = \max\left\{\frac{W^2}{W^2 + y_1^2}, \frac{1}{n}\right\}.$$

However, if $W^2/(W^2 + y_1^2) \geq n^{-1}$ then we have $W^2 + y_1^2 \geq nW^2$, which means that

$$\frac{W^2}{W^2 + y_1^2} \leq \frac{1}{n},$$

and thus we must have $M = n^{-1}$.

At this point, (3.1) has been reduced to showing that

$$\|\mathbf{y}\|_2^2 \leq \|\mathbf{y}\|_1^2 (n^{-2} y_1^2 + (1 - n^{-1})^2 W) = \|\mathbf{y}\|_1^2 \left(\frac{y_1^2}{n^2} + \frac{(1 - n^{-1})^2}{\sum_{i=2}^n y_i^{-2}} \right).$$

Actually, since $y_i \geq 1$ implies

$$\sum_{i=2}^n y_i^{-2} \leq n - 1,$$

it suffices to prove the stronger claim:

$$\|\mathbf{y}\|_2^2 \leq \|\mathbf{y}\|_1^2 \left(\frac{y_1^2}{n^2} + \frac{(1 - n^{-1})^2}{n - 1} \right) = \|\mathbf{y}\|_1^2 \left(\frac{y_1^2 + n - 1}{n^2} \right).$$

The latter claim is equivalent to the assertion that

$$\|\mathbf{y}\|_2^2 \leq \|\mathbf{y}\|_1^2 \left(\frac{\|\mathbf{y}\|_\infty^2 + n - 1}{n^2} \right) \tag{3.4}$$

for all $\mathbf{y} \in [1, \infty)^n$. Indeed, we claim that for $\mathbf{y} \in [1, L]^n$, the ratio $\|\mathbf{y}\|_2/\|\mathbf{y}\|_1$ is maximized when \mathbf{y} is of the form $\mathbf{y} = (L, 1, 1, \dots, 1)$. To see this, define the function $f : [0, L]^n \rightarrow \mathbb{R}$ by $f(\mathbf{y}) = (\|\mathbf{y}\|_2/\|\mathbf{y}\|_1)^2$. It is a simple consequence of Hölder's inequality that f is maximized over $[0, L]^n$ at the standard basis elements $\{\mathbf{e}_i\}$, such as $\mathbf{e}_1 = (1, 0, 0, \dots, 0)$. Since $\partial f/\partial y_i|_{\mathbf{y}=\mathbf{e}_1} < 0$ for $i \neq 1$, it follows that there exist some $\epsilon > 0$ such that $f(\mathbf{y})$ is maximized over $[\epsilon, L]^n$ at $(L, \epsilon, \epsilon, \dots, \epsilon)$. However, since $f(\mathbf{y}) = f(\mathbf{y}/\min_{i \in [n]} y_i)$, it follows that $f(\mathbf{y})$ is maximized over $[1, L/\epsilon]^n$ at $(L/\epsilon, 1, 1, \dots, 1)$. Since L was arbitrary, the claim follows.

Substituting the values $\|\mathbf{y}\|_2^2 = L^2 + n - 1$ and $\|\mathbf{y}\|_1^2 = (L + n - 1)^2$ into (3.4) yields (after simplifying) the equivalent inequality $n^2 \leq (L + n - 1)^2$, which clearly holds when $L \geq 1$.

Case (II): $\|\mathbf{x}\mathbf{y}\|_\infty < \|\mathbf{x}\|_\infty\|\mathbf{y}\|_\infty$.

Proving the theorem for this case is equivalent to showing that the function

$$F(\mathbf{x}, \mathbf{y}) = \frac{\|\mathbf{y}\|_2\|\mathbf{x}\mathbf{y}\|_\infty}{\|\mathbf{y}\|_1\|\mathbf{y}\|_\infty\|\mathbf{x}\|_\infty\|\mathbf{x}\mathbf{y}\|_2}$$

is upper-bounded by 1 on $\Delta_{n-1} \times [1, \infty)^n$.

Fix an $\mathbf{x} \in \Delta_{n-1}$ and pick any $\mathbf{y} \in [1, \infty)^n$. Define the quantities

$$\begin{aligned} I &= \{i \in [n] : y_i = \|\mathbf{y}\|_\infty\}, \\ w &= \max\{y_i : i \notin I\}, \\ h &= \|\mathbf{y}\|_\infty - w, \end{aligned} \tag{3.5}$$

(in words, I is the set of indices of \mathbf{y} 's maximal coordinates, w is the second largest entry in \mathbf{y} , and h is the gap between w and $\|\mathbf{y}\|_\infty$). Thus, we may write \mathbf{y} as $\mathbf{y} = \tilde{\mathbf{y}}(h)$, where

$$\begin{aligned} \tilde{\mathbf{y}}(h)_i &= \min\{y_i, w\} + \mathbf{1}_{\{i \in I\}}h \\ &= \begin{cases} w + h, & i \in I \\ y_i, & i \notin I \end{cases}. \end{aligned}$$

We claim that, for a fixed $\mathbf{x} \in \Delta_{n-1}$, the function

$$f(h) = F(\mathbf{x}, \tilde{\mathbf{y}}(h))^2$$

is decreasing for $h \geq 0$. Once this claim is proven, we may let $\mathbf{y}' = \tilde{\mathbf{y}}(0)$, compute I', w', h' from \mathbf{y}' as in (3.5) and repeat this process, obtaining a sequence of $\mathbf{y}', \mathbf{y}'', \mathbf{y}''', \dots$, until either case (I) becomes applicable or the vector $\bar{\mathbf{1}} = (1, 1, \dots, 1)$ is reached. Since by the claim $f(0) \leq f(h)$, we have

$$F(\mathbf{x}, \mathbf{y}) \leq F(\mathbf{x}, \mathbf{y}') \leq F(\mathbf{x}, \mathbf{y}'') \leq \dots$$

Note also that $F(\mathbf{x}, \bar{\mathbf{1}}) = n^{-1/2}\|\mathbf{x}\|_2^{-1} \leq 1$. Thus, showing that $f(h)$ is decreasing is all that remains to finish off case (II).

Let us write

$$\begin{aligned} \|\tilde{\mathbf{y}}\|_2^2 &= |I|(w+h)^2 + \sum_{i \notin I} y_i^2 = k(w+h)^2 + \zeta \\ \|\tilde{\mathbf{y}}\|_\infty^2 &= (w+h)^2 \\ \|\tilde{\mathbf{y}}\|_1^2 &= \left(|I|(w+h) + \sum_{i \notin I} y_i \right)^2 = (k(w+h) + \theta)^2 \\ \|\mathbf{x}\tilde{\mathbf{y}}\|_2^2 &= \sum_{i \in I} x_i^2(w+h)^2 + \sum_{i \notin I} x_i^2 y_i^2 = \eta(w+h)^2 + \nu, \end{aligned}$$

where $k, \zeta, \theta, \eta, \nu$ are defined in the obvious way.

Consider a fixed $(\mathbf{x}, \mathbf{y}) \in \Delta_{n-1} \times [1, \infty)^n$ and let I, w, h be as in (3.5). Putting $j = \operatorname{argmax}\{x_i y_i : i \in [n]\}$, we have $\|\mathbf{x}\tilde{\mathbf{y}}\|_\infty = x_j y_j$. Hence,

$$\begin{aligned} f(h) &= F(\mathbf{x}, \tilde{\mathbf{y}}(h))^2 = \frac{\|\tilde{\mathbf{y}}\|_2^2 \|\mathbf{x}\tilde{\mathbf{y}}\|_\infty^2}{\|\tilde{\mathbf{y}}\|_1^2 \|\tilde{\mathbf{y}}\|_\infty^2 \|\mathbf{x}\|_\infty^2 \|\mathbf{x}\tilde{\mathbf{y}}\|_\infty^2} \\ &= x_j^2 y_j^2 \|\mathbf{x}\|_\infty^{-2} \frac{k(w+h)^2 + \zeta}{(k(w+h) + \theta)^2 (w+h)^2 (\eta(w+h)^2 + \nu)} \\ &= x_j^2 y_j^2 \|\mathbf{x}\|_\infty^{-2} \left(k + \frac{\zeta}{(w+h)^2} \right) \\ &\quad \cdot \frac{1}{(k(w+h) + \theta)^2} \cdot \frac{1}{\eta(w+h)^2 + \nu}, \end{aligned}$$

which is obviously decreasing in h . This proves case (II).

Acknowledgments

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Reference

- [1] D. Anglun, D. Eisenstat, L. (Aryeh) Kontorovich and L. Reyzin, Lower bounds on learning random structures with statistical queries, in *ALT*, Lecture Notes in Computer Science, Vol. 6331 (Springer, 2010), pp. 194–208.