

# On the generalized convexity of quadratic functions

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## Abstract

In this paper we shall present a new approach in studying the quasiconvexity and the pseudoconvexity of a quadratic function. All classical results and new ones are obtained.

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## 1 Introduction

Generalized convexity of quadratic functions has been widely studied; the main historical references are Martos [11, 12, 13], Ferland [8], Cottle and Ferland [6], Schaible [14, 19, 18, 21].

In this paper we shall put together some results related to generalized convex quadratic functions. After noting that quasiconvexity can differ from convexity only on a proper subset  $S$  of  $\mathbb{R}^n$  and that quasiconvexity reduces to pseudoconvexity on an open set, we shall characterize the maximal domains of quasiconvexity and pseudoconvexity of a non-convex quadratic function. All the results that we are going to develop are obtained by means of an approach based on the second order characterization of pseudoconvexity. The

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ghiven approach differs from the one suggested in [22]

These results will be specified in order to obtain the criteria established by Martos [11, 12, 13] related to generalized convexity over the nonnegative orthant  $\mathbb{R}_+^n$ .

The suggested approach allows also to characterize the pseudoconvexity of a function which is the sum between a linear function and the product of affine functions.

## 2 Preliminary results

In this section we shall establish some properties of an  $n \times n$  symmetric matrix  $Q$ .

With this aim we introduce the following notations:

- $\lambda_1, \lambda_2, \dots, \lambda_n$  are the eigenvalues of the  $n \times n$  symmetric matrix  $Q$ ;
- $\{v^1, v^2, \dots, v^n\}$  is an orthonormal basis of eigenvectors associated with  $\lambda_1, \lambda_2, \dots, \lambda_n$ .  
In order to define each of the eigenvectors uniquely, we shall assume that the first component of any eigenvector is positive (this can be obtained by multiplying it by  $(-1)$  if necessary).
- $\ker Q$  is the kernel of  $Q$ , i.e.,  $\ker Q = \{x \in \mathbb{R}^n : Qx = 0\}$ ;
- $\text{rank} Q$  is the rank of  $Q$ , i.e., the maximum number of linearly independent columns (or rows) of  $Q$ ;
- $\nu_-(Q)$  is the number of the negative eigenvalues of  $Q$  (according to their multiplicity).

Regarding the number of the negative eigenvalues of  $Q$  we have the following useful lemma.

**Lemma 2.1** *Let  $Q$  be an  $n \times n$  symmetric matrix and assume the existence of two vectors  $u, w$  such that*

$$u^T Q u < 0, \quad w^T Q w < 0, \quad u^T Q w = 0.$$

*Then,  $Q$  has at least two negative eigenvalues.*

*Proof* Let  $u = \sum_{i=1}^n \alpha_i v^i$ ,  $w = \sum_{i=1}^n \beta_i v^i$ ,  $\alpha_i, \beta_i \in \mathbb{R}$ ,  $i = 1, \dots, n$ . We have

$$u^T Q u = \sum_{i=1}^n \alpha_i^2 \lambda_i, \quad w^T Q w = \sum_{i=1}^n \beta_i^2 \lambda_i, \quad u^T Q w = \sum_{i=1}^n \alpha_i \beta_i \lambda_i.$$

The assumptions imply that  $\sum_{i=1}^n \alpha_i^2 \lambda_i < 0$  and  $\sum_{i=1}^n \beta_i^2 \lambda_i < 0$ , so that at least one eigenvalue is negative. Without loss of generality assume  $\lambda_1 < 0$ . If  $\alpha_1 = 0$  or  $\beta_1 = 0$ , then obviously we have a second negative eigenvalue. If  $\alpha_1 \beta_1 \neq 0$ , we have

$$\begin{aligned} \sum_{i=1}^n (\beta_1 \alpha_i - \alpha_1 \beta_i)^2 \lambda_i &= \beta_1^2 \sum_{i=1}^n \alpha_i^2 \lambda_i + \alpha_1^2 \sum_{i=1}^n \beta_i^2 \lambda_i - 2\alpha_1 \beta_1 \sum_{i=1}^n \alpha_i \beta_i \lambda_i = \\ &= \beta_1^2 \sum_{i=1}^n \alpha_i^2 \lambda_i + \alpha_1^2 \sum_{i=1}^n \beta_i^2 \lambda_i. \end{aligned}$$

Consequently,  $\sum_{i=1}^n (\beta_1 \alpha_i - \alpha_1 \beta_i)^2 \lambda_i < 0$  so that a second negative eigenvalue exists and the thesis is achieved.  $\square$

Now we shall consider a symmetric matrix having one simple negative eigenvalue and we shall establish, for such a matrix, some fundamental properties which will be used in the next sections in characterizing the quasiconvexity and pseudoconvexity of quadratic functions.

From now on we shall assume  $\lambda_1 < 0$ ,  $\lambda_i > 0$ ,  $i = 2, \dots, p$  and  $\lambda_i = 0$ ,  $i = p+1, \dots, n$ .

The following lemma holds.

**Lemma 2.2** *Let  $Q$  be an  $n \times n$  symmetric matrix and assume  $\nu_-(Q) = 1$ . Then:*

- i) if  $u \in \mathbb{R}^n$  is such that  $u^T v^1 = 0$ , then either  $u \in \ker Q$  or  $u^T Q u > 0$ ;*
- ii)  $u \in \ker Q$  if and only if  $u^T Q u = 0$  and  $u^T v^1 = 0$ .*

*Proof* i) Let  $u = \sum_{i=1}^n \alpha_i v^i$ . We have  $0 = u^T v^1 = \alpha_1$ , so that  $u = \sum_{i=2}^n \alpha_i v^i$ . If  $u \notin \ker Q$ , there exists  $i \in \{2, \dots, p\}$  such that  $\alpha_i \neq 0$ . It follows that  $u^T Q u = \sum_{i=2}^p (\alpha_i)^2 \lambda_i > 0$ .

ii) If  $u \in \ker Q$ , obviously we have  $u^T Q u = 0$  and  $u^T v^1 = 0$ . The converse statement follows directly from i).  $\square$

Consider now the following opposite cones associated with the matrix  $Q$ :

$$T = \{x : x^T Q x \leq 0, x^T v^1 \geq 0\}, \quad -T = \{x : x^T Q x \leq 0, x^T v^1 \leq 0\}$$

We shall see in the next section that cones  $T$  and  $-T$  will play a fundamental role in characterizing the maximal domains of the quasiconvexity and pseudoconvexity of a quadratic function.

The following theorems hold, where  $\partial T$  denotes the boundary of  $T$ . Note that since  $T$  and  $-T$  are opposite cones, the properties of  $-T$  can be easily derived from the ones which will be established for  $T$ .

**Theorem 2.1** *Let  $Q$  be an  $n \times n$  symmetric matrix and assume  $\nu_-(Q) = 1$ . Then:*

- i)  $\ker Q = T \cap (-T)$ ;*
- ii)  $T$  is a pointed cone if and only if  $\ker Q = \{0\}$ .*

*Proof i)* From ii) of Lemma 2.2 we have  $\ker Q \subseteq T \cap (-T)$ . If  $x \in T \cap (-T)$  we necessarily have  $x^T Q x \leq 0$ ,  $x^T v^1 = 0$ ; consequently, i) of Lemma 2.2 implies that  $x \in \ker Q$ .

*ii)* Since  $T$  is pointed if and only if  $T \cap (-T) = \{0\}$ , the thesis follows from i). □

**Theorem 2.2** *Let  $Q$  be an  $n \times n$  symmetric matrix and assume  $\nu_-(Q) = 1$ . Then:*

- i)  $x_0 \in \text{int} T$  if and only if  $x_0^T Q x_0 < 0$  and  $x_0^T v^1 > 0$ ;*
- ii)  $x_0 \in \partial T \setminus \ker Q$  if and only if  $x_0^T Q x_0 = 0$  and  $x_0^T v^1 > 0$ ;*
- iii)  $\text{int} T \cap \text{int}(-T) = \emptyset$ ;*
- iv)  $T \cup (-T) = \{x \in \mathbb{R}^n : x^T Q x \leq 0\}$ ;*
- v)  $\text{int}(T \cup (-T)) = \text{int} T \cup \text{int}(-T)$ .*

*Proof i)* This is obvious.

*ii)* This follows by noting that  $x_0^T Q x_0 = 0$  if and only if  $x_0 \in \partial T \cup \partial(-T)$  and that  $x_0 \notin \ker Q$  if and only if  $x_0^T v^1 \neq 0$ .

*iii)* It follows from i) and from its analogous result for cone  $-T$ .

iv) This follows directly from the definitions of  $T$  and  $-T$ .

v) Since  $\text{int}(T \cup (-T)) = \{x \in \mathbb{R}^n : x^T Q x < 0\} \supseteq \text{int}T \cup \text{int}(-T)$ , we must prove that  $\text{int}T \cup \text{int}(-T) \supseteq \{x \in \mathbb{R}^n : x^T Q x < 0\}$ . Let  $x$  such that  $x^T Q x < 0$ . From Lemma 2.2 we necessarily have  $x^T v^1 \neq 0$  and the thesis follows.  $\square$

The following theorem points out the convexity of cones  $T$  and  $-T$ .

**Theorem 2.3** *Let  $Q$  be an  $n \times n$  symmetric matrix. If  $\nu_-(Q) = 1$ , then  $T$  is a closed convex cone.*

*Proof* Let  $P$  be the orthonormal matrix which has the eigenvectors  $v^1, \dots, v^n$  as columns, and let  $H$  be the diagonal matrix with the first  $p$  diagonal entries given by  $(-\lambda_1)^{-\frac{1}{2}}, (\lambda_2)^{-\frac{1}{2}}, \dots, (\lambda_p)^{-\frac{1}{2}}$ , and all the others equal to 1. It is well known that the linear transformation  $x = PHy$  reduces the quadratic form  $x^T Q x$  to the canonical form  $\sum_{i=2}^p y_i^2 - y_1^2 = \|\bar{y}\|^2 - y_1^2$ , where  $\bar{y} = (y_2, \dots, y_p)^T$ .

Consider set  $C = \{(y_1, \bar{y}) : \|\bar{y}\|^2 - y_1^2 \leq 0, y_1 \geq 0\} = \{(y_1, \bar{y}) : \|\bar{y}\| \leq y_1, y_1 \geq 0\}$ . It is easy to verify that  $C$  is a closed cone; we shall prove that  $C$  is convex. Let  $z = (z_1, \bar{z}) \in C, w = (w_1, \bar{w}) \in C$ . Since  $\|\bar{z}\| \leq z_1, \|\bar{w}\| \leq w_1$ , we have  $\|t\bar{z} + (1-t)\bar{w}\| \leq t\|\bar{z}\| + (1-t)\|\bar{w}\| \leq tz_1 + (1-t)w_1$  for all  $t \in [0, 1]$ . Consequently,  $tz + (1-t)w \in C$  for all  $t \in [0, 1]$  so that  $C$  is convex.

Taking into account that  $x^T v^1 = y^T H^T P^T v^1$  and that  $v^1 = P e^1$ , where  $e^1$  is the unit vector  $e^1 = (1, 0, \dots, 0)^T$ , we have  $x^T v^1 = y^T H e^1 = (-\lambda_1)^{-\frac{1}{2}} y^T e^1 = (-\lambda_1)^{-\frac{1}{2}} y_1$ . Consequently,  $y_1 \geq 0$  if and only if  $x^T v^1 \geq 0$  and this implies  $PH(C) = T$ . The thesis follows from the linearity of the transformation  $PH$ .  $\square$

**Remark 2.1** *Given a convex set  $C$  and a linear map  $A$ , one has  $A(\text{ri}C) = \text{ri}(AC)$ <sup>1</sup>, but, in general, the image of a closed convex set is not closed. When  $C$  is a closed convex cone*

<sup>1</sup>Let  $S$  be a convex set and let  $W$  be the smallest linear manifold containing  $S$ . Then, the relative interior of  $S$ , denoted by  $\text{ri}S$ , is the set of all interior points of  $S$  with respect to the topology induced by  $\mathbb{R}^n$  on  $W$

such that  $C \cap (-C) = \ker A$ , then  $A(\text{cl}C) = \text{cl}(AC)$ . Consequently, from i) of Theorem 2.1 and from Theorem 2.3, we have the following corollary.

**Corollary 2.1** *Let  $Q$  be an  $n \times n$  symmetric matrix and assume  $\nu_-(Q) = 1$ . Then:*

- i)  $Q(\text{int}T) = \text{ri}(Q(T))$ ,  $Q(\text{int}(-T)) = \text{ri}(Q(-T))$ ;
- ii)  $Q(T)$  and  $Q(-T)$  are closed convex cones.

Consider now the set

$$Z = \{z \in \mathbb{R}^n \setminus \{0\} : \exists w \in \text{int}T \text{ such that } z^T w = 0\}$$

and denote with  $T^+$  and  $T^-$  the positive polar and the negative polar of  $T$ , respectively. The following theorem characterizes  $Z$  in terms of the two polar cones.

**Theorem 2.4** *Let  $Q$  be an  $n \times n$  symmetric matrix. If  $\nu_-(Q) = 1$ , then  $Z = (T^+ \cup T^-)^c$ .*

*Proof* Since  $w \in \text{int}T$  if and only if either  $z^T w > 0$  for all  $z \in T^+$  or  $z^T w < 0$  for all  $z \in T^-$ , we necessarily have  $Z \cap T^+ = \emptyset$  and  $Z \cap T^- = \emptyset$ . Consequently,  $Z \subseteq (T^+ \cup T^-)^c$ . Consider now an element  $z \in (T^+ \cup T^-)^c$  and assume by contradiction that  $z^T w \neq 0 \forall w \in \text{int}T$ . The convexity of  $\text{int}T$  implies  $z^T w > 0 \forall w \in \text{int}T$  or  $z^T w < 0 \forall w \in \text{int}T$ , i.e.,  $z \in (T^+ \cup T^-)$ , and this is a contradiction. It follows that  $(T^+ \cup T^-)^c \subseteq Z$  and the thesis is achieved.  $\square$

The following lemma characterizes the image of the cones  $T$  and  $-T$  under the linear transformation  $z = Qx$ . The obtained results will play a fundamental role in characterizing the maximal domains of quasiconvexity of a quadratic function.

**Lemma 2.3** *Let  $Q$  be an  $n \times n$  symmetric matrix and assume  $\nu_-(Q) = 1$ . Then:  $Q(\text{int}T) = \text{ri}T^-$ ,  $Q(T) = T^-$ ,  $Q(\text{int}(-T)) = \text{ri}T^+$ ,  $Q(-T) = T^+$ ,  $Q((T \cup (-T))^c) = Z \cap (\ker Q)^\perp$ .*

*Proof* First of all we shall prove that  $Q(\text{int}T) \subseteq \text{ri}T^-$ ,  $Q(\text{int}(-T)) \subseteq \text{ri}T^+$ ,  $Q((T \cup (-T))^c) \subseteq (\text{ri}T^- \cup \text{ri}T^+)^c$ .

Let  $x_0 \in \text{int}T$ . Since  $x_0^T Q x_0 < 0$ ,  $Q x_0 \notin T^+$  and, taking into account Lemma 2.1,

$Qx_0 \notin Z$ . Consequently,  $Qx_0 \in T^-$  and, from Corollary 2.1,  $Q(intT) \subseteq riT^-$ . Similarly we have  $Q(int(-T)) \subseteq riT^+$ .

Now we shall prove that  $Q((T \cup (-T))^c) \cap riT^- = \emptyset$ .

Let  $z_0 \in (T \cup (-T))^c$ , i.e.  $z_0^T Q z_0 > 0$ , and let  $x_0 \in intT$ , so that  $Qx_0 \in riT^-$ . If  $Qz_0 \in riT^-$ , then  $Q([z_0, x_0]) \subseteq riT^-$  because of the convexity of  $riT^-$ . On the other hand,  $x_0 \in intT$ ,  $z_0 \notin T$  imply the existence of  $\bar{x} \in [z_0, x_0] \cap \partial T$  for which  $Q\bar{x} \in \partial T^-$ . Since  $Qx_0 \in riT^-$ , and this is a contradiction.

In a similar way it can be proven that  $Q((T \cup (-T))^c) \cap riT^+ = \emptyset$ , so that  $Q((T \cup (-T))^c) \subseteq (riT^- \cup riT^+)^c$ .

Since  $kerQ = T \cup (-T)$  implies  $T^+ \cap T^- \subseteq (kerQ)^\perp = ImQ = \{w = Qx, x \in \mathbb{R}^n\}$ , from Corollary 2.1 the thesis is achieved.  $\square$

### 3 Quadratic functions

In this section we shall characterize quadratic functions which are generalized convex.

Consider the following quadratic function

$$Q(x) = \frac{1}{2} x^T Q x + q^T x \quad (3.1)$$

where  $Q$  is an  $n \times n$  symmetric matrix,  $q \in \mathbb{R}^n$ .

We shall refer to

$$Q_0(x) = \frac{1}{2} x^T Q x \quad (3.2)$$

as the quadratic form associated with (3.1).

In order to have a self-contained paper, we recall the definition of a quasiconvex function and of a pseudoconvex function.

**Definition 3.1** *Let  $f$  be a differentiable function defined on an open set containing the convex set  $S \subseteq \mathbb{R}^n$ .*

*i)  $f$  is quasiconvex on  $S$  if*

$$x_1, x_2 \in S, f(x_1) \geq f(x_2) \Rightarrow (x_2 - x_1)^T \nabla f(x_1) \leq 0 \quad (3.3)$$

ii)  $f$  is pseudoconvex on  $S$  if

$$x_1, x_2 \in S, f(x_1) > f(x_2) \Rightarrow (x_2 - x_1)^T \nabla f(x_1) < 0 \quad (3.4)$$

**Remark 3.1** In general, a pseudoconvex function is quasiconvex too; nevertheless, quasiconvexity is equivalent to pseudoconvexity when the function does not have critical points.

A useful second order characterization of pseudoconvexity of  $Q(x)$  is the following one:

$Q(x)$  is pseudoconvex on an open convex set  $S$  if and only if (3.5) holds

$$x \in S, w \in \mathbb{R}^n, w^T(Qx + q) = 0 \Rightarrow w^T Q w \geq 0. \quad (3.5)$$

The following theorem shows that quasiconvexity reduces to convexity if the domain is the whole space  $\mathbb{R}^n$ .

**Theorem 3.1** The quadratic function  $Q(x)$  is quasiconvex on  $\mathbb{R}^n$  if and only if  $Q(x)$  is convex on  $\mathbb{R}^n$ .

*Proof* Since convexity implies quasiconvexity, the converse statement remains to be proven. By contradiction, assume that  $Q(x)$  is not convex. Then,  $Q$  has at least one negative eigenvalue  $\lambda_1$ . Let  $w$  be a normalized eigenvector associated with  $\lambda_1$  and let  $\varphi(t) = Q(tw) = \frac{1}{2} \lambda_1 t^2 + t q^T w, t \in \mathbb{R}$ . The restriction  $\varphi(t)$  has a strict local maximum point at  $\bar{t} = -\frac{q^T w}{\lambda_1}$  and consequently,  $\varphi(t)$ , and in turns  $Q(x)$ , is not quasiconvex and this contradicts the assumption.  $\square$

Theorem 3.1 implies that quasiconvexity can differ from convexity only on a proper subset  $S$  of  $\mathbb{R}^n$ . From now on, following Martos [13], we will insert the word “merely” to distinguish quadratic quasiconvex (pseudoconvex, etc.) functions that are not convex.

**Remark 3.2** Recalling that the Hessian matrix of a twice quasiconvex function evaluated at a point of its domain has at most one negative eigenvalue, a necessary condition for a quadratic function to be merely quasiconvex is that the matrix  $Q$  has one simple negative eigenvalue, i.e.,  $\nu_-(Q) = 1$ .



The following theorem shows that quasiconvexity reduces to pseudoconvexity on every open convex set of  $\mathbb{R}^n$ .

**Theorem 3.2** *The quadratic function  $Q(x)$  is quasiconvex on an open convex set  $S \subseteq \mathbb{R}^n$  if and only if it is pseudoconvex on  $S$ .*

*Proof* Since pseudoconvexity implies quasiconvexity, the converse statement remains to be proven. The thesis follows from Theorem 3.1 if  $Q(x)$  is convex, otherwise  $Q$  has one simple negative eigenvalue  $\lambda_1$ . Let  $w$  be a normalized eigenvector associated with  $\lambda_1$ . Taking into account Remark 3.1, it is sufficient to prove that the gradient of  $Q(x)$  cannot vanish. By contradiction, assume the existence of  $x_0 \in S$  such that  $\nabla Q(x_0) = Qx_0 + q = 0$ , and consider the restriction  $\varphi(t) = Q(x_0 + tw)$ . Since  $\varphi(t) = \frac{1}{2} \lambda_1 t^2 + Q(x_0)$ , this restriction has a strict local maximum point at  $t = 0$ , so that  $\varphi(t)$  and in turns  $Q(x)$ , is not quasiconvex, which contradicts the assumption.  $\square$

Note that Theorem 3.2 implies:

- a quadratic function which is merely pseudoconvex on an open convex set  $S$  has no critical points;
- a quadratic function which is merely quasiconvex on a convex set  $S$  is merely pseudoconvex at least on  $\text{int}S$ ;
- a quadratic function which is merely pseudoconvex on an open convex set  $S$  is merely quasiconvex (not necessarily pseudoconvex) on the closure of  $S$ .

**Remark 3.3** *It is important to point out that any characterization of pseudoconvexity of a quadratic function  $Q(x)$  on an open convex set  $S$  allows us to simultaneously obtain criteria for the quasiconvexity of  $Q(x)$  on the closure of  $S$ . This fact simplifies the analysis in the sense that, in order to characterize the quasiconvexity of  $Q(x)$  on  $S$ , it is sufficient to study the pseudoconvexity on the interior of  $S$ .*

Now we are able to find the maximal domains of quasiconvexity (pseudoconvexity) of a quadratic form and of a quadratic function.

**Theorem 3.3** *Let  $Q$  be an  $n \times n$  symmetric matrix. If  $\nu_-(Q) = 1$ , then the quadratic form  $Q_0(x) = \frac{1}{2} x^T Q x$  is merely quasiconvex on the closed convex cones  $T, -T$ . Furthermore,  $T$  and  $-T$  are the maximal domains of quasiconvexity of  $Q_0(x)$ .*

*Proof* Taking into account Remark 3.3, we shall prove that  $Q_0(x)$  is pseudoconvex on  $\text{int}T$ . If not, from Remark 3.1, there exist  $x_0 \in \text{int}T$ ,  $w \in \mathbb{R}^n$  such that  $w^T Q x_0 = 0$  and  $w^T Q w < 0$ . Since  $x_0^T Q x_0 < 0$ , from Lemma 2.1  $Q$  has at least two negative eigenvalues, which contradicts the assumptions. Similarly, we obtain that  $Q_0(x)$  is pseudoconvex on  $\text{int}(-T)$ . Consequently,  $Q_0(x)$  is quasiconvex on  $T$  and on  $-T$ .

The maximality of the domains remains to be proven. To see this, assume that  $Q_0(x)$  is pseudoconvex on an open set  $S$  such that  $S \cap (T \cup (-T)) \neq \emptyset$  and let  $y \in S$ ,  $y \notin T \cup (-T)$ . Then  $y^T Q y > 0$  and from Lemma 2.3,  $Qy \in Z$ . Consequently, there exists  $x_0 \in \text{int}T$  such that  $x_0^T Q y = 0$ ,  $x_0^T Q x_0 < 0$ , which contradicts (3.5). The thesis is achieved.  $\square$

Taking into account Remark 3.2, Theorem 3.3 may be re-stated as follows.

**Theorem 3.4** *A quadratic form  $Q_0(x)$  is merely quasiconvex on a convex set  $S \subset \mathbb{R}^n$ , with  $\text{int}S \neq \emptyset$ , if and only if*

- i)  $\nu_-(Q) = 1$ ;*
- ii)  $S \subseteq T$ , or  $S \subseteq -T$ .*

We shall prove that the maximal domains of quasiconvexity of a quadratic function are obtained by the ones  $(\pm T)$  associated with the quadratic form by means of a suitable translation. To this end, firstly we shall state the following theorem which gives a necessary condition for a quadratic function to be quasiconvex and which points out that, unlike the convex case, the sum of a quasiconvex function with a linear function is not, in general, quasiconvex.

**Theorem 3.5** *Assume that the quadratic function  $Q(x) = \frac{1}{2} x^T Q x + q^T x$  is merely quasiconvex on an open convex set  $S \subset \mathbb{R}^n$ . Then,  $\text{rank}Q = \text{rank}[Q, q]$ .*

*Proof* The thesis is trivial if  $q = 0$ . Consider  $\text{Im}Q = \{Qx, x \in \mathbb{R}^n\}$  and assume by contradiction that  $q \notin \text{Im}Q$ . Then, for every fixed  $x \in \mathbb{R}^n$ ,  $Qx + q \notin \text{Im}Q$  and, in

particular,  $Qx + q \notin T^+ \cup T^- \subseteq \text{Im}Q$ . From Lemma 2.3, we have  $Qx + q \in Z$  so that, from Theorem 2.4, there exists  $w \in \text{int}T$  such that  $w^T(Qx + q) = 0$ . Let  $x_0 \in S$  and consider the restriction  $\varphi(t) = Q(x_0 + tw)$ . By means of simple calculations we have  $\varphi'(0) = w^T(Qx_0 + q) = 0$ ,  $\varphi''(0) = w^TQw < 0$  so that  $t = 0$  is a strict local maximum for  $\varphi(t)$  and this implies that  $Q(x)$  is not quasiconvex on  $S$ , which contradicts the assumption. It follows that  $q \in \text{Im}Q$  or, equivalently,  $\text{rank}Q = \text{rank}[Q, q]$ .  $\square$

**Remark 3.4** Note that  $w \in \text{Im}Q$  if and only if  $w \in (\ker Q)^{\perp 2}$ . In particular,  $\text{rank}Q = \text{rank}[Q, q]$  implies that  $q \in (\ker Q)^{\perp}$ .

**Theorem 3.6** Consider the quadratic function  $Q(x) = \frac{1}{2} x^T Qx + q^T x$ .

If there exists  $s \in \mathbb{R}^n$  such that  $Qs + q = 0$ , then:

i)  $Q(x)$  is merely quasiconvex on the closed convex cones  $s + T, s - T$  if and only if  $Q_0(x) = \frac{1}{2} x^T Qx$  is merely quasiconvex on  $T, -T$ , respectively.

ii) If  $\nu_-(Q) = 1$ , then  $s + T$  and  $s - T$  are the maximal domains of quasiconvexity of  $Q(x)$  and we have

$$s + T = \{x \in \mathbb{R}^n : (x - s)^T Q(x - s) \leq 0, (v^1)^T(x - s) \geq 0\} \quad (3.6)$$

$$s - T = \{x \in \mathbb{R}^n : (x - s)^T Q(x - s) \leq 0, (v^1)^T(x - s) \leq 0\} \quad (3.7)$$

*Proof* i)  $Q(x)$  is pseudoconvex on  $s \pm \text{int}T$  if and only if (3.5) holds with  $S = s \pm \text{int}T$ . Since  $x \in s \pm \text{int}T$  if and only if  $x - s = u \in \pm \text{int}T$ , we have  $Qx + q = Q(x - s) = Qu$  so that  $Q(x)$  is merely pseudoconvex on  $s \pm \text{int}T$  if and only if  $Q_0(x)$  is merely pseudoconvex on  $\pm \text{int}T$  and the thesis follows.

ii) Theorem 3.3 implies that  $\pm T$  are the maximal domains of quasiconvexity of  $Q_0(x)$  so that, taking into account i),  $s \pm T$  are the maximal domains of quasiconvexity of  $Q(x)$ .

Finally,  $x \in s \pm T$  if and only if  $x - s \in \pm T$ , so that (3.6), and (3.7) hold.

The proof is complete.  $\square$

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<sup>2</sup>Given a linear subspace  $W \subseteq \mathbb{R}^n$ ,  $W^{\perp} = \{z \in \mathbb{R}^n : z^T w = 0, \forall w \in W\}$

**Remark 3.5** *If  $Q$  is a singular matrix, then a stationary point of  $Q(x)$  is not unique. However, the characterization of the maximal domains of quasiconvexity is independent of the particular stationary point used. To see this, let  $s_1, s_2$  be two distinct stationary points, i.e.,  $Qs_1 + q = Qs_2 + q = 0$ . We have  $s_1 = s_2 + u$ ,  $u \in \ker Q \subset T \cup (-T)$ . It follows that  $s_1 \pm T = s_2 + u \pm T = s_2 \pm T$ .*

The previous results allow us to characterize the merely quasiconvexity of a quadratic function.

**Theorem 3.7** *The quadratic function  $Q(x)$  is merely quasiconvex on a convex set  $S$  with nonempty interior if and only if*

- i)  $\nu_-(Q) = 1$ ;*
- ii) there exists  $s \in \mathbb{R}^n$  such that  $Qs + q = 0$ ;*
- iii)  $S \subseteq s \pm T$ .*

*Proof* Assume that  $Q(x)$  is merely quasiconvex on  $S$ . We necessarily have  $\nu_-(Q) = 1$  and from Theorem 3.5, *ii)* follows; *iii)* is a direct consequence of Theorem 3.6.

Conversely, the thesis follows from Theorem 3.3 and from Theorem 3.6. □

**Corollary 3.1** *If  $Q(x)$  is merely quasiconvex on a convex set  $S$  with nonempty interior, then  $Q_0(x)$  is merely quasiconvex on  $S - s$ , where  $s$  is such that  $Qs + q = 0$ .*

*Proof* The thesis follows from Theorem 3.7, taking into account Theorem 3.3. □

The following examples clarify the results given in Theorem 3.4 and in Theorem 3.7.

**Example 3.1** *Consider the quadratic form  $Q_0(x) = 2x_1^2 - x_2^2 - x_1x_2$ .*

*We have  $Q = \begin{bmatrix} 4 & -1 \\ -1 & -2 \end{bmatrix}$ ,  $\lambda_1 = 1 - \sqrt{10} < 0$ ,  $\lambda_2 = 1 + \sqrt{10} > 0$ ,  $v^1 = \frac{v}{\|v\|}$  with  $v = (1, 3 + \sqrt{10})^T$ .*

*Theorem 3.4 implies that  $Q_0(x)$  is quasiconvex on the maximal domains  $T, -T$ . It is easy to verify that  $T = \{x = \alpha(1, 1)^T + \beta(-1, 2)^T, \alpha, \beta \geq 0\}$ , so that the positive and negative*

polars of  $T$  are respectively,

$$T^+ = \{x = \alpha_1(-1, 1)^T + \beta_1(2, 1)^T, \alpha_1, \beta_1 \geq 0\},$$

$$T^- = \{x = \alpha_1(-1, 1)^T + \beta_1(2, 1)^T, \alpha_1, \beta_1 \leq 0\}.$$

Now we shall verify that the image of  $T$  under the linear transformation  $Q$  is  $T^-$ .

$$\text{In fact, } Q(T) = \{y = \alpha Q(1, 1)^T + \beta Q(-1, 2)^T, \alpha, \beta \geq 0\} = \{y = -3\alpha(-1, 1)^T - 3\beta(2, 1)^T, \alpha, \beta \geq 0\} = \{y = \alpha_1(-1, 1)^T + \beta_1(2, 1)^T, \alpha_1, \beta_1 \leq 0\} = T^-.$$

By means of similar calculations we can verify that  $Q(-T) = T^+$ .

Let us note that the nonsingularity of  $Q$  implies that the quadratic function  $Q(x) = Q_0(x) + q^T x$  is quasiconvex for every  $q \in \mathbb{R}^n$  on the maximal domains  $s + T$  and  $s - T$ , where  $s = -Q^{-1}q$ .

**Example 3.2** Consider the quadratic function  $Q(x) = -x_1^2 - x_2^2 - 2x_1x_2 + 2x_1 + 2x_2$ . We

$$\text{have } Q = \begin{bmatrix} -2 & -2 \\ -2 & -2 \end{bmatrix}, q = (2, 2)^T, \lambda_1 = -4 < 0, \lambda_2 = 0, v^1 = \frac{v}{\|v\|} \text{ with } v = (1, 1)^T.$$

Note that  $Q(x)$  is a concave function; nevertheless, since  $\text{rank } Q = \text{rank } [Q, q]$ ,  $Q(x)$  is also merely quasiconvex on  $s + T$  and on  $s - T$ , where  $s$  is any vector of the kind  $s = (s_1, -s_1 + 1)^T$ ,  $s_1 \in \mathbb{R}$ . Taking into account that  $x^T Q x \leq 0 \forall x \in \mathbb{R}^2$ , we have  $s + T = \{x \in \mathbb{R}^2 : (v^1)^T(x - s) \geq 0\} = \{x \in \mathbb{R}^2 : x_1 + x_2 - 1 \geq 0\}$ .

At last, we shall characterize the merely quasiconvexity of a quadratic function over a half-space.

**Theorem 3.8**  $Q(x)$  is merely quasiconvex on the half-space  $H = \{x \in \mathbb{R}^n : h^T x + h_0 \geq 0\}$  if and only if (3.8) holds:

$$\nu_-(Q) = 1, \ker Q = h^\perp, \exists \beta \in \mathbb{R} : q = \beta h, h_0 \leq \beta \frac{\|h\|^4}{h^T Q h}. \quad (3.8)$$

*Proof* Assume that  $Q(x)$  is merely quasiconvex on  $H$ . From Theorem 3.7, we have  $H \subseteq s + T \subseteq \Gamma = \{x \in \mathbb{R}^n : (v^1)^T(x - s) \geq 0\}$ , or  $H \subseteq s - T \subseteq \Gamma_1 = \{x \in \mathbb{R}^n : (v^1)^T(x - s) \leq 0\}$ . Since  $H \subseteq \Gamma$  or  $H \subseteq \Gamma_1$ ,  $\partial H$  and  $\partial \Gamma$  are necessarily parallel hyperplanes so that  $h = kv^1$ ,  $k \neq 0$ , i.e.,  $h$  is an eigenvector associated with the negative eigenvalue  $\lambda_1$ . Obviously,  $k > 0$  implies  $H \subseteq \Gamma$ , while  $k < 0$  implies  $H \subseteq \Gamma_1$ . We shall limit ourselves to considering the case  $k > 0$  since the other one is perfectly analogous.

Note that  $H \subseteq \Gamma$  if and only if  $h_0 \leq -h^T s$ ; when  $h_0 = -h^T s$  we have  $H = s + T = \Gamma$  and  $T = \{x \in \mathbb{R}^n : x^T Q x \leq 0, h^T x \geq 0\}$ .

Since the quadratic form  $\frac{1}{2}x^T Q x$  is merely quasiconvex on  $T$  and on  $-T$ , by means of continuity, we have  $\frac{1}{2}x^T Q x \leq 0, \forall x \in \mathbb{R}^n$  and, because  $\nu_-(Q) = 1$ , this implies that  $\ker Q = h^\perp$  and  $\text{Im} Q = \{kh, k \in \mathbb{R}\}$ . Since  $\text{rank} Q = \text{rank}(Q, q)$ , there exists  $\beta \in \mathbb{R}$  such that  $q = \beta h$ . If  $\beta = 0$ , we have  $s \in \ker Q = h^\perp$  so that  $h_0 \leq -h^T s = 0$  and (3.8) holds. If  $\beta \neq 0$ , we can choose any  $s \in s_0 + h^\perp$ , where  $Q s_0 = -q$ . In particular,  $s = (s_0 + h^\perp) \cap \text{Im} Q$  is an eigenvector of  $Q$  and thus  $Q s = \lambda_1 s$ . It follows that  $h_0 \leq -h^T s = -\frac{h^T Q s}{\lambda_1} = \frac{h^T \beta h}{\lambda_1} = \beta \frac{\|h\|^2}{\lambda_1} = \beta \frac{\|h\|^4}{h^T Q h}$ .

Conversely, by choosing  $s = -\frac{\beta}{\lambda_1} h$ , it is easy to verify that (3.8) implies i), ii), and iii) of Theorem 3.7.  $\square$

**Corollary 3.2**  $Q(x)$  is merely quasiconvex on the half-space  $H = \{x \in \mathbb{R}^n : h^T x + h_0 \geq 0\}$  if and only if  $Q = \mu h h^T, q = \beta h$ , with  $\mu < 0$  and  $h_0 \leq \frac{\beta}{\mu}$ .

## 4 Quadratic functions of nonnegative variables

By specifying the results given in the previous section, it is possible to establish criteria for generalized convex quadratic functions on  $\mathbb{R}_+^n$ . These results were obtained for the first time by Martos in [11, 12], by introducing of the concept of positive subdefinite matrices. Now we shall characterize the quasiconvexity of a quadratic form on the nonnegative orthant. The first result points out the relationships between the nonnegative orthant and maximal cone  $T$ .

**Theorem 4.1** The quadratic form  $Q_0(x)$  is merely quasiconvex on  $\mathbb{R}_+^n$  if and only if

i)  $\nu_-(Q) = 1$ ;

ii)  $\mathbb{R}_+^n \subseteq T$ .

*Proof* From Theorem 3.4,  $Q_0(x)$  is merely quasiconvex on  $\mathbb{R}_+^n$  if and only if i) holds and either  $\mathbb{R}_+^n \subseteq T$  or  $\mathbb{R}_+^n \subseteq -T$ . This last inclusion cannot hold; in fact  $(v^1)^T x \leq 0, \forall x \in \mathbb{R}_+^n$  implies that  $v^1 \in \mathbb{R}_-^n$  and this is a contradiction since the first nonzero component of  $v^1$

is positive. The thesis follows.  $\square$

The following theorem characterizes a quadratic form on the nonnegative orthant in terms of the sign of the elements of matrix  $Q$ .

**Theorem 4.2** *The quadratic form  $Q_0(x)$  is merely quasiconvex on  $\mathbb{R}_+^n$  if and only if*

i)  $\nu_-(Q) = 1$ ;

ii)  $Q \leq 0$ <sup>3</sup>.

*Proof* Assume that i) and ii) hold and let  $\Gamma$  be the subspace spanned by the normalized eigenvectors associated with the nonnegative eigenvalues of  $Q$ ; we have  $\Gamma = \{x \in \mathbb{R}^n : (v^1)^T x = 0\}$ . Since  $x^T Q x \geq 0$  for all  $x \in \Gamma$ , and  $x^T Q x \leq 0$  for all  $x \in \mathbb{R}_+^n$ ,  $\Gamma$  is a supporting hyperplane to  $\mathbb{R}_+^n$  at the origin, so that  $v^1 \in \mathbb{R}_+^n$ . Consequently, the elements of  $\mathbb{R}_+^n$  satisfy the inequalities  $x^T Q x \leq 0$ ,  $(v^1)^T x \geq 0$  so that  $\mathbb{R}_+^n \subseteq T$  and the thesis follows from Theorem 4.1.

Assume now that  $Q_0(x)$  is merely quasiconvex on  $\mathbb{R}_+^n$ .

From Theorem 4.1, i) holds and furthermore  $\mathbb{R}_+^n \subseteq T$  so that  $x^T Q x \leq 0$  for all  $x \in \mathbb{R}_+^n$ ; in particular  $(e^i)^T Q e^i = q_{ii} \leq 0$ ,  $i = 1, \dots, n$ . Consider now the submatrix of  $Q$ ,

$$Q_{ij} = \begin{bmatrix} q_{ii} & q_{ij} \\ q_{ij} & q_{jj} \end{bmatrix} \text{ and the restriction } \varphi(x_i, x_j) = \frac{1}{2}(q_{ii}x_i^2 + 2q_{ij}x_i x_j + q_{jj}x_j^2).$$

Since  $\varphi(x_i, x_j) \leq 0$ ,  $\forall (x_i, x_j) \in \mathbb{R}_+^2$ , we have  $q_{ij} \leq 0$  when  $q_{ii}q_{jj} = 0$ . Consider the case  $q_{ii} < 0$ ,  $q_{jj} < 0$ . The quasiconvexity of  $\varphi$  implies that  $Q_{ij}$  has at most one negative eigenvalue, so that  $q_{ii}q_{jj} - q_{ij}^2 \leq 0$ . If  $q_{ii}q_{jj} - q_{ij}^2 < 0$ , the equation  $q_{ii}x_i^2 + 2q_{ij}x_i x_j + q_{jj}x_j^2 = 0$  has, for every fixed  $x_j$  (or  $x_i$ ), two roots which cannot be positive since  $\varphi(x_i, x_j) \leq 0$ ,  $\forall (x_i, x_j) \in \mathbb{R}_+^2$ ; consequently,  $q_{ij} \leq 0$ .

If  $q_{ii}q_{jj} - q_{ij}^2 = 0$ , we have  $\varphi(x_i, x_j) = \frac{1}{2q_{ii}}(q_{ii}x_i + q_{ij}x_j)^2$ . This function has a line  $r$  of critical points which are global maximum points; since the quasiconvexity of  $\varphi$  implies that  $r \cap \text{int}\mathbb{R}_+^2 = \emptyset$ , we necessarily have  $q_{ij} \leq 0$ .

It follows that  $q_{ij} \leq 0$ ,  $\forall i, j = 1, \dots, n$ , i.e.,  $Q \leq 0$  and the proof is complete.  $\square$

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<sup>3</sup> $Q \leq 0$  means  $q_{ij} \leq 0$ ,  $\forall i, j$

**Theorem 4.3** Let  $Q(x)$  be merely quasiconvex on  $\mathbb{R}_+^n$ . Then,  $Q_0(x)$  is merely quasiconvex on  $\mathbb{R}_+^n$ .

*Proof* From ii) of Theorem 3.6, either  $\mathbb{R}_+^n \subseteq s + T$  or  $\mathbb{R}_+^n \subseteq s - T$ . Let  $v_j^1 > 0$  be the first nonzero component of  $v^1$ ; since  $te^j \in \mathbb{R}_+^n$ ,  $\forall t > 0$ , we have  $(v^1)^T(te^j - s) = tv_j^1 - (v^1)^T s > 0$ , for a large enough  $t$ . It follows that  $\mathbb{R}_+^n \subseteq s + T$ . Consequently, we must prove that  $\mathbb{R}_+^n \subseteq T$ , i.e.,  $x^T Q x \leq 0$ ,  $(v^1)^T x \geq 0$ ,  $\forall x \in \mathbb{R}_+^n$ . Assume the existence of  $\bar{x}$  such that  $\bar{x}^T Q \bar{x} > 0$  ( $(v^1)^T \bar{x} < 0$ ). Since  $t\bar{x} \in \mathbb{R}_+^n$ ,  $\forall t > 0$ , for a large enough  $t$  we have  $(t\bar{x} - s)^T Q(t\bar{x} - s) > 0$  ( $(v^1)^T(t\bar{x} - s) < 0$ ) and this is a contradiction.

Consequently,  $\mathbb{R}_+^n \subseteq T$ , so that  $Q_0$  is merely quasiconvex on  $\mathbb{R}_+^n$ .  $\square$

The following example shows that the converse statement of Theorem 4.3 does not hold; we need some additional assumptions on the vector  $q$  which will be given in Theorem 4.4.

**Example 4.1** Consider the quadratic function  $Q(x_1, x_2) = -x_1 x_2 + x_1 - x_2$ .

We have  $Q = \frac{1}{2} \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix}$ ,  $q = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ .

$Q_0(x_1, x_2) = -x_1 x_2$  is merely quasiconvex on  $\mathbb{R}_+^2$  according to Theorem 4.2. On the other hand,  $Q(x_1, x_2)$  is not quasiconvex on  $\mathbb{R}_+^2$ , since its restriction on line  $x_2 = x_1 + 2$  has a strict local maximum point at  $(1, 3)$ .

**Theorem 4.4** A quadratic function  $Q(x) = \frac{1}{2} x^T Q x + q^T x$  is merely quasiconvex on  $\mathbb{R}_+^n$  if and only if

- i)  $\nu_-(Q) = 1$ ;
- ii)  $Q \leq 0$ ;
- iii) there exists  $s \in \mathbb{R}^n$  such that  $Qs + q = 0$ ,  $q^T s \geq 0$ ;
- iv)  $q \leq 0$ .

*Proof* Assume that i)-iv) hold. From Theorem 3.6 we must prove that  $\mathbb{R}_+^n \subseteq s + T = \{x \in \mathbb{R}^n : (x - s)^T Q(x - s) \leq 0, (v^1)^T(x - s) \geq 0\}$ .

We have  $(x - s)^T Q(x - s) = x^T Q x + 2q^T x - q^T s$ , so that ii), iii) and iv) imply that  $(x - s)^T Q(x - s) \leq 0$ ,  $\forall x \in \mathbb{R}_+^n$ . From Theorem 4.2 and Theorem 4.1 we have  $\mathbb{R}_+^n \subseteq T$



and this implies that  $v^1 \in \mathfrak{R}_+^n$ . On the other hand,  $s^T v^1 = \frac{1}{\lambda_1} s^T Q v^1 = -\frac{1}{\lambda_1} q^T v^1 \leq 0$ , and consequently,  $(v^1)^T(x - s) \geq 0, \forall x \in \mathfrak{R}_+^n$ , so that  $\mathfrak{R}_+^n \subseteq T$ .

Assume now that  $Q(x)$  is merely quasiconvex on  $\mathfrak{R}_+^n$ . From Theorem 3.7 we have  $\nu_-(Q) = 1, \exists s \in \mathfrak{R}^n : Qs + q = 0$ , and  $\mathfrak{R}_+^n \subseteq s + T$ , while from Theorem 4.3 and from Theorem 4.2, we have  $Q \leq 0$  and  $\mathfrak{R}_+^n \subseteq T$ . It remains to be proven that  $q \leq 0$  and  $q^T s \geq 0$ . The inclusion  $\mathfrak{R}_+^n \subseteq s + T$  implies that  $0 \in s + T$ , i.e.,  $s \in -T$ . Consequently,  $s^T Q s \leq 0$  and since  $q^T s = -s^T Q s$ , we have  $q^T s \geq 0$ . Furthermore, from Lemma 2.3,  $Qs \in T^+$ , i.e.,  $q \in T^-$ ; it follows that  $q^T x \leq 0 \forall x \in T$  and, in particular,  $q^T x \leq 0 \forall x \in \mathfrak{R}_+^n$  so that  $q \leq 0$ .

The proof is complete. □

## 5 Pseudoconvexity of a quadratic function on a closed set

In this section we shall characterize the maximal domains of pseudoconvexity of a non-convex quadratic function. In particular, we are interested in analyzing pseudoconvexity on the nonnegative orthant  $\mathfrak{R}_+^n$  since many extremum quadratic problems have a feasible region contained in  $\mathfrak{R}_+^n$  and not just in the open set  $\text{int}\mathfrak{R}_+^n$ . Since  $\mathfrak{R}_+^n$  is a closed set, we shall refer to the following definition of pseudoconvexity at a point.

**Definition 5.1** *Let  $f$  be a differentiable function defined on an open set of  $\mathfrak{R}^n$  containing the convex set  $S$ .  $f$  is pseudoconvex at  $x_0 \in S$  if*

$$x \in S, f(x) < f(x_0) \Rightarrow (x - x_0)^T \nabla f(x_0) < 0.$$

Obviously, a function is pseudoconvex on a convex set if and only if it is pseudoconvex at every point of the set.

Since a non-convex quadratic function is quasiconvex on a convex set  $S$  if and only if it is pseudoconvex on  $\text{int}S$ , we must further investigate the study of the pseudoconvexity on the boundary of  $S$ , starting from the maximal domains of quasiconvexity of a quadratic

form.

Regarding this, the following lemma holds.

**Lemma 5.1** *Consider the quadratic form  $Q_0(x) = \frac{1}{2} x^T Q x$ , and assume  $\nu_-(Q) = 1$ . Then,  $Q_0(x)$  is pseudoconvex at  $x_0 \in \pm T$  if and only if  $\nabla Q_0(x_0) = Q x_0 \neq 0$ .*

*Proof* Since  $Q_0(x)$  is merely pseudoconvex on  $\pm \text{int}T$  (see Theorem 3.3 and Theorem 3.2), we must investigate the boundary of cones  $T$  and  $-T$ . We shall consider  $\partial T$  since the other case is analogous.

$Q_0(x)$  is pseudoconvex at  $x_0 \in \partial T$  if and only if (5.9) holds:

$$x_0 \in \partial T, x \in T, Q_0(x) < Q_0(x_0) \Rightarrow (x - x_0)^T Q x_0 < 0. \quad (5.9)$$

Obviously, (5.9) implies that  $\nabla Q_0(x_0) = Q x_0 \neq 0$ .

Conversely, let  $x_0 \in \partial T$ ; we necessarily have  $Q_0(x_0) = 0$ , so that  $Q_0(x) < Q_0(x_0) = 0$  implies that  $x \in \text{int}T$ . On the other hand,  $Q x_0 \neq 0$  implies that  $Q x_0 \in T^- \setminus \{0\}$  so that  $x^T Q x_0 < 0, \forall x \in \text{int}T$ . Since  $(x - x_0)^T Q x_0 = x^T Q x_0$ , the thesis is achieved.  $\square$

The following theorems, which are a direct consequence of Lemma 5.1, characterize the maximal domains of pseudoconvexity of a nonconvex quadratic form.

**Theorem 5.1** *Consider the quadratic form  $Q_0(x)$  and assume that  $\nu_-(Q) = 1$ . Then the following properties hold:*

- i)  $Q_0(x)$  is merely pseudoconvex on the maximal domains  $T \setminus \ker Q, -T \setminus \ker Q$ ;*
- ii)  $Q_0(x)$  is merely pseudoconvex on  $T \setminus \{0\}, -T \setminus \{0\}$  if and only if  $Q$  is nonsingular.*

Theorem 5.1 may be re-stated as follows.

**Theorem 5.2** *A quadratic form  $Q_0(x)$  is merely pseudoconvex on a convex set  $S$  with nonempty interior if and only if*

- i)  $\nu_-(Q) = 1$ ;*
- ii)  $S \subseteq T \setminus \ker Q$ , or  $S \subseteq -T \setminus \ker Q$ .*

Taking into account i) of Theorem 2.1, the maximal domains  $T \setminus \ker Q, -T \setminus \ker Q$  can be characterized by means of the inequalities  $(v^1)^T x > 0, (v^1)^T x < 0$ , respectively. More exactly, we have the following theorem.

**Theorem 5.3** Consider the quadratic form  $Q_0(x) = \frac{1}{2} x^T Q x$  and assume  $\nu_-(Q) = 1$ . Then, the maximal domains of pseudoconvexity of  $Q_0(x)$  are given by

$$\begin{aligned} T \setminus \ker Q &= \{x \in \mathbb{R}^n : x^T Q x \leq 0, (v^1)^T x > 0\} \\ -T \setminus \ker Q &= \{x \in \mathbb{R}^n : x^T Q x \leq 0, (v^1)^T x < 0\}. \end{aligned}$$

The relation between the pseudoconvexity of a quadratic function and the pseudoconvexity of the corresponding quadratic form is specified in the following theorem.

**Theorem 5.4** Consider the quadratic function  $Q(x) = \frac{1}{2} x^T Q x + q^T x$ , and assume the existence of  $s \in \mathbb{R}^n$  such that  $Qs + q = 0$ . Then,  $Q(x)$  is pseudoconvex on  $s \pm T$  if and only if  $Q_0(x) = \frac{1}{2} x^T Q x$  is pseudoconvex on  $\pm T$ .

*Proof*  $Q(x)$  is pseudoconvex at  $x_0 \in s + T$  if and only if

$$x \in s + T, Q(x) < Q(x_0) \Rightarrow \nabla Q(x_0)^T (x - x_0) < 0. \quad (5.10)$$

Set  $y_0 = x_0 - s \in T$ ,  $y = x - s \in T$ . We have  $Q(x) = \frac{1}{2}(x - s)^T Q(x - s) - \frac{1}{2}s^T Q s = Q_0(y) - \frac{1}{2}s^T Q s$ ,  $Q(x_0) = Q_0(y_0) - \frac{1}{2}s^T Q s$ . It follows that  $Q(x) < Q(x_0)$  if and only if  $Q_0(y) < Q_0(y_0)$ . Furthermore,  $\nabla Q(x_0) = Qx_0 + q = Q(x_0 - s) = Qy_0 = \nabla Q_0(y_0)$ , so that  $\nabla Q(x_0)^T (x - x_0) < 0$  if and only if  $\nabla Q_0(y_0)^T (y - y_0) < 0$ . Consequently, (5.10) is equivalent to

$$y_0 \in T, y \in T, Q_0(y) < Q_0(y_0) \Rightarrow \nabla Q_0(y_0)^T (y - y_0) < 0$$

i.e., the pseudoconvexity of  $Q(x)$  on  $s + T$  is equivalent to the pseudoconvexity of  $Q_0(y)$  on  $T$ .

Analogously, the pseudoconvexity of  $Q(x)$  on  $s - T$  is equivalent to the pseudoconvexity of  $Q_0(y)$  on  $-T$ .  $\square$

The following lemma extends Lemma 5.1 to a quadratic function.

**Lemma 5.2** Consider the quadratic function  $Q(x) = \frac{1}{2} x^T Q x + q^T x$  with  $\nu_-(Q) = 1$ , and assume the existence of  $s \in \mathbb{R}^n$  such that  $Qs + q = 0$ . Then,  $Q(x)$  is merely pseudoconvex at  $x_0 \in s \pm T$  if and only if  $\nabla Q(x_0) = Qx_0 + q \neq 0$ .

*Proof* From Theorem 5.4,  $Q(x)$  is merely pseudoconvex at  $x_0 \in s \pm T$  if and only if  $Q_0$  is merely pseudoconvex at  $y_0 = x_0 - s \in \pm T$ .

Since  $Qy_0 = Q(x_0 - s) = Qx_0 + q$ , the thesis follows from Lemma 5.1.  $\square$

As a direct consequence of the previous results, we have the following characterization of the maximal domains of pseudoconvexity of a quadratic function.

**Theorem 5.5** *The quadratic function  $Q(x) = \frac{1}{2} x^T Qx + q^T x$  is merely pseudoconvex on a convex set  $S$  with nonempty interior if and only if*

i)  $\nu_-(Q) = 1$ ;

ii) there exists  $s \in \mathbb{R}^n$  such that  $Qs + q = 0$ ;

iii)  $S \subseteq s + (T \setminus \ker Q) = \{x \in \mathbb{R}^n : (x - s)^T Q(x - s) \leq 0, (v^1)^T(x - s) > 0\}$ , or  $S \subseteq s - (T \setminus \ker Q) = \{x \in \mathbb{R}^n : (x - s)^T Q(x - s) \leq 0, (v^1)^T(x - s) < 0\}$ .

## 5.1 Pseudoconvexity on the nonnegative orthant

The above criteria can be specified to the case where  $S$  is the nonnegative orthant.

**Theorem 5.6** *Let  $Q_0(x) = \frac{1}{2} x^T Qx$  be merely quasiconvex on  $\mathbb{R}_+^n$ . Then,  $Q_0(x)$  is merely pseudoconvex on  $\mathbb{R}_+^n \setminus \{0\}$  if and only if  $Q$  does not contain a column (or a row) of zeros.*

*Proof* From Lemma 5.1,  $Q_0(x)$  is pseudoconvex on  $\mathbb{R}_+^n \setminus \{0\}$  if and only if  $Qx \neq 0 \forall x \in \mathbb{R}_+^n \setminus \{0\}$ . By denoting with  $q^j$  the  $j$ -th column of  $Q$ ,  $j = 1, \dots, n$ , we have  $Qx = \sum_{j=1}^n x_j q^j$ ,  $x_j \geq 0$ . Since  $Q \leq 0$  (see Theorem 4.2),  $Qx = 0$  if and only if for some  $j$  we have  $q^j = 0$ .  $\square$

**Theorem 5.7** *Let  $Q(x) = \frac{1}{2} x^T Qx + q^T x$  be merely quasiconvex on  $\mathbb{R}_+^n$ . Then,  $Q(x)$  is merely pseudoconvex on  $\mathbb{R}_+^n$  if and only if  $q \neq 0$ .*

*Proof* From Lemma 5.2,  $Q(x)$  is pseudoconvex on  $\mathbb{R}_+^n$  if and only if  $Qx + q \neq 0 \forall x \in \mathbb{R}_+^n$ . Since  $Q \leq 0$ ,  $q \leq 0$  (see Theorem 4.4), the thesis follows.  $\square$

By applying Theorem 4.2 and Theorem 5.6 to a  $2 \times 2$  matrix we obtain the following criteria of quasiconvexity and of pseudoconvexity of a quadratic form.

**Theorem 5.8** Consider the matrix  $Q = \begin{bmatrix} \alpha & \beta \\ \beta & \gamma \end{bmatrix}$ . Then, the quadratic form  $Q_0(x) = \frac{1}{2}x^T Q x$  is merely quasiconvex on  $\mathbb{R}_+^2$  if and only if

i)  $\alpha \leq 0, \beta \leq 0, \gamma \leq 0, (\alpha, \beta, \gamma) \neq (0, 0, 0)$ ;

ii)  $\det Q = \alpha\gamma - \beta^2 \leq 0$ .

Furthermore,  $Q_0(x)$  is pseudoconvex on  $\mathbb{R}_+^2 \setminus \{(0, 0)\}$  if and only if in addition to i) and ii) we have  $(\alpha, \beta) \neq (0, 0)$  and  $(\beta, \gamma) \neq (0, 0)$ .

**Example 5.1** Consider the matrices

$$A = \begin{bmatrix} \alpha & 0 \\ 0 & 0 \end{bmatrix}, \alpha < 0; \quad B = \begin{bmatrix} 0 & \beta \\ \beta & 0 \end{bmatrix}, \beta < 0;$$

$$C = \begin{bmatrix} 0 & 0 \\ 0 & \gamma \end{bmatrix}, \gamma < 0; \quad D = \begin{bmatrix} \alpha & \beta \\ \beta & 0 \end{bmatrix}, \alpha < 0, \beta < 0.$$

The quadratic forms associated with all the matrices are quasiconvex on  $\mathbb{R}_+^2$  but only the quadratic forms associated with the matrices  $B$  and  $D$  are pseudoconvex on  $\mathbb{R}_+^2 \setminus \{(0, 0)\}$ .

## 6 A special case

The necessary and sufficient conditions stated in the previous sections are, in general, not easy to use for testing the quasiconvexity (pseudoconvexity) of a quadratic function. Nevertheless, when  $Q(x)$  has a particular structure, it is possible to obtain a characterization that is easy to test. In this section we shall consider the following class of functions:

$$f(x) = (a^T x + a_0)(b^T x + b_0) + c^T x. \quad (6.11)$$

**Theorem 6.1** Consider the function  $f$  in (6.11) and assume that the vectors  $a$  and  $b$  are linearly independent. Then,  $f$  is merely quasiconvex on a convex set  $S \subset \mathbb{R}^n$  with

nonempty interior if and only if

- i) there exist  $\alpha, \beta \in \mathbb{R}$  such that  $c = \alpha a + \beta b$ ;
- ii)  $S \subseteq \{x \in \mathbb{R}^n : a^T x + a_0 + \beta \geq 0, b^T x + b_0 + \alpha \leq 0\}$  or  $S \subseteq \{x \in \mathbb{R}^n : a^T x + a_0 + \beta \leq 0, b^T x + b_0 + \alpha \geq 0\}$ .

*Proof* We have  $f(x) = \frac{1}{2}x^T Qx + q^T x + q_0$ , where

$$Q = ab^T + ba^T, \quad q = b_0 a + a_0 b + c, \quad q_0 = a_0 b_0.$$

The linear independence of  $a$  and  $b$  implies that  $\dim(\text{Im}Q) = 2$ , where  $\text{Im}Q = \{z = \mu_1 a + \mu_2 b, \mu_1, \mu_2 \in \mathbb{R}\}$ ; consequently,  $\dim(\text{ker}Q) = n - 2$ . Taking into account that the quadratic form  $x^T Qx = 2a^T x b^T x$  is not constant in sign, we necessarily have a unique negative eigenvalue, i.e.,  $\nu_-(Q) = 1$ . From Theorem 3.7,  $f$  is quasiconvex on  $S$  if and only if there exists  $s \in \mathbb{R}^n$  such that  $Qs + q = 0$  and  $S \subseteq s + T$  or  $S \subseteq s - T$ . We have  $Qs + q = 0$  if and only if  $q \in \text{Im}Q$  or, equivalently, if and only if there exist  $\alpha, \beta \in \mathbb{R}$  such that  $c = \alpha a + \beta b$ , i.e., if and only if i) holds. Furthermore,  $Qs + q = (b^T s + b_0 + \alpha)a + (a^T s + a_0 + \beta)b$  so that  $Qs + q = 0$  if and only if  $b^T s = -(b_0 + \alpha)$  and  $a^T s = -(a_0 + \beta)$ . By means of simple calculations we have  $(x - s)^T Q(x - s) = 2(a^T x + a_0 + \beta)(b^T x + b_0 + \alpha)$  and thus  $(x - s)^T Q(x - s) \leq 0$  if and only if ii) holds.

The proof is complete. □

**Remark 6.1** *When  $a$  and  $b$  are linearly dependent,  $f$  is convex on  $\mathbb{R}^n$  or it is concave on  $\mathbb{R}^n$ . In this last case  $f$  turns out to be quasiconvex on a convex set  $S$  if and only if  $c = \alpha a$  and  $S$  is contained in one of the two half-spaces associated with the hyperplane given by the set of critical points of the function.*

**Corollary 6.1** *Consider the function  $f$  in (6.11) and assume that  $a$  and  $b$  are linearly independent. Then,  $f$  is merely pseudoconvex on a convex set  $S \subset \mathbb{R}^n$  with nonempty interior if and only if*

- i) there exist  $\alpha, \beta \in \mathbb{R}$  such that  $c = \alpha a + \beta b$ ;
- ii)  $S \subseteq \{x \in \mathbb{R}^n : a^T x + a_0 + \beta > 0, b^T x + b_0 + \alpha \leq 0\} \cup \{x \in \mathbb{R}^n : a^T x + a_0 + \beta \geq 0, b^T x + b_0 + \alpha < 0\}$ .

$0, b^T x + b_0 + \alpha < 0\}$  or

$S \subseteq \{x \in \mathbb{R}^n : a^T x + a_0 + \beta < 0, b^T x + b_0 + \alpha \geq 0\} \cup \{x \in \mathbb{R}^n : a^T x + a_0 + \beta \leq 0, b^T x + b_0 + \alpha > 0\}$ .

*Proof* The thesis follows from Lemma 5.2, taking into account that

$\nabla f(x_0) = 0$  if and only if  $x_0 \in \{x \in \mathbb{R}^n : a^T x + a_0 + \beta = 0, b^T x + b_0 + \alpha = 0\}$ .  $\square$

In order to characterize the quasiconvexity of  $f$  on  $\mathbb{R}_+^n$ , we shall state, firstly, the following lemma.

**Lemma 6.1** Consider the matrix  $Q = ab^T + ba^T$ . Then,  $Q \leq 0$  if and only if  $a \geq 0, b \leq 0$  or  $a \leq 0, b \geq 0$ .

*Proof* Obviously, if  $a \geq 0, b \leq 0$  or  $a \leq 0, b \geq 0$ , then  $Q \leq 0$ . Conversely, since the thesis is trivial if  $a = 0$  or  $b = 0$ , we shall consider the case  $a \neq 0, b \neq 0$ . Assume by contradiction, the existence of  $i, j$  such that  $a_i > 0, a_j < 0$  and consider the submatrix

$Q_{ij} = \begin{bmatrix} 2a_i b_i & a_i b_j + a_j b_i \\ a_i b_j + a_j b_i & 2a_j b_j \end{bmatrix}$ . If  $b_i b_j \neq 0$ ,  $a_i b_i \leq 0, a_j b_j \leq 0$  imply that  $b_i < 0, b_j > 0$ , respectively, so that  $a_i b_j + a_j b_i > 0$  and this is absurd. If  $b_i = 0$  and  $b_j \neq 0$ , we have

$Q_{ij} = \begin{bmatrix} 0 & a_i b_j \\ a_i b_j & 2a_j b_j \end{bmatrix}$ , so that  $a_j b_j \leq 0$  implies that  $b_j > 0$  while  $a_i b_j \leq 0$  implies that  $b_j < 0$  and, once again, we get a contradiction. The case  $b_j = 0, b_i \neq 0$  is analogous, so that the case  $b_i = 0, b_j = 0$  remains to be considered. Let  $k$  be such that  $b_k \neq 0$  and consider the submatrix

$\begin{bmatrix} a_i b_k + a_k b_i & a_i b_j + a_j b_i \\ 2a_k b_k & a_k b_j + a_j b_k \end{bmatrix} = \begin{bmatrix} a_i b_k & 0 \\ 2a_k b_k & a_j b_k \end{bmatrix}$ ;  $a_i b_k < 0$  implies that  $b_k < 0$  while  $a_j b_k < 0$  implies that  $b_k > 0$  and this is absurd. Consequently, we have

$a \geq 0$  or  $a \leq 0$ . For symmetric reasons, the components of  $b$  also have the same sign so that, necessarily,  $a \geq 0, b \leq 0$  or  $a \leq 0, b \geq 0$ . The thesis is achieved.  $\square$

**Theorem 6.2** Consider the function  $f$  in (6.11) and assume that  $a$  and  $b$  are linearly independent. Then,  $f$  is merely quasiconvex on  $\mathbb{R}_+^n$  if and only if there exist  $\alpha, \beta \in \mathbb{R}$  such that  $c = \alpha a + \beta b$  and one of the following conditions holds:

i)  $a \geq 0, b \leq 0, \alpha \leq -b_0, \beta \geq -a_0$ ;

ii)  $a \leq 0, b \geq 0, \alpha \geq -b_0, \beta \leq -a_0$ .

*Proof* From Theorem 4.4 we have  $Q = ab^T + ba^T \leq 0$ , while from Lemma 6.1 we have  $a \geq 0, b \leq 0$  or  $a \leq 0, b \geq 0$ . The thesis follows from Theorem 6.1.  $\square$

**Corollary 6.2** Consider the function  $f$  in (6.11) and assume that  $a$  and  $b$  are linearly independent. Then,  $f$  is merely pseudoconvex on  $\mathbb{R}_+^n$  if and only if there exist  $\alpha, \beta \in \mathbb{R}$  such that  $c = \alpha a + \beta b$  and one of the following conditions holds:

i)  $a \geq 0, b \leq 0$  and  $\alpha < -b_0, \beta \geq -a_0$  or  $\alpha \leq -b_0, \beta > -a_0$ ;

ii)  $a \leq 0, b \geq 0$  and  $\alpha > -b_0, \beta \leq -a_0$  or  $\alpha \geq -b_0, \beta < -a_0$ .

*Proof* Referring to Theorem 5.7, it is sufficient to note that  $b_0 a + a_0 b + c \neq 0$  if and only if  $a_0 + \beta \neq 0$  or  $b_0 + \alpha \neq 0$ .  $\square$

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