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A lower bound on the sum of the largest components of a nonnegative vector

Charles Audet

GERAD & Département de mathématiques et de génie industriel, Polytechnique Montréal, Montréal (Québec) Canada, H3C 3A7

`charles.audet@gerad.ca`

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Abstract: We study the function returning the sum of the k components of largest magnitude of a vector. We show that if a nonnegative vector x is such that its Euclidean norm is greater than or equal to one, and if the integer k is an upper bound on the Manhattan norm of x , then there exists k components of x whose sum is greater than or equal to one.

Key Words: Sum of largest components of a vector, order statistics.

1 Introductory note

We consider the function that returns the sum of the largest k components of a vector in $\mathbb{R}^n$, where k is an integer between 1 and n . Formally, define the partial sum function $s_k : \mathbb{R}^n \rightarrow \mathbb{R}$ as

$$s_k(x) := \sum_{i=1}^k x_{[i]}$$

where $x_{[i]}$ denotes the i -th largest component of the vector $x \in \mathbb{R}^n$. When sorted in increasing order, the values $x_{[n]} \leq x_{[n-1]} \leq \dots \leq x_{[1]}$ are known as order statistics [3]. The textbooks [1, 2] show that this partial sum function s_k is convex by using elementary linear programming duality theory.

In the present note, we show that any $x \in \mathbb{R}_+^n$ whose Euclidean and Manhattan norms satisfy $\|x\|_2 \geq 1$ and $\|x\|_1 \leq k$ for some integer $k \leq n$, is such that the sum of the k largest components of x is at least one, i.e., $s_k(x) \geq 1$. The inequality is trivial when $k = 1$ because the only admissible vectors are the coordinate vectors. The inequality is also trivial when $k = n$ because $s_n(x) = \|x\|_1 \geq \|x\|_2$ for any nonnegative vector x .

2 Proof of the main result

The following Lemma is used twice in the demonstration of the main result.

Lemma 2.1 *Let $m \in \mathbb{N}$, $M = \{1, 2, \dots, m\}$, and α, β, γ be nonnegative real numbers satisfying $\alpha \leq \beta$. If a and b belong to $\Omega = \{c \in \mathbb{R}^n : \|c\|_1 = \gamma, \alpha \leq c_i \leq \beta \forall i \in M\}$, and if all the components of a , except for possibly one, are at their bound α or β then $b^T b \leq a^T a$.*

Proof. Let a and b in Ω be such that the components of a , except for possibly one, are either α or β . Choose $c \in \Omega$ so that $c^T c \geq d^T d$ for all $d \in \Omega$. In particular, we have $c^T c \geq b^T b$.

If there exists two indices $i \neq j$ of M such that $\alpha < c_i \leq c_j < \beta$, then $\delta := \min\{c_i - \alpha, c_j + \delta\} > 0$ satisfies $\alpha \leq c_i - \delta < c_j + \delta \leq \beta$ and

$$(c_i - \delta)^2 + (c_j + \delta)^2 = c_i^2 + c_j^2 + 2\delta(c_j - c_i) + 2\delta^2 > c_i^2 + c_j^2.$$

This situation would contradict the fact that $c^T c$ is maximal. Therefore, at most one component of c differs from α and β .

Furthermore, since $\|a\|_1 = \|c\|_1$ and $\alpha \leq a_i \leq \beta$ et $\alpha \leq c_i \leq \beta$ for all $i \in M$, it follows that a and c have the same entries, with possibly a different order. We conclude that $a^T a = c^T c \geq b^T b$. $\square$

We can now state and prove the principal result.

Theorem 2.2 *If $k \leq n$ is an integer such that $k \geq \|x\|_1$ for some $x \in \mathbb{R}_+^n$ with $\|x\|_2 \geq 1$, then $s_k(x) \geq 1$.*

Proof. Consider the pair $x \in \mathbb{R}_+^n$ and $k \in \mathbb{N}$ with $\|x\|_2 \geq 1$ and $k \geq \|x\|_1$. Without any loss in generality, and by reordering the components if necessary, assume that $x_1 \geq x_2 \geq \dots \geq x_n \geq 0$ and that $1 \leq k < n$. Furthermore, the case where $x_k = 0$ is trivial.

Set $v = x_k > 0$, and by an Euclidian division, find $p \in \mathbb{N}$ and $0 \leq q < v$ such that

$$pv + q = \sum_{i=k+1}^n x_i = \|x\|_1 - s_k(x).$$

Construct the vector $y \in \mathbb{R}_+^n$ as follows :

$$y_i = \begin{cases} s_k(x) - (k-1)v & \text{if } i = 1 \\ v & \text{if } 2 \leq i \leq p+k \\ q & \text{if } i = p+k+1 \\ 0 & \text{if } p+k+2 \leq i \leq n. \end{cases}$$

With this construction, we have $s_k(x) = s_k(y)$, and in addition, the one-norm of y satisfies

$$\begin{aligned} \|y\|_1 &= y_1 + (p+k-1)v + q = y_1 + (k-1)v + (pv+q) \\ &= \sum_{i=1}^k x_i + \sum_{i=k+1}^n x_i \\ &= \|x\|_1 \leq k. \end{aligned}$$

Multiplying this last inequality by the scalar $v > 0$, and by reordering the terms we get

$$-(pv+q)v \geq (y_1 - k + (k-1)v)v. \quad (1)$$

Let us now analyze the Euclidean norm of y . On the one hand, both vectors $(y_1, y_2, \dots, y_k) = (y_1, v, v, \dots, v)$ and $(x_1, x_2, \dots, x_k)$ belong to the set $\Omega \subset \mathbb{R}^k$ of Lemma 2.1 with $\alpha = v \leq \beta = y_1$ and $\gamma = \sum_{i=1}^k x_i$. In addition, all components of the first vector are at their bounds and consequently

$$\sum_{i=1}^k x_i^2 \leq \sum_{i=1}^k y_i^2. \quad (2)$$

On the other hand, both vectors $(y_{k+1}, y_{k+2}, \dots, y_n) = (v, v, \dots, q, 0, 0, \dots, 0)$ and $(x_{k+1}, x_{k+2}, \dots, x_n)$ belong to $\Omega \subset \mathbb{R}^{n-k}$ of Lemma 2.1 with $\alpha = 0 < \beta = v$ et $\gamma = pv+q$. Once again, all components of the first vector, except possibly the one whose value is q , are at one of their bounds. It follows that

$$\sum_{j=k+1}^n x_j^2 \leq \sum_{j=k+1}^n y_j^2. \quad (3)$$

Combining Equations (2) and (3) with the assumption on the Euclidean norm of x leads to

$$\|y\|_2^2 = y_1^2 + (p+k-1)v^2 + q^2 \geq \|x\|_2^2 \geq 1, \quad (4)$$

and therefore

$$\begin{aligned} y_1^2 + (k-1)y_1v &\geq y_1^2 + (k-1)v^2 && \text{since } y_1 \geq v \\ &\geq 1 - pv^2 - q^2 && \text{by Equation (4)} \\ &\geq 1 - (pv+q)v && \text{since } v \geq q \\ &\geq 1 + (y_1 - k + (k-1)v)v && \text{by Equation (1).} \end{aligned} \quad (5)$$

Finally, consider the quantity

$$\begin{aligned} (y_1 + 1 - v)(y_1 + (k-1)v - 1) &= y_1^2 + (k-1)y_1v - y_1 + (1-v)(y_1 + (k-1)v - 1) \\ &\geq 1 + (y_1 - k + (k-1)v)v \\ &\quad - y_1 + (1-v)(y_1 + (k-1)v - 1) && \text{by (5)} \\ &= 1 + y_1v - kv + (k-1)v^2 \\ &\quad - y_1 + y_1 + (k-1)v - 1 - y_1v - (k-1)v^2 + v \\ &= 0. \end{aligned}$$

However, $y_1 \geq v$ implies that both $(y_1 + 1 - v)$ and $(y_1 + (k-1)v - 1)$ are nonnegative, and

$$s_k(x) = s_k(y) = y_1 + (k-1)v \geq 1.$$

□

The Theorem may be trivially stated in $\mathbb{R}^n$ rather than $\mathbb{R}_+^n$ as follows. If $k \leq n$ is an integer such that $k \geq \|x\|_1$ for some $x \in \mathbb{R}^n$ with $\|x\|_2 \geq 1$, then $s_k(|x|) \geq 1$, where $|x|$ denotes the vector in $\mathbb{R}^n$ whose components are the absolute values of those of x .

As a final coment, observe that the Theorem cannot be made tighter since for any positive integer $k \leq \sqrt{n}$, the vector $x \in \mathbb{R}_+^n$ defined as

$$x_i = \begin{cases} \frac{1}{k} & \text{if } i \in \{1, 2, \dots, k^2\} \\ 0 & \text{if } i \in \{k^2 + 1, k^2 + 2, \dots, n\} \end{cases}$$

satisfies $\|x\|_2 = 1$, $\|x\|_1 = k$ and $s_k(x) = 1$.

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