OPTIMIZERS FOR SUB–SUMS SUBJECT TO A SUM– AND A SCHUR–CONVEX CONSTRAINT WITH APPLICATIONS TO ESTIMATION OF EIGENVALUES

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Abstract. A complete solution is presented for the problem of determining the sets of points at which the functions $(x_1, \ldots, x_n) \mapsto x_k + \ldots + x_l$, subject to the constraints $M \ge x_1 \ge \ldots \ge$ $x_n \ge m$, $x_1 + x_2 + ... + x_n = a$, and $g(x_1) + g(x_2) + ... + g(x_n) = b$, with g strictly convex continuous, assume their maxima and minima. Applications are given.

0. Introduction

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Let $m, M \in \overline{\mathbb{R}} = \mathbb{R} \cup \{-\infty, +\infty\} = [-\infty, +\infty], \ a, b \in \mathbb{R}, \text{ and } n \in \mathbb{Z}_{\geqslant 1}.$ **0. Introduction**

Let $m, M \in \overline{\mathbb{R}} = \mathbb{R} \cup \{-\infty, +\infty\} = [-\infty, +\infty]$, $a, b \in \mathbb{R}$, and $n \in \mathbb{Z}_{\ge 1}$.

For an extended function $g : \mathbb{R} \to]-\infty, +\infty]$ and integers $1 \le k \le l \le n$, define $S_g^{kl}(\mathbf{x}) = \sum_{i=1}^{l} g(x_i)$. We also use $\sum_{i=k} g(x_i)$. We also use the abbreviations $S_g = S_g^{1n}$, $S^{kl} = S_{id}^{kl}$, and $S = S^{1n}$. For convex continuous $g : \mathbb{R} \to]-\infty, +\infty]$, define the spaces $+\infty$ } = [- ∞ , + ∞], *a*, *b*
- ∞ , + ∞] and integers 1
abbreviations $S_g = S_g^{1n}$, S^k
 ∞ , + ∞], define the spaces

$$
X=X(m,M,a,b;n;g)=\{x\in\mathbb{R}^n:M\geqslant x_1\geqslant x_2\geqslant\ldots\geqslant x_n\geqslant m, S(x)=a, S_g(x)=b\}.
$$

The main objective of this paper is the determination of the subsets of *X* where the functions S^{kl} assume their maxima and minima (called also maximum sets or maximizers, and minimum sets or minimizers) via an elementary technique; i.e. avoiding Karush-Kuhn-Tucker theory, see e.g. [3]. The calculation of upper and lower bounds for the functions S^{kl} , of necessity always the best possible, is then a simple matter of evaluation of the functions at points in their maximizers and minimizers.

This article puts many of the results estimating various functions of eigenvalues and singular values in terms of the trace and/or determinant etc. (e.g. $[5]$, $[9]$, $[10]$, $[11]$, $[12]$, $[13]$, $[14]$, $[17]$, $[18]$), in particular those in the last four papers mentioned, under a general umbrella; many bounds obtained by Wolkowicz and Styan $[17]$, $[18]$ are implied by our results.

Section 1 recalls some basics of the modern theory of convex functions; facts yielding information about the spaces $X(m, M, *)$, in particular also for $n = 3$ via generalized barycentric coordinates in the plane, are also proved. In section 2, after

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presenting in lemma 6 the heart of our technique and theorem 7 that implicitly contains an algorithm to test when the spaces X are nonempty, we show that for all integers $1 \leq k \leq l \leq n$ and $m, M \in \overline{\mathbb{R}}$ the sets minimizer $(S^{kl}) = \{ \mathbf{x} \in X : S^{kl}(\mathbf{x}) = \inf \{ S^{kl}(\mathbf{u}) : S^{kl}(\mathbf{u}) > 0 \}$ $\mathbf{u} \in X$ } and maximizer $(S^{kl}) = \{ \mathbf{x} \in X : S^{kl}(\mathbf{x}) = \sup\{S^{kl}(\mathbf{u}) : \mathbf{u} \in X\} \}$ consist of elements having a simple structure. Indeed theorems 8 and 9 show that the minimization or maximization of the functions *Skl* can be reduced to that of solving equations in two unknowns or to that of finding the intersections of a plane convex curve with a straight line. The extent to which these equations can be explicitly solved depends of course on the convex function *g*. The brief section 3 indicates that our theorems allow to calculate explicitly the infimum and supremum of certain subsets of *X* in the lattice induced by majorization order on *X* In section 4 we show how our results can be used to estimate partial sums of the eigenvalues of matrices having real spectra, given on the matrix additional information like trA and tr A^2 ; trA and tr A^3 ; trA and det A; etc. This way we gain deeper understanding of results found earlier by the second and fourth authors on this subject.

Concerning presentation, we make differentiability assumptions on the convex function *g*. These are made for simplicity, since they do not impose restrictions on our applications, since they broaden accessibility of the paper, and since at spots they lead to further illuminating observations. As far as we see, the main theorems 8 and 9 can be stated and proved without differentiability assumptions following the indications in the concluding section 5. Though the basic idea of our method is simple, we apologize for that there are sometimes many cases to consider, the number and implications of which are at times best controlled by applying a modest amount of the symbology of logic. Our objects frequently depend on many parameters. To lighten notation, we usually suppress a fair amount of these; the context provides the remaining ones. Finally, for precision we indicate references sometimes in ways like '[2], $p12c-5$ ' meaning 'see [2] page 12, approximately 5 centimeters from the last text row'.

1. Basic Notations, Definitions, and Lemmas

a. Some of the symbols and conventions we use are these:

b. Concerning convex functions it simplifies matters if we adopt definitions used in modern texts on convex analysis or optimization; foremost we mention Rockafellar's definitive treatise $[16]$ and the recent book of Borwein and Lewis $[3]$. Let E be an Euclidean space and $g: E \to [-\infty, +\infty]$ be a real valued function on E. The *domain* of *g* is dom $g = \{x \in E : g(x) < \infty\}$; *g* is called *proper* if dom $g \neq \emptyset$ and for no $x \in E$, $g(x) = -\infty$, see [3], p44c2; [16], p23c4. The function *g* is *convex* if its *epigraph* epi $g = \{(x, r) \in E \times \mathbb{R} : g(x) \leq r\}$ is a convex set, see [3], p43c-2; [16], p23; strictly convex if its epigraph is strictly convex, see [7], p98c-2. Proper convex functions in this sense have convex domains and satisfy for $x, y \in E$ the usual inequality $g(\lambda x + (1 - \lambda)y) \leq \lambda g(x) + (1 - \lambda)g(y)$, see [3], p46c6; strictly convex functions the strict inequality. As we see, every convex/strictly convex function $g: I \to \mathbb{R}$ (*I* is an interval) in the traditional sense (satisfying the inequality/strict inequality) can be trivially extended to $\mathbb R$ putting its values outside *I* equal to $+\infty$ and thus viewed as a convex function in the sense here defined, see $[16]$, p23c-3.

PROPOSITION 1. Let $k \in \mathbb{Z}_{\geqslant 1}$, $\alpha_1, \ldots, \alpha_k \in \mathbb{R}_{>0}$, $a, b \in \mathbb{R}$. Assume the function $g : \mathbb{R} \to [-\infty, +\infty]$ *satisfies the following: g is proper and strictly convex, dom g is open, and g is differentiable and unbounded on* dom*g Let* $\mu = \frac{a}{k}$, $\mathbf{m} = \mu$ α*i ,* $\mathbf{m} = \mu \mathbf{1}_k$ *, and*

let $\mathscr{S} = \mathscr{S}((\alpha_1, \ldots, \alpha_k), a, b)$ *be the set of solutions of the system of two equations*

i.
$$
S(\mathbf{x}) = \sum_{i=1}^{k} \alpha_i x_i = a
$$
, ii. $G(\mathbf{x}) = \sum_{i=1}^{k} \alpha_i g(x_i) = b$.

Then $H = \{x : x \text{ solves (i)}\}$ *is a hyperplane and* $\mathscr S$ *is either empty/the point* $\{m\}$

/homeomorphic to a sphere in H according to if $G(m) > b/G(m) = b/G(m) < b$ *.*

Proof. It is clear that *H* is a hyperplane containing **m**. We can write $H = \{m + t\mathbf{r} :$ $t \in \mathbb{R}, \mathbf{r} = (r_1, \ldots, r_k) \in \mathbb{R}^k, \sum_{i=1}^k \alpha_i r_i = 0, ||\mathbf{r}|| = 1$ *I*. *k.* **MERIROSKI, O. PIRHURKO, A. VIRTANEN
** *H* **according to if** $G(\mathbf{m}) > b/G(\mathbf{m}) = b/G(\mathbf{m}) < b$ **.

s a hyperplane containing m**. We can write $H = \{\mathbf{m}+t\mathbf{r}:$
 $\sum_{i=1}^{k} \alpha_i r_i = 0, ||\mathbf{r}|| = 1\}$. Every $\mathbf{x} \in H \setminus {\$ representation $\mathbf{x} = \mathbf{m} + t(\mathbf{x})\mathbf{r}(\mathbf{x})$ with $t(\mathbf{x}) > 0$, and thus we can view *G* henceforth as defined by $G(\mathbf{m} + t\mathbf{r}) = \sum_{k}^{k} \alpha_{i}g(\mu + t r_{i})$ as a f $\sum_{i=1} \alpha_i g(\mu + tr_i)$ as a function on the Euclidean space *H*.

Claim 0. *G* is strictly convex on *H*. Let **x**, **y** be two points in dom *G* and $\lambda \in]0,1[$. Then

$$
G(\lambda \mathbf{x} + (1 - \lambda)\mathbf{y}) = \sum_{i=1}^{k} \alpha_i g(\lambda x_i + (1 - \lambda)y_i)
$$

\$\leqslant \sum_{i=1}^{k} \alpha_i (\lambda g(x_i) + (1 - \lambda)g(y_i))\$
\$\leqslant \lambda G(\mathbf{x}) + (1 - \lambda)G(\mathbf{y}).\$

Since we are supposing $\mathbf{x} \neq \mathbf{y}$, there exists *i* such that $x_i \neq y_i$; hence the strict convexity of g implies the strictness of the inequality.

Claim 1. dom $G \neq \emptyset \Leftrightarrow m \in \text{dom } G \Leftrightarrow \mu \in \text{dom } g$. \Box The implications \Leftarrow are clear. Now let $t \in \mathbb{R}$ be such that $\mathbf{m} + t\mathbf{r} \in \text{dom } G$. There are *i, j* such that $r_i \leq 0 \leq r_i$, and so we get $g(\mu + tr_i) < +\infty$ and $g(\mu + tr_i) < +\infty$. If $t \geq 0$, we have $\mu + tr_i \leq \mu \leq \mu + tr_i$, and the reverse inequalities in case $t \leq 0$. In any case the convexity of dom *g* insures $\mu \in g$, as we wished to show.

Claim 2. dom *G* is open. \int If dom *G* is empty, there is nothing to prove. Suppose now dom $G \neq \emptyset$ and let $\mathbf{x} \in H$. Clearly $\mathbf{x} \in \text{dom } G$ iff, for all *i*, $x_i \in \text{dom } g$. Since dom *g* is open, there is $\varepsilon > 0$ such that $x_i + \cdot$ $-\varepsilon$, $+\varepsilon$ \subseteq dom *g* for all *i*. But then the open set $(\mathbf{x} + \mathbf{I} - \varepsilon, +\varepsilon \mathbf{I}^n) \cap H$ in *H* is an open neighborhood of **x** in *H*, contained in $dom G \trianglelefteq$

Claim 3. If dom $G \neq \emptyset$, then minimizer $(G) = {\bf{m}}$. [By claims 1 and 2, **m** and μ are interior points of dom G and dom g respectively. Hence the calculation $\frac{d}{dt}\tilde{G}(\mathbf{m} + t\mathbf{r})|_{t=0} = \sum_{i} \alpha_i r_i g'(\mu) = 0$ is well defined and shows that **m** is a critical point. Since *G* is a strictly convex function, **m** is the unique minimizer of *G*, see [3], p16c-7, 19c-5. \triangleleft

Claim 4. For any $r \in \mathbb{R}$, the level sets $L_r = \{ \mathbf{x} \in H : G(\mathbf{x}) \leq r \}$ are convex compact subsets of dom *G*. If $r > G(m)$, then $m \in int(L_r) = \{x : G(x) < r\}$. [We can assume $L_r \neq \emptyset$. By claims 2 and 3, there exists $\rho > 0$ such that the sphere ∂B , $B = B(\mathbf{m}, \rho) \subset H$, pertains to dom *G*. Let $d(\mathbf{r}) = G(\mathbf{m} + \rho \mathbf{r}) - G(\mathbf{m}) > 0$. The function *d* assumes on ∂B its minimum, say at point \mathbf{r}_0 . From an application of the formulae in $\left[15\right]$, p98c1 or $\left[16\right]$, p242c2 we get

$$
G(\mathbf{m} + t\mathbf{r}) \geqslant G(\mathbf{m} + \rho \mathbf{r}) + (t - \rho) \frac{dG}{dt}\Big|_{t=\rho}
$$

$$
\geqslant G(\mathbf{m}) + d(\mathbf{r}) + (t - \rho) \frac{d(\mathbf{r})}{\rho}
$$

$$
= G(\mathbf{m}) + \frac{t}{\rho}d(\mathbf{r})
$$

\n
$$
\geq G(\mathbf{m}) + \frac{t}{\rho}d(\mathbf{r}_0).
$$

Choose t_0 such that $G(\mathbf{m}) + \frac{t_0}{\rho}d(\mathbf{r}_0) = r$. We see: for all $\mathbf{x} \in H$ with $\mathbf{x} \notin B(\mathbf{m}, t_0)$, there holds $G(\mathbf{x}) > r$. Hence the set L_r is bounded. The remaining claims follow from combining [16], p28c-2, 51c-3, 52c-8, 59c2 or can otherwise be left to the reader. $\frac{1}{2}$

We can now conclude the proof. Case: $G(m) > b$. If $m \in \text{dom } G$, then by claim 3, for all $\mathbf{x} \in H$, $G(\mathbf{x}) \geq G(\mathbf{m}) > b$. If $\mathbf{m} \notin \text{dom } G$, then claim 1 says $G(\mathbf{x}) = +\infty$ for all $\mathbf{x} \in H$, and we come to the same conclusion. In any case we see $\mathscr{S} = \emptyset$. Case: $G(\mathbf{m}) = b$. Then claim 3 gives $\mathcal{S} = {\mathbf{m}}$. Case: $G(\mathbf{m}) < b$. Then, using $\mathscr{S} = {\mathbf{x} : G(\mathbf{x}) \leq b} \setminus {\mathbf{x} : G(\mathbf{x}) < b}$, we have by ([16], p59c1) that $\mathscr{S} = \partial L_b$. Now [2], corollary 11.3.4 completes the proof. \Box

LEMMA 2. Assume the hypotheses of proposition 1 specialized to the case $k = 3$ *and* $G(m) < b$. Then $\mathscr{S} = \mathscr{S}((\alpha, \beta, \gamma), a, b)$ *is a convex, rectifiable, smooth curve in H. If s* \mapsto $(x(s), y(s), z(s))$ *is a regular parametric representation (e.g. w.r.t. curve length), then with a dot denoting differentiation w.r.t. s : for each s with* $\dot{x} \cdot \dot{y} \cdot \dot{z} \neq 0$ there holds $sgn(\dot{x}, \dot{y}, \dot{z}) \in \{(+,-, +), (-, +, -)\}$.
 b. *for each s with* $\dot{x} \cdot \dot{y} \cdot \dot{z} \neq 0$ there holds $sgn(\dot{x}, \dot{y}, \dot{z}) \in \{(+, -, +), (-, +, -)\}$.

a. $\dot{x} = 0 \Leftrightarrow z = y$, $\dot{y} = 0 \Leftrightarrow x = z$, $\dot{z} = 0 \Leftrightarrow x = y$;

Proof. a. It is known that the boundary of a compact convex plane region is rectifiable. Thus we obtain from the previous lemma that we can parametrize $\mathscr S$ in the form $s \mapsto (x(s), y(s), z(s))$, where *s* is the arc length measured from a certain point onwards. (In a cartesian interpretation of coordinates, $\dot{x}^2 + \dot{y}^2 + \dot{z}^2 \equiv 1$.) Since the curve $\mathscr S$ satisfies (i) and (ii) of proposition 1, we find upon differentiation the relations

$$
0 = \alpha \dot{x} + \beta \dot{y} + \gamma \dot{z},
$$

\n
$$
0 = \alpha g'(x)\dot{x} + \beta g'(y)\dot{y} + \gamma g'(z)\dot{z}
$$

\n
$$
= (\alpha \dot{x} + \beta \dot{y} + \gamma \dot{z})g'(z) + \alpha \dot{x}(g'(x) - g'(z)) + \beta \dot{y}(g'(y) - g'(z))
$$

\n
$$
= \alpha \dot{x}(g'(x) - g'(z)) + \beta \dot{y}(g'(y) - g'(z)).
$$
\n(2)

Note that the nonvanishing of the tangent vector and (1) imply that for each s at most one of the quantities $\dot{x}(s)$, $\dot{y}(s)$, $\dot{z}(s)$ vanishes. Thus if $\dot{x} = 0$, we find $g'(y) - g'(z) = 0$; hence, by monotonicity of g' ([15], p10c-1), $y = z$ follows. Conversely, and using that $x = y = z$ is impossible, we get that $y = z$ implies $\dot{x} = 0$. Proceeding analogously for the pairs x, z and x, y proves (a).

b. Assume $x > y > z$. Then $g'(x) > g'(y) > g'(z)$. Thus we see that $\text{sgn}(x)\text{sgn}(y) = -1$. Interchanging in the calculation leading to (2) the roles of $\alpha g'(x)\dot{x}$ and $\gamma g'(z)\dot{z}$, we get similarly that $sgn(\dot{z})sgn(\dot{z}) = -1$. Similar considerations can obviously be made in all cases where $x \neq y \neq z \neq x$. \Box

It is instructive and for our later applications indeed useful to interpret the formulae of lemma 2 geometrically. Select three noncollinear points *X Y Z* in a plane. We coordinatize its points by triples (x, y, z) satisfying $\alpha x + \beta y + \gamma z = a$ via the better

known barycentric coordinates $(\hat{x}, \hat{y}, \hat{z})$ w.r.t. *X*, *Y*, *Z* (see [2], p81) via the formulae *X*, *Y*, \ddot{Z} and parameters α , β , γ , α .

 $\hat{x} = \frac{\alpha}{a}x$, $\hat{y} = \frac{\beta}{a}y$, $\hat{z} = \frac{\gamma}{a}z$. We call (x, y, z) generalized barycentric coordinates w.r.t.

X, Y, Z and parameters α , β , γ , a.

LEMMA 3. In a generalized barycentric system with parameters LEMMA 3. In a generalized barycentric system with parameters α , β , γ , α , an *equation* $cx + dy + ez = f$ with $(0, 0, 0) \nsim (c, d, e) \nsim (\alpha, \beta, \gamma)$ represents a line *through points* $Z' \in \{z = 0\}$ *and* $Y' \in \{y = 0\}$ *(see the rhs-figure) located so that*

$$
\overrightarrow{XZ'} = \frac{\beta}{a} \cdot \frac{ac - \alpha f}{\beta c - \alpha d} \overrightarrow{XY}, \qquad \overrightarrow{XY'} = \frac{\gamma}{a} \cdot \frac{ac - \alpha f}{\gamma c - \alpha e} \overrightarrow{XZ}.
$$

Proof. Expressed in barycentric coordinates, $cx + dy + ez = f$ turns into $\frac{ac}{\alpha}\hat{x} +$ $\frac{ad}{\beta}\hat{y} + \frac{ae}{\gamma}\hat{z} = f$. Given our hypotheses, this represents a line for it is interpretable as the intersection of two planes. Since $\hat{x} + \hat{y} + \hat{z} = 1$, one calculates that this line intersects $\{z = 0\} = \{\hat{z} = 0\}$ at a point *Z'* whose barycentric coordinates are given by $\hat{x}(Z') = \frac{\alpha \beta f - \alpha ad}{\beta ac - \alpha ad}$, $\hat{y}(Z') = \frac{\beta ac - \alpha \beta f}{\beta ac - \alpha ad}$. We see (e.g. applying [2], p295c and elementary triangle geometry, or otherwise) that if *P* is any point on the line $\hat{z} = 0$, then $\overrightarrow{XP} = \hat{y}(P) \cdot \overrightarrow{XY}$. From this follows the first assertion; the second is similar. \Box

EXAMPLE 4. Consider the lhs-figure where we have chosen points X, Y, Z to define a triangle of unit side length. In such a system the coefficients of \overrightarrow{XY} , \overrightarrow{XZ} are interpretable as signed distances from X to Z' and X to Y' respectively.

The rest of the figure is to be interpreted with $(\alpha, \beta, \gamma, a) = (2, 5, 3, 1)$. The ${y = 0}$ that define the triangle; the cevians $x = y$ through *Z*, $x = z$ through *Y*, $z = y$

through *X*, and the line $3x + y = 0.5$. The figure shows further a convex curve which we assume to be the solution of some equation $2g(x) + 5g(y) + 3g(z) = b$ for a certain convex *g*. According to lemma 2, at the point of intersection with $x = y$, $y = z$, $z = x$ the tangents to $\mathscr S$ are parallel to segments *XY*, *YZ*, and *ZX* respectively. We further find a shadowed open region in which the coordinates of any point satisfy $x > y > z$. There are six such regions defined solely by the lines through the point $x = y = z$. The arrows indicate increase or decrease of coordinates *x*, *y*, *z* of a point moving along $\mathscr S$ in that region counterclockwise. The lines $x = 0$, $y = 0$, $z = 0$ define seven disjoint regions. The function $P \mapsto \text{sgn}(x, y, z)$ is constant in each of these regions. We have indicated the values of this function for each of the regions by $- + +$, etc. in the rhs-figure.

LEMMA 5. Assume the hypotheses and notation of lemma 2. Define for $m, M \in$ $[-\infty, +\infty]$ the set $\mathscr{S}' = \mathscr{S}'(m, M) = \{(x, y, z) \in \mathscr{S}((\alpha, \beta, \gamma), a, b) : M \geq x \geq 0\}$ $y \ge z \ge m$ }. Let $(\xi', \eta') \in \mathscr{S}((\alpha, \beta + \gamma), a, b)$ with $\xi' \ge \eta'$ and $(\xi, \eta) \in$ $\mathscr{S}((\alpha + \beta, \gamma), a, b)$ with $\xi \geq \eta$. Let

$$
(x_L, y_L, z_L) = \begin{cases} (\xi', \eta', \eta') & \text{if } \xi' \le M, \\ (M, y, z) \in \mathcal{S}' & \text{if } \xi' > M, \end{cases}
$$

$$
(x_R, y_R, z_R) = \begin{cases} (\xi, \xi, \eta) & \text{if } \eta \ge m, \\ (x, y, m) \in \mathcal{S}' & \text{if } \eta < m. \end{cases}
$$

These points are all uniquely defined. Assume c, d, e $\in \mathbb{R}_{\geq 0}$ *satisfy, in addition to the hypothesis of lemma 3, the inequality* $\frac{c}{\alpha}$ > max $\{\frac{d}{\beta}, \frac{e}{\gamma}\}\$. Put $\varepsilon = \text{sgn}\left(\frac{\gamma}{\beta} \frac{\beta c - \alpha d}{\gamma c - \alpha e}\right)$ β*c*^α*d* $\gamma c - \alpha e$ $\frac{\alpha}{\beta} \frac{x_L - x_R}{y_R - y_L} - 1$). Then

$$
\text{minimize} (cx + dy + ez | \mathcal{S}'(m, M)) = \begin{cases} \{(x_R, y_R, z_R)\} & \text{if } \varepsilon = +1, \\ \{(x_L, y_L, z_L), (x_R, y_R, z_R)\} & \text{if } \varepsilon = 0, \\ \{(x_L, y_L, z_L)\} & \text{if } \varepsilon = -1. \end{cases}
$$

Proof. We reason along the lines of the figures. If *M* is sufficiently large or *m* sufficiently small, then $\mathscr{S}'(m, M)$ is just the part of \mathscr{S} lying in the shadowed region $x \ge y \ge z$ of the lhs-figure and thus limited at the left by a point (x, y, z) with $y = z$, coinciding evidently with the point (ξ', η', η') formed from the solution (ξ', η') with $\xi' \geq \eta'$ in $\mathscr{S}((\alpha, \beta + \gamma), a, b)$; and on the right by a point with $x = y$ formed similarly from the decreasing solution (ξ, η) in $\mathcal{S}((\alpha + \beta, \gamma), a, b)$. If *M* is smaller than the *x*-coordinate of the point in $\mathscr S$ satisfying $x \geq y = z$ and/or *m* larger than the *z*-coordinate of the point in $\mathscr S$ satisfying $x = y \ge z$, then upon watching the changes of *x* - and *z*-coordinates of a point moving in the shadowed region, we see that the arc is defined on the left by a point of the form $(M, *, *)$ and/or on the the right by a point of the form $(*,*,m)$, and the claim follows again – the rhs-figure tries to illustrate these possibilities. the arc is defined on the left by a point inoving in the shadowed region, we see that
the arc is defined on the left by a point of the form $(M, *, *)$ and/or on the the right
by a point of the form $(*, *, m)$, and the claim foll

Let us now define the 'steepness' of a line through two points $Y' \in \{y = 0\}$,

 $cx + dy + ez = f$ is *any* line, then lemma 3 yields the formula steepness $(l_f) = \frac{\gamma(\beta c - \alpha d)}{\beta(\gamma c - \alpha e)}$. Of course the steepness does not depend on *f*

We now assume c, d, e to have the meaning and satisfy the hypotheses reserved for them. Define for real *f* the line $l_f = \{cx + dy + ez = f\}$. The hypotheses ensure that the denominators figuring in the expressions given in lemma 3 for $\overrightarrow{XZ'}$ and $\overrightarrow{XY'}$, $Z' = Z'(f)$, $Y' = Y'(f)$ – these being the points of intersections of the line l_f with $z = 0$ and $y = 0$ respectively – are all positive. If $f = 0$, the fractions are indeed 1. We see that as f increases, Y' and Z' wander toward X . Let \overline{l} be the line through points (x_L, y_L, z_L) and (x_R, y_R, z_R) . It is a consequence of lemma 2a and the convexity of arc \mathscr{S}' that we always shall have $0 <$ steepness $(\bar{l}) < 1$. The minimizer of the function $(x, y, z) \mapsto cx + dy + ez$ restricted to \mathscr{S}' is the subset of \mathscr{S}' that line l_f hits first as f increases. This set consists of the point (x_R, y_R, z_R) only if steepness (l_f) > steepness (\bar{l}) , of both points (x_L, y_L, z_L) , (x_R, y_R, z_R) if the steepnesses are equal, and of point (x_R, y_R, z_R) only if steepness (l_f) < steepness (\bar{l}) . The line \bar{l} obeys the equation $(y_R - y_L)x + (x_L - x_R)y + 0z + (x_Ry_L - x_Ly_R) = 0$. Calculating its steepness, the claim follows. \Box

2. Maximizers and Minimizers

This section presents the main results; unfortunately many of these results are technical at first reading. However, just as is the case with lemma 5 of the previous section, large chunks of the technicalities evaporate for spaces in which $m = -\infty, M =$ $+\infty$; so the reader is invited to always reflect first what happens in this case.

Fix a space $X = X(m, M, a, b; n; g)$ with g as in proposition 1. It will be sometimes convenient to think of $\mathbf{x} = (x_1, \ldots, x_n) \in X$ as augmented by $x_0 = M$, $x_{n+1} = m$ section, large chunks of the technicalities evaporate for spaces in which $m = -\infty$, $M = +\infty$; so the reader is invited to always reflect first what happens in this case.
Fix a space $X = X(m, M, a, b; n; g)$ with g as in proposit indicate the position of a descent at *i* in a form like $\left(\ldots \right)$ and write descents(**x**) for the set of descents of \bf{x} (in X). We extend this definition in a natural manner to subtuples $\mathbf{x}(k : l)$. For example, if $\mathbf{x} = (3, 3, 1, 1, -2) \in X(-5, 3, 6, *, 5; g)$, then descents $(\mathbf{x}) = \{2, 4, 5\}$, and descents $(\mathbf{x}(3:4))$ is $\{0, 2\}$ in terms of relative positions and $\{2, 4\}$ in terms of absolute ones. No confusion will arise.

Let $\mathbf{x} \in X(m, M, *)$. If there exists $\mathbf{i} = (i_0, \ldots, i_5) \in \mathbb{Z}^6$ satisfying $0 \leq i_0 < i_1 \leq i_2$ $i_2 < i_3 \leq i_4 < i_5 \leq n$, and *r*, *s*, $t \in \mathbb{Z}_{>0}$ and *x*, *y*, $z \in \mathbb{R}$ such that **x** can be thought of as having the form Let $\mathbf{x} \in X(m, M, *)$. If there $i_2 < i_3 \le i_4 < i_5 \le n$, and $r, s, t \in$
as having the form
 $\bullet \mathbf{x} = \mathbf{x}(x, y, z) = (\dots \stackrel{i_0}{\circ})$
amenable to an $\uparrow \downarrow \uparrow - \mathbf{i}$ -motion; i

• $\mathbf{x} = \mathbf{x}(x, y, z) = (\dots \overset{i_0}{\geq} x \mathbf{1}_r \overset{i_1}{\geq} \dots \overset{i_2}{\geq} y \mathbf{1}_s \overset{i_3}{\geq} \dots \overset{i_4}{\geq} z \mathbf{1}_t \overset{i_5}{\geq} \dots)$, then **x** is **i**-motion; i.e. to a replacement $\mathbf{x} \mapsto \mathbf{x}' = \mathbf{x}(x', y', z')$ with $x' > x, y' < y, z' > z$ such that $\mathbf{x}' \in X$; • $\mathbf{x} = \mathbf{x}(x, y, z) = (\dots \overset{t_0}{>}$
amenable to an $\uparrow \downarrow \uparrow - \mathbf{i}$ -motion; i
 $x' > x, y' < y, z' > z$ such that \mathbf{x}'
• $\mathbf{x} = \mathbf{x}(x, y, z) = (\dots \overset{i_0}{>}$
amenable to an $\downarrow \uparrow \downarrow - \mathbf{i}$ -motion; i

• $\mathbf{x} = \mathbf{x}(x, y, z) = (\dots \geq 1) \geq x \mathbf{1}_r \geq \dots \geq y \mathbf{1}_s \geq \dots \geq z \mathbf{1}_t \geq \dots),$ then **x** is **i**-motion; i.e. to a replacement $\mathbf{x} \mapsto \mathbf{x}' = \mathbf{x}(x', y', z')$ with $x' < x, y' > y, z' < z$ such that $\mathbf{x}' \in X$ $x, y' \le y, z' > z$ such that **x**
 $\bullet \mathbf{x} = \mathbf{x}(x, y, z) = (\dots \ge 0, z)$

suable to an $||\uparrow|$ - **i**-motion;
 $x, y' > y, z' < z$ such that **x**

Note that $||\uparrow|$ - **i**-movability

Note that $\|\cdot\|$ – **i**-movability of **x** for some **i** is implied by (and implies) the existence of $i_0, i_3 \in$ descents(**x**) such that $0 \le i_0 < i_0 + 2 \le i_3 \le n - 1$. In this case

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we can find *x*, *y*, *z*, *r*, *s*, *t*, *i*₁, *i*₂, *i*₄, *i*₅ (usually in many ways) such that **x** is $||$ $||$ - **i**movable with $\mathbf{i} = (i_0, \ldots, i_5)$. Similarly there exist $i_1, i_5 \in \text{descents}(\mathbf{x})$ with $1 \leq i_2 <$ $i_2 + 2 \leq i_5 \leq n$ if and only if these can be completed to an $\mathbf{i} = (i_0, \ldots, i_5)$ (usually We can find $x, y, z, r, s, t, i_1, i_2, i_4, i_5$ (usually in
movable with $\mathbf{i} = (i_0, \dots, i_5)$. Similarly there exity $i_2 + 2 \leq i_5 \leq n$ if and only if these can be com
in various ways) such that **x** is $||\cdot|| - \mathbf{i}$ -movable. in various ways) such that **x** is $|||| = i$ -movable. A complete specification of a motion would make necessary also the indication of one of the amounts $x' - x$, $y' - y$, $z' - z$, but these quantities are irrelevant for us. Of course **i** and r , s , t depend on each other; namely we have $i_1 - i_0 = r$, $i_3 - i_2 = s$, $i_5 - i_4 = t$. As with descents, movability depends via the values of *m* and *M* on the space *X* to which we refer. Context will avoid confusions.

LEMMA 6. For integers k, l with $1 \leq k \leq l \leq n-1$, consider a space $X =$ $X(m, M, a, b; n; g)$ and let $\mathbf{x} \in X$

a. *If there exist* $i_0, i_3 \in$ descents(**x**) with $0 \le i_0 \le k - 1, l + 1 \le i_3 \le n - 1$, then $\mathbf{x} \notin \text{maximize}(S^{kl})$.

b. If there exist $i_0, i_3 \in$ descents(**x**) with $0 \le i_0 \le k - 2, k \le i_3 \le n - 1$, then $\mathbf{x} \notin$ minimizer*(S^{kl})*.

b'. If there exist $i_0, i_3 \in$ descents(**x**) with $0 \le i_0 < i_0 + 2 \le i_3 \le l$, then $\mathbf{x} \notin$ minimizer (S^{1l}) .

c. If there exist $i_2, i_5 \in$ descents(**x**) with $1 \leq i_2 \leq l-1, l+1 \leq i_5 \leq n$, then $\mathbf{x} \notin \text{maximize}(S^{kl})$.

d. If there exist $i_2, i_5 \in$ descents(**x**) with $1 \leq i_2 \leq k - 2, l \leq i_5 \leq n$, then $\mathbf{x} \notin$ minimizer (S^{kl}) .

d'. If there exist $i_2, i_5 \in$ descents(**x**) with $l \leq i_2 < i_2 + 2 \leq i_5 \leq n$, then $\mathbf{x} \notin$ minimizer (S^{1l}) .

Proof. a. Choose $i_0 = \max\{i \in \text{descents}(\mathbf{x}) : i \leq k - 1\}$. Then (check that it is possible to) complete i_0 , i_3 to $\mathbf{i} = (i_0, \ldots, i_5)$ via choices $i_1 = k$, $s = 1$, $i_4 = i_3$, $t = 1$, d'. If there exist $i_2, i_5 \in \mathbf{x}$
 $\mathbf{x} \notin \text{minimize}(S^{1l})$.
 Proof. a. Choose $i_0 = \text{max}$

possible to) complete i_0, i_3 to $\mathbf{i} =$

such that \mathbf{x} is $\exists \downarrow \uparrow - \mathbf{i}$ -movable. such that **x** is $\|\cdot\|$ – **i**-movable. Indeed, we can now think of **x** as having the form *Proof.* **a. Choose** $i_0 = \max\{i \in \text{descents}(\mathbf{x}) : i \leq k - 1\}$ **. Then (check that it is possible to) complete** i_0, i_3 **to** $\mathbf{i} = (i_0, \ldots, i_5)$ **via choices** $i_1 = k, s = 1, i_4 = i_3, t = 1$ **, such that x** is $\uparrow \downarrow \uparrow - \mathbf{i}$ -movable. I **i**-motion $\mathbf{x} \mapsto \mathbf{x}' \in X$. Since y and z are to the right of the *l*^{\prime} th entry, the motion increases exactly the k'th component of **x** by the amount $x' - x$, while leaving the other components of $\mathbf{x}(k : l)$ unaltered. Hence $S^{kl}(\mathbf{x}') > S^{kl}(\mathbf{x})$, proving the claim.

In the remainder of this proof we will be more succinct and specify **i** usually only to some necessary amount.

b. Choose $i_3 \ge k$ minimal; $i_1 = i_2 = k - 1$, $i_4 = i_3$. Then we can think of **x** as having the form $\mathbf{x} = (\dots \overset{i_0}{\geq} x \mathbf{1}_r \overset{k-1}{\geq} y \mathbf{1}_s \overset{i_3}{\geq} z \mathbf{1}_t \overset{i_5}{\geq} \dots).$ $k^d(\mathbf{x}') > S^{kl}(\mathbf{x})$, proving the claim.

e succinct and specify **i** usually only
 $i_4 = i_3$. Then we can think of **x** as
 $\geq \ldots$). Consider an $||\cdot|| = \mathbf{i}$ -motion $\mathbf{x} \mapsto \mathbf{x}'$. If $i_3 \ge l$, then all entries at positions in $\{k, ..., l\}$ decrease; if $i_3 < l$, then all entries changed at positions outside $\{k, ..., l\}$ increase. These positions define a nonempty set. Either way, the con all entries changed at positions outside $\{k, \ldots, l\}$ increase. These positions define a nonempty set. Either way, the conservation of the sum of the entries of **x** by the motion $(i.e. S(\mathbf{x}) = S(\mathbf{x}'))$ guarantees $S^{kl}(\mathbf{x}') < S^{kl}(\mathbf{x})$, proving the claim.

b'. Choose $i_3 \le l$ maximal, put $i_4 = i_3$ and t maximal. An $\mathbb{1}$ – **i**-motion $\mathbf{x} \mapsto \mathbf{x}'$ replaces some entries of **x** outside positions $\{1, \ldots, l\}$ by larger values, while no such entry is replaced by a smaller value. So from sum conservation the claim follows.

c. Choose $i_4 = i_3 = l$, $i_1 = i_2 \le l - 1$ maximal (i.e. *s* maximal), $r = 1$, $i_5 > l$ *l* minimal (i.e. *t* maximal). Then we can think of **x** as $\mathbf{x} = (\dots \geqslant x \geqslant y \mathbf{1}_s \geqslant z \mathbf{1}_t \geqslant \dots).$ 754 A. Kovačec, J. K. N

c. Choose $i_4 = i_3 = l$, $i_1 = i_2$

minimal (i.e. *t* maximal). Then we

Subjecting **x** to a $\|\cdot\|$ – **i**-motion to **i**-motion to obtain $\mathbf{x}' \in X$ will in case $i_1 \leq k - 1$: increase all entries in $\mathbf{x}(k: l) = y\mathbf{1}_{s'}$ (for some $s' \leqslant s$), and in case $k - 1 < i_1$: decrease all entries moved outside of positions $\{k, \ldots, l\}$. Thus sum conservation implies $S^{kl}(\mathbf{x}') > S^{kl}(\mathbf{x})$.

d. Choose $i_5 \geq l$ minimal, $i_1 = i_2 \leq k - 2$ maximal, $s = 1$, $i_3 = i_4$, t_1 maximal. We can think of **x** as $\mathbf{x} = (\dots \overset{i_0}{\geq} x \mathbf{1}_r \overset{i_1}{\geq} y \overset{i_3}{\geq} z \mathbf{1}_t \overset{i_5}{\geq} \dots)$. An associated $\downarrow \uparrow \downarrow$ -motion $\mathbf{x} \mapsto \mathbf{x}'$ decreases all the entries in positions in $\{k, \dots, l\}$ that are $\downarrow \uparrow \downarrow$ -motion $\mathbf{x} \mapsto \mathbf{x}'$ decreases all the entries in positions in $\{k, \ldots, l\}$ that are moved; hence $S^{kl}(\mathbf{x}) > S^{kl}(\mathbf{x}')$. \mathbf{r} and \mathbf{r} and \mathbf{r} and \mathbf{r}

d'. Choose $i_2 \ge l$ minimal. Put $i_1 = i_2$ and *r* maximal. We can now apply an **i**-motion $\mathbf{x} \mapsto \mathbf{x}'$. All entries changed in positions $\{1, \ldots, l\}$ are diminished, so the claim follows. \Box

The following theorem answers completely which spaces *X* are nonempty and, for certain types of points, whether they exist in a space *X* or not. Such criteria can be useful in non-interactive software implementations of our method. A simple idea and results from $[6]$ are used: the complicated formulae in (iv) arise from explicit enumeration of all vertices of a certain polytope given there; note that the third line of (iv) will be true only in a few exceptional cases. The paper $[6]$ works with ascending chains $m \leq x_1 \leq \ldots \leq x_n \leq M$ instead of the descending $M \geq x_1 \geq \ldots \geq m$ used here; to ease comparison, the notation follows that paper as closely as possible.

THEOREM 7. a. Let $X = X(m, M, a, b; n; g)$. If $m, M \in \mathbb{R}$, define $f(l) = (l 1/m + (n + 1 - l)M$. Then $X \neq \emptyset$ if and only if $ng(\frac{a}{n}) \leq b$, $m \leq M$, and one of the *following conditions is satisfied, where quantifications are meant over integers:*

i. $m = -\infty, M = +\infty$.

ii. $M \in \mathbb{R}$ and $(m = -\infty \vee a \in [f(2), f(1)])$ and $\exists \bar{n}, 1 \leq \bar{n} \leq n$, $(n - \bar{n})g(M) +$ $\bar{n}g(\frac{a-(n-\bar{n})M}{\bar{n}})$ $\frac{(n-n)M}{\bar{n}}$) $\geqslant b$. wing conditions is satisfied, where quantifications are meant over integers:

i. $m = -\infty$, $M = +\infty$.

ii. $M \in \mathbb{R}$ and $(m = -\infty \vee a \in [f(2), f(1)])$ and $\exists \bar{n}, 1 \leq \bar{n} \leq n$, $(n - \bar{n})g(M) + \frac{-(n - \bar{n})M}{\bar{n}} \geq b$.

iii. $m \in \mathbb{$

 $n, \bar{n}g(\frac{a-(n-\bar{n})m}{\bar{n}}) + (n-\bar{n})g(m) \geqslant b.$

iv. $(m, M \in \mathbb{R} \text{ and } a \in [f(n), f(2)] \wedge ((\exists n_1, n_2, 0 \leq n_1 < \frac{a - mn}{M-m}, 0 \leq n_2 <$ $\frac{Mn-a}{M-m}$, $n_2g(M) + (n - n_1 - n_2)g(\frac{a - mn_1 - Mn_2}{n - n_1 - n_2}) + n_1g(m) \geq b) \vee (\exists n_1, n_2 \in \mathbb{Z}_{\geqslant 1}, n_1m +$ $n_2M = a$, $n_1 + n_2 = n$, and $n_2g(M) + n_1g(m) \ge b$).

b. Given integers $n_1, n_2 \in \mathbb{Z}_{\geqslant 0}$, let $\tilde{a} = a - Mn_1 - mn_2$, $\tilde{b} = b - n_1 g(M) - n_2 g(m)$. *Then* $(M1_{n_1}, x, m1_{n_2}) \leftrightarrow x$ *furnishes a 1-1 correspondence between the points of the form of the* $\text{lhs}(\leftrightarrow)$ *in* $X(m, M, a, b; n; g)$ *and the points in* $\mathbf{x} \in X(m, M, \tilde{a}, \tilde{b}; n - n_1$ n_2 ; *g*). In particular this allows via (a) to decide whether $X(m, M, a, b; n; g)$ has points *of the form* $(M1_{n_1}, \mathbf{x}, m1_{n_2}).$

Proof. a. We apply proposition 1 with $k = n$, $\alpha_1 = \ldots = \alpha_n = 1$. Define $G(\mathbf{x}) = \mathbf{a}_n$ $\sum_{i} g(x_i)$, $H = \{ \mathbf{x} \in \mathbb{R}^n : \sum x_i = a \}, P = P(m, M, a; n) = \{ \mathbf{x} \in H : M \ge x_1 \dots \ge x_m \}$ $x_n \ge m$, $\mathscr{S} = \mathscr{S}(\mathbf{1}_n, a, b) = {\mathbf{x} \in H : G(x) = b}$, and $X = X(m, M, a, b; n; g)$. Clearly $X = \mathcal{S} \cap P$ and *P* is convex. Let ext *P* be the set of extreme points of *P*. *Claim* 1. $X \neq \emptyset \Leftrightarrow (G(m) \leq b) \wedge (P \text{ is unbounded } \vee \exists p \in \text{ext } P, b \leq G(p)).$

 $\kappa \Leftarrow$: The hypotheses imply of course that $P \neq \emptyset$; hence, as is easily seen, $\mathbf{m} = \frac{a}{n} \mathbf{1}_n \in P$. Now, there always exists $\mathbf{p} \in P$ such that $b \leq G(\mathbf{p})$: if P is bounded, this is part of the hypothesis; if P is unbounded, this follows by proposition 1 (or most directly by claim 4 in its proof. By convexity of P and continuity of G , there exists $\mathbf{p}' \in P$, $\mathbf{p}' \in [\mathbf{m}, \mathbf{p}] \subseteq H$ with $G(\mathbf{p}') = b$. So $\mathbf{p}' \in \mathcal{S} \cap P = X$.

 \Rightarrow : The hypothesis implies of course $\mathscr{S} \neq \emptyset \neq P$, and so **m** \in *P* and, by proposition 1, $G(\mathbf{m}) \leq b$. If *P* is unbounded, then rhs(\Leftrightarrow) is trivially true. If *P* is bounded, then *P* is a convex polytope and hence the convex hull of its extreme points, called vertices. If we had $G(\mathbf{p}) < b$ for all $\mathbf{p} \in \text{ext } P$, then Jensen's inequality ([15], p212c7) implies $G(\mathbf{x}) < b$ for all $\mathbf{x} \in P$, hence $\mathcal{S} \cap P = \emptyset$, a contradiction. ded, then *P* is a convex polytope and hence the convex hun of its extreme points,
d vertices. If we had $G(\mathbf{p}) < b$ for all $\mathbf{p} \in \text{ext } P$, then Jensen's inequality ([15],
c7) implies $G(\mathbf{x}) < b$ for all $\mathbf{x} \in P$, hen

Note that $G(\mathbf{m}) = ng(\frac{a}{n})$, and that *P* is unbounded iff $m = -\infty, M = +\infty$. It suffices, thus, to show the following claim.

 $\Box \Rightarrow$: First, let *P* satisfy (ii). Since $M \in \mathbb{R}$, then *P* is bounded. If $a \in$ $[f(2), f(1)]$, then $P(m, M) = P(-\infty, M)$, see [6], lemma 2f. By ([6], theorem 5), $ext(P(-\infty, M))$ consists of the points $\mathbf{p} = \mathbf{p}(n_1, n_2) = (M\mathbf{1}_{n_2}, \xi\mathbf{1}_{n_1})$ where $n_1 \in \mathbb{Z}_{\geqslant 1}$, $n_1 \le n$, $n_1 + n_2 = n$, $\xi = \xi(n_1, n_2) = \frac{a - n_2 M}{n_1}$. Hence $G(\mathbf{p}(n_1, n_2)) \ge b$ is equivalent to $n_2g(M) + n_1g(\frac{a-n_2M}{n_1}) \geq b$ for some n_1 and n_2 . Thus rhs(\Leftrightarrow) follows after a simple renotation. If *P* satisfies (iii), then one can infer rhs (\Leftrightarrow) by using reasoning similar to that just given for (ii), using [6], lemma 2f, theorem 6; if *P* satisfies (iv), one uses $[6]$, theorem 4d.

 \Leftarrow : The hypothesis implies that *P* is a nonempty polytope of one of the types $P(-\infty, M)$, $P(m, M)$ or $P(m, +\infty)$ with $m, M \in \mathbb{R}$. The polytopes $P(-\infty, M)$ and $P(m, +\infty)$ have known extreme points. Using them and the hypothesis yields (ii) and (iii). Finally suppose that $P(m, M)$ satisfies the hypothesis. Since $P(m, M) \neq \emptyset$, it follows that $a \in [f(n+1), f(1)] = [f(n+1), f(n)] \cup [f(n), f(2)] \cup [f(2), f(1)]$ and these cases are precisely covered by (ii,iii,iv), and we are done. \triangleleft

b. The easy considerations are left to the reader. \Box

In theorems 8 an 9 below the structures of the sets maximizer $(S^{kl}|X)$ and minimizer $(S^{kl}|X)$, $1 \le k \le l \le n$, are determined. Note that theorem 8e relegates the calculation of the sets maximizer(S^{kn}), $2 \le k \le n$, to the calculation of the sets minimizer($S^{1,k-1}$). These latter are determined in theorem 9f. Theorem 9e relegates the calculation of the sets minimizer(S^{kn}) for $2 \le k \le n$ to the calculation of the sets maximizer($S^{1,k-1}$). Note that the determination of the latter fall under the domain of theorem 8bcd and the answers are more explicit than those for the minimizers of the $S^{1,k-1}$. Interestingly the optimizers of S^{kl} depend usually only on one of the parameters k, l ; hence by determining one of these optimizers we frequently have many others.

THEOREM 8. (maximizers) Given integers k, l with $1 \leq k \leq l \leq n-1$ and a *space* $X = X(m, M, a, b; n; g) \neq \emptyset$, define the sets $U = \{ \mathbf{x} \in X : \mathbf{x}(1 : l) = M\mathbf{1}_l \}$, $V = \{ (\xi \mathbf{1}_l, \eta \mathbf{1}_{n-l}) \in X : \xi, \eta \in \mathbb{R} \}, \quad W = \{ \mathbf{x} \in X : \mathbf{x}(l+1:n) = m \mathbf{1}_{n-l} \}.$ Then:

o. maximizer $(S^{1n}) = X$.

a. $\emptyset \neq$ maximizer $(S^{kl}) \subseteq U \cup V \cup W$; *if two of the sets U, V, W are nonempty*, *they are equal singletons.*

b. *If* $U \neq \emptyset$, *then* maximizer $(S^{kl}) = U$.

c. If $V \neq \emptyset$, then there exist ξ, η , with $M \geq \xi \geq \eta \geq m$, such that $maximize(S^{kl}) = V = \{(\xi \mathbf{1}_l, \eta \mathbf{1}_{n-l})\}.$

d. If $W \neq \emptyset$, then, letting $\tilde{a} = a - (n - l)m$, $\tilde{b} = b - (n - l)g(m)$, and $\tilde{X} = X(m, M, \tilde{a}, \tilde{b}; l; g)$, there holds maximizer $(S^{kl}) = \{ \mathbf{x} \in X : \mathbf{x}(1 : k - 1) \in I \}$ $\min\{S^{1,k-1}|\tilde{X}\}, \mathbf{x}(l+1:n) = m\mathbf{1}_{n-l}\}.$

e. *For* $2 \le k \le n$, maximizer(S^{kn}) = minimizer($S^{1,k-1}$).

Proof. o. Obvious.

a. Consider, to the extent existing, $\mathbf{u} \in U$, $\mathbf{v} \in V$, $\mathbf{w} \in W$. These *n*-tuples have the following forms: $\mathbf{u} = (M_1, \ldots, M, u_{l+1}, \ldots, u_n), \mathbf{v} = (\xi, \ldots, \xi, \eta, \ldots, \eta), \mathbf{w} =$ $(w_1, \ldots, w_l, m, \ldots, m)$. From the facts that $S^{1n}(\mathbf{u}) = S^{1n}(\mathbf{v}) = S^{1n}(\mathbf{w}) = a$ and that **u**, **v**, **w** are decreasing with entries in $[m, M]$, it is easy to see that $\mathbf{u} \succeq \mathbf{v}$, $\mathbf{w} \succeq \mathbf{v}$, $\mathbf{u} \succeq \mathbf{v}$ **w**. Hence, if $\mathbf{u} \neq \mathbf{v}$, we have by [8], p64c-4, p54c8 the contradiction $b = \sum g(u_i) > \sum g(v_i) = b$. Hypothesizing $\mathbf{u} \neq \mathbf{w}$ and $\mathbf{v} \neq \mathbf{w}$ leads to similar contradictions. Since X is compact and the func $\sum g(v_i) = b$. Hypothesizing $\mathbf{u} \neq \mathbf{w}$ and $\mathbf{v} \neq \mathbf{w}$ leads to similar contradictions. Since
X is compact and the functions S^{kl} continuous, we know that maximizer(S^{kl}) $\neq \emptyset$.
Now lemma 6ac tells us \overline{X} is compact and the functions S^{kl} continuous, we know that maximizer $(S^{kl}) \neq \emptyset$. Now lemma 6ac tells us that if $\mathbf{x} \in \text{maximize}(S^{kl})$, then the following logical formula holds true: $((\text{descents}(x) \cap \{0, \ldots, k-1\} = \emptyset) \vee (\text{descents}(x) \cap \{l+1, \ldots, n-1\})$ Rewriting this formula as a disjunction of four conjunctions using the distributive law, the set of all **x** satisfying this formula is seen to be precisely $U \cup V \cup W$, yielding (a).

b. The claim is obvious, since for any $\mathbf{x} \in X \setminus U$ and $\mathbf{u} \in U$ we have of course $S^{kl}(\mathbf{x}) < (l - k + 1)M = S^{kl}(\mathbf{u}).$

c. For any real ξ , η we have obviously $(\xi \mathbf{1}_l, \eta \mathbf{1}_{n-l}) \in X$ iff $(\xi, \eta) \in \mathcal{S}((l, n-l))$ *l*), *a*, *b*) and $M \ge \xi \ge \eta \ge m$. Now $\mathcal{S}((l, n-l), a, b)$ is by proposition 1 a point or a 'sphere' in the one-dimensional space $H_1 = \{(x, y) : lx + (n - l)y = a\}$; hence it consists of at most two points, (ξ, η) and (ξ', η') say, which define a closed (possibly degenerated) interval containing $\frac{a}{n}$ **1**₂. Hence one of the points, (ξ, η) say, satisfies $\xi \geqslant n$.

d. Under the hypothesis we infer from (a) that maximizer $(S^{kl}) \subset W$, and hence $\mathbf{x}(l+1:n) = m\mathbf{1}_{n-l}$ for any $\mathbf{x} \in \text{maximize}(S^{kl})$. Fix such **x** Then $S^{kl}(\mathbf{x}) \geq S^{kl}(\mathbf{x}')$ for all $\mathbf{x}' \in X$, hence $S^{1,k-1}(\mathbf{x}) + S^{l+1,n}(\mathbf{x}) \leq S^{1,k-1}(\mathbf{x}') + S^{l+1,n}(\mathbf{x}')$ for all $\mathbf{x}' \in X$, in particular for those in *W*, for which latter we get $S^{1,k-1}(\mathbf{x}) \leq S^{1,k-1}(\mathbf{x}')$. But $\mathbf{x}' \in W$ iff it is of the form $\mathbf{x}' = (\tilde{\mathbf{x}}, m\mathbf{1}_{n-l})$ with $\tilde{\mathbf{x}} \in \tilde{\bar{X}}$, and (d) follows.

e. This is a consequence of $S^{1,k-1}(\mathbf{x}) + S^{kn}(\mathbf{x}) = a$. \Box

THEOREM 9. (minimizers) Given integers k, l with $2 \leq k \leq l \leq n-1$ and a *space* $X = X(m, M, a, b; n; g) \neq \emptyset$, define $U = \{ \mathbf{x} \in X : \mathbf{x}(1 : k - 1) = M \mathbf{1}_{k-1} \}$, $V = \{ (\xi \mathbf{1}_{k-1}, \eta \mathbf{1}_{n-k+1}) \in X : \xi, \eta \in \mathbb{R} \}, W = \{ \mathbf{x} \in X : \mathbf{x}(k : n) = m \mathbf{1}_{n-k+1} \}.$ Then: o. minimizer $(S^{1n}) = X$.

a. $\emptyset \neq$ minimizer $(S^{kl}) \subset U \cup V \cup W$; *if two of the sets U, V, W are nonempty*, *they are equal singletons.*

b. If $U \neq \emptyset$, then letting $\tilde{a} = a - (k-1)M$, $\tilde{b} = b - (k-1)g(M)$, and $\tilde{X} = X(m, M, \tilde{a}, \tilde{b}; n - k + 1; g)$, there holds minimizer $(S^{kl}) = \{ \mathbf{x} \in X : \mathbf{x}(1 : k - 1) = 0 \}$ $M, \mathbf{x}(k:n) \in \text{maximize}(S^{l-k+2,n-k+1}|\tilde{X})$

c. *If* $V \neq \emptyset$, then there exist $M \geq \xi \geq \eta \geq m$ such that minimizer $(S^{kl}) = V =$ $\{(\xi \mathbf{1}_{k-1}, \eta \mathbf{1}_{n-k+1})\}.$

d. *If* $W \neq \emptyset$, *then* minimizer $(S^{kl}) = W$

e. *For* $2 \le k \le n$, minimizer(S^{kn}) = maximizer($S^{1,k-1}$

f. *The set* minimizer (S^{1l}) *can be obtained by an algorithm given below; it has only finitely many points, each of which has the form* $(M1_r, x, y1_s, z, m1_t)$ *; <i>if* (m, M) = $(-\infty, +\infty)$, find the unique $(\xi, \eta) \in \mathscr{S}((n-1, 1), a, b)$ with $\xi \geqslant \eta$, and the unique $(\xi', \eta') \in \mathscr{S}((1, n-1), a, b)$ with $\xi' \geqslant \eta'$, and calculate $\varepsilon = \text{sgn}(1 - l + \frac{\xi' - \xi}{\xi - \eta'})$. *Then*

minimize
$$
(S^{1l})
$$
 =
$$
\begin{cases} \{(\xi \mathbf{1}_{n-1}, \eta)\} & \text{if } \varepsilon = +1, \\ \{(\xi \mathbf{1}_{n-1}, \eta), (\xi', \eta' \mathbf{1}_{n-1})\} & \text{if } \varepsilon = 0, \\ \{(\xi', \eta' \mathbf{1}_{n-1})\} & \text{if } \varepsilon = -1. \end{cases}
$$

Proof. o. Clear.

a. The proof follows by consistently substituting in the proof of theorem 7a, *l* by $k-1$, using that compactness of *X* and continuity of S^{kl} yields minimizer $(S^{kl}) \neq \emptyset$, and using lemma 6bd in a way similar to the usage of lemma 6ac in the proof of theorem 7a.

b,c,d,e. These parts are proved in complete analogy with the parts (d,c,b,e) respectively of theorem 7. The somewhat complicated indices appearing in *S*^{\dots} in (b) express in disguise the fact that one seeks to maximize the sum of the components x_{l+1}, \ldots, x_n .

f. By lemma 6b'd', if $\mathbf{x} \in \text{minimizer}(S^{1l})$, then there do not exist $i_0, i_3 \in$ descents(**x**) with $0 \le i_0 < i_0 + 2 \le i_3 \le l$ and there do not exist $i_2, i_5 \in$ descents(**x**) with $l \le i_2 < i_2 + 2 \le i_5 \le n$. In particular it follows that $|\text{descents}(\mathbf{x}(1 : l))| \le 2$ and $|\text{descents}(\mathbf{x}(l+1:n))| \leq 2$; where these descent sets have to be treated with $\mathbf{x}(1:l)$ considered as extended to $(x_0, \mathbf{x}(1 : l), x_{l+1})$ and $\mathbf{x}(l+1 : n)$ to $(x_l, \mathbf{x}(l+1 : n), x_{n+1})$. Furthermore if the number of descents in one of these extended truncations of **x** is 2 then the descents are adjacent. It follows that the extended $\mathbf{x}(1 : l)$ has one of the forms $(x_0 \ge x_1 \ge x_2 = \ldots = x_l = x_{l+1}), (x_0 = x_1 \ge x_2 \ge x_3 = \ldots = x_l = x_{l+1}),$ \ldots $(x_0 = x_1 = x_2 = x_3 = \ldots \ge x_l \ge x_{l+1}$, where always one or both of the \ge can be strict inequality signs. We can think of $(x_l, \mathbf{x}(l+1:n), x_{n+1})$ in a similar manner. It follows that every such **x** lies in one of the sets $U_{rst} = U_{r,s,t} = \{(M\mathbf{1}_r, x, y\mathbf{1}_s, z, m\mathbf{1}_t) \in$ *X*}, $r, s, t \in \mathbb{Z}_{\geqslant 0}$, $r+s+t=n-2$. This gives us for the determination of minimizer (S^{1l}) the following algorithm:

- Determine the family $R = \{(r, s, t) : U_{rst} \neq \emptyset\}$.
- For each $\mathbf{r} = (r, s, t) \in R$ determine $u_{\mathbf{r}} \in \text{minimize}(S^{1l} | U_{\mathbf{r}})$ as follows: Using in lemma 5 (L5) the definitions $\alpha = 1$, $\beta = s$, $\gamma = 1$, $c = 1$, $d = l - r - 1$ $1, e = 0, a(L5) = a - rM - tm, b(L5) = b - rg(M) - tg(m),$ and consequently $\varepsilon = \text{sgn}\left(-l + r + 1 + \frac{x_L - x_R}{y_R - y_L}\right)$, determine $\{(\bar{x}, \bar{y}, \bar{z})\} = \text{minimize}(1x + (l - r - 1))$ 1 *y* $|\mathscr{S}'|$. Put $\mathbf{u}_{rst} = (M\mathbf{1}_r, \bar{x}, \bar{y}\mathbf{1}_s, \bar{z}, m\mathbf{1}_t)$.

• Calculate $m_{\mathbf{r}} = S^{1l}(u_{\mathbf{r}})$ and $m = \min\{m_{\mathbf{r}} : \mathbf{r} \in R\}$, and determine $R' = \{\mathbf{r} \in R :$ $m_{\bf r} = m$

• Then minimizer $(S^{1l}|X) = \bigcup_{\mathbf{r} \in R'}$ minimizer $(S^{1l}|U_{\mathbf{r}})$.

Finally, it is an easy task to see that in case $(m, M) = (-\infty, +\infty)$ the algorithm reduces to the process given in (f). \Box

3. Infimum and Supremum of Spaces $X(*, a, *, n; *)$ in the Lattice $(\mathscr{D}(n, a), \prec)$

A. KOVAČEC, J. K. MERIKOSKI, O. PIKHURKO, A. VIRTANEN
 Infimum and Supremum of Spaces $X(*, a, *, n; *)$ in the Lattice $(\mathcal{D}(n, a), \preceq)$

Bapat ([1], p62, 63) and others have observed that for $r = 0$ the set $\mathcal{D}(n, *) \cap {\mathbf{x}}$ $\mathbf{x} \geq r$, endowed with weak majorization order \preceq_w , has lattice theoretic properties. From this the corresponding properties follow at once for any fixed *r* Note that a subset of $\mathcal{D}(n, a)$ is bounded w.r.t. \prec iff it is bounded above w.r.t. \prec , and also iff it is bounded in the sense of Euclidean metric. Furthermore, every subset of $\mathcal{D}(n, a)$ is bounded below w.r.t. \leq by $\frac{a}{n} \mathbf{1}_n$.

We can now use part of Bapat's discussion to obtain part of the following (using partially his notations).

COROLLARY 10. a. Let $X = X(m, M, a, b; n; g)$ be nonempty. For $l = 1, 2, ..., n$, *let* $\alpha_l = \inf_{\mathbf{x} \in X} S^{1l}(\mathbf{x})$ *and* $\alpha'_l = \sup_{\mathbf{x} \in X} S^{1l}(\mathbf{x})$ *. Define* $\delta = \delta(X) = (\alpha_1, \alpha_2 - \alpha_1, \dots, \alpha_n - \delta)$ α_{n-1}) and $\eta = \eta(X) = (\alpha'_1, \alpha'_2 - \alpha'_1, \ldots, \alpha'_n - \alpha'_{n-1})$. Then δ and η are the infimum *and supremum of X in the partial order* $(\mathscr{D}(n, a), \preceq)$; *in other words* δ *and* n *satisfy:*

o. $\delta, \eta \in \mathscr{D}(n, a)$.

i. *For all* $\mathbf{x} \in X$, *there holds* $\delta \preceq \mathbf{x} \preceq \eta$.

ii. If δ', η' are any two elements for which the (corresponding) (o, i) hold, then $\delta' \preceq \delta \preceq \eta \preceq \eta'$

b. $(\mathscr{D}(n, a), \preceq)$ *is a conditionally complete lattice.*

Proof. a. We noted that for all $\mathbf{x} \in X$ we have $\mathbf{x} \succeq \frac{a}{n} \mathbf{1}_n$. So with obvious modifications we can reason with our *X* as Bapat $(1]$, lemma 3 and its proof) does with (his) *S*, to obtain from $\sum \delta_i = \alpha_n = a$, the facts concerning δ claimed in (o), (i), *i* **y** $l \in \{2, ..., n-1\}$. Choose $\mathbf{v} \in \text{maximize}(S^{1,l-1}), \mathbf{v}' \in \mathbf{v}'$ *Proof.* a. We noted that for all $\mathbf{x} \in X$
modifications we can reason with our X as Bay
(his) S, to obtain from $\sum_i \delta_i = \alpha_n = a$, the
and (ii). As for η , fix any $l \in \{2, ..., n - 1\}$ maximizer(S^{1l}), $\mathbf{v}'' \in \text{maximir}(\mathcal{S}^{1,l+1})$. By theorem 8a, \mathbf{v}' is in one of the sets *U*, *V*, *W* constructed for the pair $(k, l) = (1, l)$ and it follows that $(\mathbf{v}', \mathbf{v}')_1 \in \tilde{X} =$ $X(m, M, 2a, 2b; 2n; g)$ is in the corresponding \tilde{U}, \tilde{V} or \tilde{W} defined in the space \tilde{X} for the pair $(1, 2l)$. From this it follows that $(\mathbf{v}', \mathbf{v}')_1 \in \text{maximize}(S^{1,2l} | \tilde{X})$ and so $2\alpha'_l = S^{1,2l}((\mathbf{v}', \mathbf{v}')_l) \geqslant S^{1,2l}((\mathbf{v}, \mathbf{v}'')_l) \geqslant \alpha'_{l-1} + \alpha'_{l+1}$. This inequality is also seen to hold for $l = 1$ if we put $\alpha'_0 := 0$. Thus η is decreasing. Since $S^{1l}(\eta) = \alpha'_l$, it follows that $\mathbf{x} \preceq \eta$ for all $\mathbf{x} \in X$; the remaining thing to show for η is also an immediate consequence of its construction.

b. We have to show that every nonempty subset *S* of $\mathscr{D}(n, a)$, bounded below/above w.r.t. \leq , has an infimum/supremum (in $\mathcal{D}(n, a)$). We have observed in (a) that for showing the existence of the infimum one can proceed precisely as in Bapat $[1]$ adding the fact that the elements of S all have sum a , so as to permit expressing his claims with \preceq instead of with \preceq_w . Bapat's corollary 4 can be used in a similar fashion to show the existence of the supremum of a set *S* bounded above. \Box

It is interesting to note that while Bapat's non-constructive supremum definition η_{Ban} is applicable to our spaces *X* and we have by his uniqueness proofs $\eta_{\text{Ban}}(X)$ = $\eta(X)$, the two η 's are not always the same. For example, if we apply our construction to the set $S = \{(12, 2, 2, 2), (6, 6, 6, 0)\}\,$, then we obtain the non-decreasing element $\eta = (12, 2, 4, 0)$; the trade-off for our constructiveness is, hence, less general validity.

4. Applications to Eigenvalue Estimation

In this section we show how our results can be applied to estimate eigenvalues, given the traces of certain powers and/or the determinant of a matrix having only real 4. **Applications to Eigenvalue Estimation**
In this section we show how our results can be applied to estimate eigenvalues,
given the traces of certain powers and/or the determinant of a matrix having only real
eigenvalues are strictly convex. The idea is to use that for an $n \times n$ -matrix *A* and any $k \in \mathbb{Z}_{\geq 0}$, we have tr $A^k = \sum_{i=1}^{n} \lambda_i^k$, and, if $i=1$ λ_i^k , and, if *A* has positive eigenvalues, $-\ln \det A = -\sum \ln \lambda_i$.

To gauge our estimates, which are throughout best possible (apart of rounding which was done at the second decimal mostly) against an explicit example the reader may wish to use the real symmetric matrix

$$
A = \begin{bmatrix} 4 & 0 & 2 & 3 \\ 0 & 5 & 0 & 1 \\ 2 & 0 & 6 & 0 \\ 3 & 1 & 0 & 7 \end{bmatrix} \text{ that has } \begin{cases} \lambda = (9.376, 6.423, 4.775, 1.426), \\ \text{tr}A = 22, \\ \text{tr}A^2 = 154, \\ \text{tr}A^3 = 1201, \\ \text{tr}A^4 = 9954, \\ \text{det}A = 410. \end{cases}
$$

This matrix appeared in $[17]$.

As follows from the remarks preceding theorem 8, once determined the optimizers of the S^{ll} , we actually know most of the optimizers of the S^{kl} , and therefore the best possible estimates for sums of eigenvalues of the type $\lambda_k + \ldots + \lambda_l$. So we limit ourselves to just indicate the bounds for the eigenvalues themselves.

PROBLEM 1. Let $\mathscr A$ be the class of all 4×4 matrices A of real eigenvalues with trace 22 (i.e. tr $A = 22$) and the trace of whose square is 154 (i.e. tr $A^2 = 154$). Give bounds for the eigenvalues.

Solution: The 4-tuple of eigenvalues $\lambda(A) = (\lambda_1 \geq \ldots \geq \lambda_4)$ of $A \in \mathcal{A}$ lies in the space $X = X(-\infty, +\infty, 22, 154; 4; x^2)$. We know beforehand that $\lambda(A) \in X$ but could use theorem 7a to infer from $4(\frac{22}{4})^2 = 121 < 154$ that $X \neq \emptyset$. For each $A \in \mathcal{A}$ we have $S^{\prime\prime}$ (minimizer $(S^{\prime\prime})$) $\leq \lambda_l(A) \leq S^{\prime\prime}$ (maximizer $(S^{\prime\prime})$). Since $m = -\infty, M =$ $+\infty$, the sets *U* and *W* of theorems 8 and 9 are empty for all $l \in \{1, 2, 3\}$ for which they are defined; the sets *V* are consequently nonempty and obtainable from the decreasing solutions (ξ, η) to the systems $\mathcal{S}((l, 4-l), 22, 154)$. For $l = 1, 2, 3$ we obtain the solutions $\mathbf{v}_1 = (10.48 \, \mathbf{1}_1, 3.84 \, \mathbf{1}_3), \mathbf{v}_2 = (8.37 \, \mathbf{1}_2, 2.63 \, \mathbf{1}_2), \mathbf{v}_3 =$ $(7.16 \, \mathbf{1}_3, 0.53 \, \mathbf{1}_1)$, respectively. Using theorems 9c and 8c for $k = l = 2$, we obtain $S^{22}(\mathbf{v}_1) = 3.84 \le \lambda_2(A) \le S^{22}(\mathbf{v}_2) = 8.37$, similarly for $k = l = 3$ we obtain $S^{33}(v_2) = 2.63 \le \lambda_3(A) \le S^{33}(v_3) = 7.16$. Using theorem 9e for $k = n = 4$, and then theorem 8c, we have S^{44} (minimizer(S^{44})) = S^{44} (maximizer(S^{13})) = S^{44} (\mathbf{v}_3) = $0.53 \le \lambda_4(A)$, while $\lambda_4(A) \le S^{44}(\text{maximizer}(S^{44})) = S^{44}(\text{minimizer}(S^{13}))$. Now minimizer (S^{13}) is determined via the algorithm given in the proof of theorem 9f with $l = 3$. In that algorithm of course $r = t = 0$, $s = n - 2 = 2$. We find minimizer(S^{13}) = **v**₃, and hence the upper bound $\lambda_4(A) \leq 3.84$.

PROBLEM 2. Let $\mathscr A$ be the class of all 4×4 matrices *A* of real positive eigenvalues with tr $A = 22$ and det $A = 410$. Give bounds for the eigenvalues.

Solution: We have $\ln \det A = 6.02$. Similarly as in Problem 1, we have to search for minimizer(S^{*ll*}) and maximizer(S^{*ll*}); this time in space $X(-\infty, +\infty, 22, -6.02; 4; g)$ defined by the proper strictly convex function $g(x) = \begin{cases} -\ln x & \text{for } x > 0, \\ \ln x & \text{for } x < 0 \end{cases}$ The $+\infty$ for $x \le 0$. The union of the spaces *V* is given this time by the 4-tuples $\mathbf{v}_1 = (12.36 \cdot \mathbf{1}_1, 3.21 \cdot \mathbf{1}_3), \mathbf{v}_2 =$ $(8.66 \text{ } 1_2, 2.34 \text{ } 1_2), v_3 = (6.92 \text{ } 1_3, 1.24 \text{ } 1_1), \text{ from where we read off the bounds } \lambda_1(A) \leq 1_2$ 12.36, $S^{22}(\mathbf{v}_1) = 3.21 \le \lambda_2(A) \le S^{22}(\mathbf{v}_2) = 8.66$, $S^{33}(\mathbf{v}_2) = 2.34 \le \lambda_3(A) \le$ $S^{33}(v_3) = 6.92$, $1.24 \le \lambda_4(A)$. Invoking theorem 9f, we also find minimizer (S^{11}) $\{v_3\}$ and maximizer $(S^{44}) = \{v_1\}$; thus we have the bounds $6.92 \le \lambda_1(A)$ and $\lambda_4(A) \leqslant 3.21$.

Note that a matrix with real eigenvalues $\lambda_1 = \lambda_2 = 12.606$, $\lambda_3 = \lambda_4 = -1.606$ has the specified trace and determinant, but does not satisfy the bounds given. Reason: the negative eigenvalues impede rewriting the information on the determinant as a sum of logarithms of eigenvalues equality.

Another reason for requiring nonnegative eigenvalues happens when admission of negativity leads us out of the convexity region of the function *g* as is the case with $g(x) = x^3$.

PROBLEM 3. Let $\mathscr A$ be the class of all 4×4 matrices *A* of real nonnegative eigenvalues with tr $A = 22$ and tr $A^3 = 1201$. Give bounds for the eigenvalues.

Solution: We apply our theorems to the space $X = X(0, +\infty, 22, 1201; 4; x^3)$. It happens that there exist 4-tuples $(\xi \mathbf{1}_l, \eta \mathbf{1}_{4-l}) \in X$ for $l = 1, 2$. Indeed minimizer (S^{22}) $= \{(10.04 \text{ } 1_1, 3.99 \text{ } 1_3)\} = \text{maximize}(S^{11}) \text{ and } \text{minimize}(S^{33}) = \{(8.35 \text{ } 1_2, 2.65 \text{ } 1_2)\}$ $=$ maximizer(S^{22}). So the estimates easiest to obtain are $\lambda_1(A) \le 10.04$, 3.99 \le $\lambda_2(A) \leq 8.35$, $2.65 \leq \lambda_3(A)$. Furthermore, minimizer (S^{44}) = maximizer (S^{13}) = ${x \in X : x_4 = 0}$ by theorems 9e and 8ad, for it happens so that the (only) point of form $(\xi \cdot 1_3, \eta \cdot 1_1)$ in $X(-\infty, +\infty)$, namely $(7.37, -11)$, does not lie in $X(0, +\infty)$. So we find $0 \le \lambda_4$. To find a lower bound for $\lambda_1(A)$ we invoke theorem 9f with $l = 1$. The algorithm yields $r = 0$, $s = 1$, $t = 1$; hence minimizer(S^{11}) = {(7.71·1₂, 6.59·1₁, 0)}, and so 7.71 $\leq \lambda_1(A)$. Finally we find an upper bound for $\lambda_4(A)$. By theorem 8e, maximizer (S^{44}) = minimizer (S^{13}) . To find this minimizer, theorem 9f tells us to put $l = 3, r = 0$. With $t = 0, s = 2$, we find minimizer(S^{13}) = {(10.04.1₁, 3.99.1₃)}, and so $\lambda_4(A) \leq 3.99$.

PROBLEM 4. Let $\mathscr A$ be the class of all 4×4 matrices *A* of real eigenvalues with tr $A = 22$ and tr $A^4 = 9954$. Give bounds for the eigenvalues.

Solution: We can use the space $X(-\infty, +\infty, 22, 9954; 4; x^4)$. We then find by the techniques of the previous problems without difficulties that minimizer (S^{22}) = maximizer (S^{11}) = maximizer (S^{44}) = minimizer (S^{13}) = { $(9.77 \text{·}1_1, 4.08 \text{·}1_3)$ }, minimizer (S^{33}) = $\{(8.38 \cdot 1_2, 2.62 \cdot 1_2)\}\$ = maximizer (S^{22}) , and minimizer (S^{11}) = minimizer (S^{44}) = $\{(7.59 \cdot 1_3, -77 \cdot 1_1)\}$ = maximizer (S^{33}) . Thus we have $7.59 \le \lambda_1(A) \le 9.77$, $4.08 \le$ $\lambda_2(A) \leqslant 8.38, \ \ 2.62 \leqslant \lambda_3(A) \leqslant 7.59, \ \ -0.77 \leqslant \lambda_4(A) \leqslant 4.08.$

5. Notes and Concluding Remarks

a. There should be no difficulties in extending our theorems to the case of continuous not necessarily differentiable convex functions by using the theory of subgradients and subdifferentials (see e.g. [16], p214c-2, 215c2). In particular a combination of [16], p223c5 and the fact [16], p264c2 should permit to establish proposition 1, claim 3. The essence of lemma 2 is that if we move via $s \mapsto (x(s), y(s), z(s))$ along the curve \mathscr{S} . then (x, y, z) change according to \inf or \inf . This can be established via the theory of majorization. Indeed on the plane $x + y + z = a$ such changes to (x', y', z') are the only possible that allow escaping majorization-comparability (and hence by $[8]$, p92c8, inequalities of the type $g(x) + g(y) + g(z) < g(x') + g(y') + g(z')$ or its reverse).

 $\sum g(x_i) = b$ with a strictly convex function *g*, be substituted by a more general one? b. To what degree could one of the defining conditions of our spaces X , namely $(*)$ *i* Few of the more conventional generalizations seem to be possible without at least some technical troubles. Do our theorems remain essentially valid if one substitutes *g* by an unbounded quasiconvex function, i.e. a function with convex level-sets? If not before, we expect major troubles in lemma 2b. While an unbounded convex function, even after 'tilting' it with respect to its defining domain yields convex levelsets, a quasiconvex function does not have this property. We also expect troubles with the validity of lemma 2b if we generalize $(*)$ to $\sum g_i(x_i) = b$ with g_i convex. The reason is that results of the type $\mathbf{x} \preceq \mathbf{y} \Rightarrow \sum_{i} \alpha_i g_i(x_i) \leq \sum_{i} \alpha_i g_i(y_i)$, do *not* hold in general (but are valid and were used if all g_i are equal to a certain convex g , see [8], p92c-3).

c. Could we generalize, on the spaces defined in this paper, and with its methods, the functions *Skl* to be optimized? Again not much seems to be possible in this direction. We run into troubles in lemma 6. Let \dot{S}^{kl} : $\mathbb{R}^n \to \mathbb{R}$ be a function dependent only of variables x_k, \ldots, x_l . At the bottom, in lemma 6 we invariably use the reasoning that a positive or negative change of the value $\sum_{i=1}^{l} x_i$, that is of $S^{kl}(\mathbf{x})$. $\sum_{i=k} x_i$, that is of $S^{kl}(\mathbf{x})$, be reflected in a positive or negative change of the value $S^{kl}(\mathbf{x})$. This means that we have $S^{kl}(\mathbf{x}) = S^{kl}(\mathbf{y})$ iff $S^{kl}(\mathbf{x}) = S^{kl}(\mathbf{y})$. Hence the function $S^{kl}(\mathbf{x}) = g(S^{kl}(\mathbf{x}))$ for some monotone realvalued function *g*. So the generalizations possible in the sense intended are trivial.

d. The observation (c) is deplorable, for it means for example that we cannot deduce from our results that, given a space $X = X(0, +\infty, a, b; n; x^2)$, one has for certain ξ, η, ξ', η' that minimizer($\prod_{i=1}^{n} x_i$) = {($\xi \mathbf{1}_1, \eta \mathbf{1}_{n-1}$)} and maximizer($\prod_{i=1}^{n} x_i$) = $\{(\xi \mathbf{1}_{n-1}, \eta \mathbf{1}_1)\}\;$ a fact equivalent to one proved in [4]. ample that we cannot deduce $n; x^2$, one has for certain
and maximizer($\prod_{i=1}^n x_i$) =
 $\infty, +\infty, a, b; n; x^2$) \cap {**x** :

e. Given a decreasing *n*-tuple **a**, put $\tilde{X} = X(-\infty, +\infty, a, b; n; x^2) \cap {\mathbf{x}}$: $\mathbf{a} \succeq \mathbf{x}$. Using Karush-Kuhn-Tucker theory ([5], p308) find that the structure of maximizer $(S^{11}|\tilde{X})$ is of the form $(\xi \mathbf{1}_1, \eta \mathbf{1}_t, \mathbf{a}(t+1:n))$. Can problems of this type solved by our method?

f. Summing up, in a sense we now understand better why we so often find a simple structure for the optimizers. And yet, as the results mentioned in (d) and (e) painfully remind us, our understanding is still incomplete.

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