

## ON DIFFERENTIABILITY OF METRIC PROJECTIONS IN $\mathbf{R}^n$ , 1: BOUNDARY CASE

ALEXANDER SHAPIRO

**ABSTRACT.** This paper is concerned with metric projections onto a closed subset  $S$  of a finite-dimensional normed space. Necessary and in a sense sufficient conditions for directional differentiability of a metric projection at a boundary point of  $S$  are given in terms of approximating cones. It is shown that if  $S$  is defined by a number of inequality constraints and a constraint qualification holds, then the approximating cone exists.

**1. Introduction.** Let  $\|\cdot\|$  be a norm defined on the  $n$ -dimensional real vector space  $\mathbf{R}^n$  and  $S$  be a closed subset of  $\mathbf{R}^n$ . The set-valued metric projection onto  $S$  is the point-to-set mapping  $\Omega: \mathbf{R}^n \rightrightarrows S$  which corresponds to an  $x \in \mathbf{R}^n$  the set of elements of  $S$  closest to  $x$ , that is

$$\Omega x = \{y \in S: \|x - y\| = d(x, S)\},$$

where  $d(x, S)$  denotes the distance from  $x$  to  $S$ ,

$$d(x, S) = \inf\{\|x - y\|: y \in S\}.$$

For every  $x$  the set  $\Omega x$  is nonempty and compact, although possibly is not a singleton. We consider a selection mapping  $P_S: \mathbf{R}^n \rightarrow S$  associated with  $\Omega$ , which is defined as assigning to  $x$  a closest point  $y = P_S x$  in  $S$ , so that  $y \in \Omega x$ . One may note that if  $x \in S$ , then  $P_S x = x$  and  $P_S$  is continuous at  $x$ .

It is well known that if  $S$  is convex and the norm is strictly convex, then  $\Omega x$  is a singleton and hence  $P_S x$  is uniquely defined for all  $x$ . In this case it was shown in a number of publications that  $P_S$  is directionally differentiable at every  $x \in S$ , e.g., [5, 7, 8, 12]. In this article we investigate  $P_S$  at a boundary point of  $S$  for an arbitrary closed set  $S$ . We give necessary and in a sense sufficient conditions for directional differentiability of  $P_S$  in terms of approximating cones. It will be shown that if  $S$  is defined by a number of inequality constraints, then under mild assumptions the approximating cone exists and thus  $P_S$  is directionally differentiable.

**2. Cone approximation and directional differentiability of metric projections.** The following concept of approximating cones has proved to be useful, e.g., in sensitivity analysis of nonlinear programs (see Shapiro [11]) and in deriving some asymptotic results in inequality constrained estimation [2, 10].

---

Received by the editors November 13, 1985.

1980 *Mathematics Subject Classification* (1985 *Revision*). Primary 41A29, 41A50.

*Key words and phrases.* Metric projection, normed space, distance function, directional differentiability, approximating cone.

DEFINITION 1. We say the set  $S$  is approximated at  $x_0 \in S$  by a closed cone  $\mathcal{C}$ , called an approximating cone, if

$$(2.1) \quad \lim_{\substack{x \rightarrow x_0 \\ x \in S}} \frac{d(x - x_0, \mathcal{C})}{\|x - x_0\|} = 0$$

and

$$(2.2) \quad \lim_{\substack{y \rightarrow 0 \\ y \in \mathcal{C}}} \frac{d(x_0 + y, S)}{\|y\|} = 0.$$

We recall that a function (or mapping)  $f(x)$  is said to be directionally differentiable at a point  $x$  if the limit (directional derivative)

$$f'(x, y) = \lim_{t \rightarrow 0^+} \frac{f(x + ty) - f(x)}{t}$$

exists for all  $y \in \mathbf{R}^n$ . In the following theorem we give a characterization of approximating cones in terms of directional differentiability of the distance function.

THEOREM 1. The approximating cone  $\mathcal{C}$ , to  $S$  at  $x_0 \in S$ , exists iff the distance function  $\delta(\cdot) = d(\cdot, S)$  is directionally differentiable at  $x_0$ , in which case

$$(2.3) \quad \delta'(x_0, \cdot) = d(\cdot, \mathcal{C}).$$

PROOF. First we observe that

$$(2.4) \quad \|P_S x - x_0\| \leq \|P_S x - x\| + \|x - x_0\| \leq 2\|x - x_0\|$$

and hence  $P_S$  is continuous at  $x_0$ . Now suppose that the approximating cone  $\mathcal{C}$  exists. Then it follows from (2.1) that there is  $y = y(x) \in \mathcal{C}$  such that

$$\|P_S x - x_0 - y\| = o(\|P_S x - x_0\|),$$

which by (2.4) becomes

$$(2.5) \quad \|P_S x - x_0 - y\| = o(\|x - x_0\|).$$

We have that

$$(2.6) \quad d(x, S) = \|x - P_S x\| \geq \|x - x_0 - y\| - \|P_S x - x_0 - y\|.$$

Noting that  $d(x - x_0, \mathcal{C}) \leq \|x - x_0 - y\|$  and applying (2.5) we obtain from (2.6) that

$$d(x, S) \geq d(x - x_0, \mathcal{C}) + o(\|x - x_0\|).$$

In a similar way it follows from (2.2) that

$$d(x - x_0, \mathcal{C}) \geq d(x, S) + o(\|x - x_0\|)$$

and hence

$$(2.7) \quad d(x, S) = d(x - x_0, \mathcal{C}) + o(\|x - x_0\|).$$

Since  $\mathcal{C}$  is a cone, the function  $d(\cdot, \mathcal{C})$  is positively homogeneous. Together with (2.7) this implied that for every  $y$

$$d(x_0 + ty, S) = td(y, \mathcal{C}) + o(t),$$

i.e.  $\delta(\cdot)$  is directionally differentiable at  $x_0$  and (2.3) follows.

Conversely, suppose that  $\delta(\cdot)$  is directionally differentiable at  $x_0$  and consider the cone

$$(2.8) \quad \mathcal{C} = \{y : \delta'(x_0, y) = 0\}.$$

Note that the distance function  $\delta(\cdot)$  and then the directional derivative  $\delta'(x_0, \cdot)$  are Lipschitz and hence are continuous functions. Consequently,  $\mathcal{C}$  is a closed cone. Now we show that  $S$  is approximated at  $x_0$  by  $\mathcal{C}$ . Since  $\delta(\cdot)$  is Lipschitz, the directional derivative  $\delta'(x_0, y)$  gives the first-order approximation of  $\delta(\cdot)$  uniformly in  $y$ , i.e.

$$(2.9) \quad \delta(x_0 + y) = \delta'(x_0, y) + o(\|y\|)$$

(e.g., Demyanov and Rubinov [4, Lemma 3.2]). Then (2.2) follows immediately from (2.8) and (2.9). Also, because of (2.9) and since  $\delta(x_0 + y) = 0$  for  $x_0 + y \in S$ , we have that

$$\delta'(x_0, y)/\|y\| = \delta'(x_0, y/\|y\|) \rightarrow 0$$

as  $y \rightarrow 0$  and  $x_0 + y \in S$ . Since the function  $d(\cdot, \mathcal{C})$  is continuous, this implies by the standard argument of compactness that  $d(y/\|y\|, \mathcal{C}) \rightarrow 0$  as  $y \rightarrow 0$  and  $x_0 + y \in S$ . Consequently,  $d(y, \mathcal{C}) = o(\|y\|)$  for  $x_0 + y \in S$  and (2.1) follows.  $\square$

It follows from Theorem 1 that if the approximating cone exists, then it can be represented in the form (2.8) and consequently is unique. Moreover, since all norms in  $\mathbf{R}^n$  are equivalent, the definition of approximating cones is independent of a particular choice of the norm  $\|\cdot\|$ . Therefore, Theorem 1 implies the following result:

**COROLLARY 1.** *If for a given set  $S$  and a certain norm the distance function is directionally differentiable at  $x \in S$ , then it is directionally differentiable at  $x$  for any other norm.*

If the set  $S$  is convex, then the distance function  $d(\cdot, S)$  is also convex and hence is directionally differentiable (e.g. [9]). Consequently, in this case the approximating cone exists at every  $x \in S$  and coincides with the tangent cone  $\text{cl} \bigcup \{\lambda(S - x) : \lambda > 0\}$ .

Now we formulate the main result of this paper. With the cone  $\mathcal{C}$  and norm  $\|\cdot\|$  is associated the set-valued metric projection  $\Pi: \mathbf{R}^n \rightrightarrows \mathcal{C}$ ,

$$\Pi y = \{z \in \mathcal{C} : \|y - z\| = d(y, \mathcal{C})\}.$$

One can see that for every  $y$  the set  $\Pi y$  is nonempty and compact and that the mapping  $\Pi$  is positively homogeneous, i.e.  $\Pi(ty) = t\Pi y$  for all  $t > 0$ .

**THEOREM 2.** *Let  $S$  be approximated at  $x_0 \in S$  by a cone  $\mathcal{C}$ . Then for every  $y \neq 0$ ,*

$$(2.10) \quad \lim_{t \rightarrow 0^+} \frac{d(P_S(x_0 + ty), x_0 + t\Pi y)}{t} = 0.$$

**PROOF.** Suppose that (2.10) does not hold. Then there exists a vector  $\bar{y} \neq 0$  and a sequence  $t_n \rightarrow 0^+$  of positive numbers such that

$$(2.11) \quad d(z_n, t_n \Pi \bar{y}) \geq \varepsilon t_n,$$

where  $z_n = P_S(x_0 + t_n \bar{y}) - x_0$  and  $\varepsilon$  is a positive constant. As in (2.4) we have that  $\|z_n\| \leq 2\|t_n \bar{y}\|$  and hence by the argument of compactness the sequence  $\{t_n\}$

can be chosen in such a way that  $t_n^{-1}z_n$  converges to a vector  $\bar{z}$ , say. Note that it is implied by the condition (2.1) of the definition of approximating cones that  $\bar{z} \in \mathcal{C}$ . It follows from (2.11) that  $d(t_n^{-1}z_n, \Pi\bar{y}) \geq \varepsilon$ , and hence by the continuity of  $d(\cdot, \Pi\bar{y})$  that  $d(\bar{z}, \Pi\bar{y})$  is greater than zero. Consequently  $\bar{z}$  does not belong to the set  $\Pi\bar{y}$ .

On the other hand,

$$\begin{aligned} \|\bar{y} - \bar{z}\| &= t_n^{-1}\|t_n\bar{y} - t_n\bar{z}\| = t_n^{-1}(\|t_n\bar{y} - z_n\| + o(t_n)) \\ &= t_n^{-1}[d(x_0 + t_n\bar{y}, S) + o(t_n)]. \end{aligned}$$

Moreover, by Theorem 1 we have that  $t_n^{-1}d(x_0 + t_n\bar{y}, S) \rightarrow d(\bar{y}, \mathcal{C})$  and hence  $\|\bar{y} - \bar{z}\| = d(\bar{y}, \mathcal{C})$ . This implies that  $\bar{z} \in \Pi\bar{y}$ , a contradiction.  $\square$

It follows from Theorem 2 that if the approximating cone  $\mathcal{C}$  exists and  $\Pi\bar{y} = \{\bar{z}\}$  is a singleton for some  $\bar{y} \in \mathbf{R}^n$ , then

$$P_S(x_0 + t\bar{y}) = x_0 + t\bar{z} + o(t), \quad t > 0.$$

Thus  $\bar{z}$  is the directional derivative of  $P_S$  in the direction  $\bar{y}$ . In this sense existence of the approximating cone is a sufficient condition for directional differentiability of  $P_S$ . On the other hand, it is clear that if  $P_S$  is directionally differentiable at  $x_0 \in S$ , then the distance function  $d(\cdot, S)$  also is, and hence the approximating cone exists. Consequently, existence of the approximating cone is always a necessary condition for directional differentiability of  $P_S$ .

**3. Cone approximation of a set defined by inequality constraints.** The concept of approximating cones closely resembles the standard notion of Bouligand's contingent cone  $T_S(x_0)$ , which can be defined as follows:

$$T_S(x_0) = \left\{ y : \liminf_{t \rightarrow 0^+} \frac{d(x_0 + ty, S)}{t} = 0 \right\}.$$

(For an equivalent definition and elementary properties of  $T_S(x_0)$  see, e.g., [1, pp. 176–179].) Theorem 1 implies that whenever the approximating cone exists it coincides with the contingent cone  $T_S(x_0)$ . In the following example we demonstrate the difference between contingent and approximating cones.

**EXAMPLE.** Consider the set  $S = \{2^{-k} : k = 1, 2, \dots\} \cup \{0\}$  in  $\mathbf{R}$ . Then  $T_S(0) = \mathbf{R}_+$ , whereas the distance function  $d(\cdot, S)$  is not differentiable at zero in the positive direction, and hence the approximating cone does not exist at  $x_0 = 0$ .

In the remainder of this section we study the case where  $S$  is defined by an inequality constraint as follows:

$$S = \{x : g(x) \leq 0\}.$$

We suppose that the constraint function  $g(x)$  is uniformly directionally differentiable at  $x_0 \in S$ , i.e.  $g(x_0 + y) - g(x_0) = g'(x_0, y) + o(\|y\|)$ , and the directional derivative  $g'(x_0, y)$  is continuous in  $y$ . This holds, for example, if  $g(x)$  is directionally differentiable and locally Lipschitz. Let  $g(x_0) = 0$  and consider the cone

$$(3.1) \quad \mathcal{C} = \{y : g'(x_0, y) \leq 0\}.$$

We say that the *nondegeneracy condition* holds at  $x_0$  if

$$\mathcal{C} = \text{cl}\{y : g'(x_0, y) < 0\}.$$

We remark here that if  $g(x)$  is Fréchet differentiable at  $x_0$  (with nonzero gradient) and hence  $g'(x_0, y)$  is linear in  $y$ , then the cone  $\mathcal{C}$  is a half-space and the nondegeneracy condition holds automatically.

**THEOREM 3.** *Suppose that the nondegeneracy condition holds at the point  $x_0$ . Then  $S$  is approximated at  $x_0$  by the cone  $\mathcal{C}$ .*

**PROOF.** We have to show that conditions (2.1) and (2.2) are satisfied. Suppose that condition (2.1) does not hold. Then there exists a sequence  $\{x_n\}$  in  $S$  converging to  $x_0$  such that  $y_n/\|y_n\|$ , with  $y_n = x_n - x_0$ , tends to a vector  $\bar{y} \notin \mathcal{C}$ . But

$$0 \geq g(x_0 + y_n)/\|y_n\| = g'(x_0, y_n/\|y_n\|) + o(\|y_n\|)/\|y_n\|.$$

Consequently, we obtain from the continuity of  $g'(x_0, \cdot)$  that  $g'(x_0, \bar{y}) \leq 0$ , a contradiction.

Now suppose that (2.2) does not hold. Then there exists a sequence  $\{y_n\}$  in  $\mathcal{C}$  converging to zero such that  $d(x_0 + y_n, S) \geq 2\varepsilon\|y_n\|$  for some  $\varepsilon > 0$  and  $y_n/\|y_n\|$  tends to a vector  $\bar{y}$ . Since  $\mathcal{C}$  is closed,  $\bar{y} \in \mathcal{C}$ . Moreover, because of the nondegeneracy condition there exists a vector  $y^*$  such that  $g'(x_0, y^*) < 0$ ,  $\|y^*\| = \|\bar{y}\|$  and  $\|y^* - \bar{y}\| < \varepsilon$ . Consider the sequence  $y_n^* = y_n + \|y_n\|(y^* - \bar{y})$ . Then

$$d(x_0 + y_n^*, S) \geq d(x_0 + y_n, S) - \|y_n\| \|y^* - \bar{y}\| \geq \varepsilon\|y_n\|$$

and  $\|y_n^*/\|y_n^*\| \rightarrow 1$ . Consequently,  $x_0 + y_n^*$  does not belong to  $S$  for all  $n$  and  $y_n^*/\|y_n^*\|$  tends to  $y^*$ . From the continuity of  $g'(x_0, \cdot)$  we have that  $g'(x_0, y_n^*/\|y_n^*\|) < -\delta$  for some  $\delta > 0$  and hence  $g'(x_0, y_n^*) < -\delta\|y_n^*\|$ . Moreover, since  $g(x_0) \leq 0$  and because of the uniform directional differentiability of  $g(x)$ ,

$$g(x_0 + y_n^*) \leq g'(x_0, y_n^*) + o(\|y_n^*\|) \leq -\delta\|y_n^*\| + o(\|y_n^*\|).$$

Consequently,  $g(x_0 + y_n^*) \leq 0$  and hence  $x_0 + y_n^* \in S$  for sufficiently large  $n$ , a contradiction.  $\square$

Now let the set  $S$  be defined by a (possibly infinite) number of inequality constraints as follows:

$$S = \{x: g_\alpha(x) \leq 0, \alpha \in \Lambda\}.$$

Then we can replace these inequality constraints by one constraint  $g(x) \leq 0$  with  $g$  being the max-function.

$$g(x) = \sup\{g_\alpha(x): \alpha \in \Lambda\}.$$

By the well-known theorem on Danskin [3] we have that if

- (a) The index set  $\Lambda$  is compact,
- (b)  $g_\alpha(\cdot)$  is continuously differentiable for all  $\alpha \in \Lambda$ ,
- (c)  $g(x, \alpha) = g_\alpha(x)$  together with  $\nabla_x g(x, \alpha)$  are continuous on  $\mathbf{R}^n \times \Lambda$ .

Then  $g(x)$  is uniformly directionally differentiable and

$$(3.2) \quad g'(x_0, y) = \max\{y^T \nabla g_\alpha(x_0): \alpha \in \Lambda^*(x_0)\},$$

where  $\Lambda^*(x) = \{\alpha \in \Lambda: g(x) = g_\alpha(x)\}$ .

In this case the nondegeneracy condition means that

(d) There exists a vector  $b$  such that  $b^T \nabla g_\alpha(x_0) < 0$  for all  $\alpha \in \Lambda^*(x_0)$  (i.e. the Mangasarian-Fromovitz constraint qualification [6]).

Here the directional derivative  $g'(x_0, \cdot)$  is representable as the pointwise maximum (3.2) of a family of linear functions and hence is convex. Consequently, the cone  $\mathcal{C}$  given in (3.1) is convex. Therefore we obtain that if conditions (a)–(d) are satisfied and the norm  $\|\cdot\|$  is strictly convex, then  $P_S$  is directionally differentiable at the point  $x_0$ .

## REFERENCES

1. J. P. Aubin and A. Cellina, *Differential inclusions*, Grundlehren Math. Wiss., Band 264, Springer-Verlag, Berlin and New York, 1984.
2. H. Chernoff, *On the distribution of the likelihood ratio*, Ann. Math. Statist. **25** (1954), 573–578.
3. J. M. Danskin, *The theory of max-min and its applications to weapons allocation problems*, Econometrics and Operations Research, vol. 5, Springer-Verlag, Berlin and New York, 1967.
4. V. F. Demyanov and A. M. Rubinov, *On quasidifferentiable mappings*, Math. Operationsforsch. Statist. Ser. Optimization **14** (1983), 3–21.
5. A. Haraux, *How to differentiate the projection on a convex set in Hilbert space. Some applications to variational inequalities*, J. Math. Soc. Japan **29** (1977), 615–631.
6. O. L. Mangasarian and S. Fromovitz, *The Fritz John necessary optimality conditions in the presence of equality and inequality constraints*, J. Math. Anal. Appl. **17** (1967), 37–47.
7. G. P. McCormick and R. Tapia, *The gradient projection method under mild differentiability conditions*, SIAM J. Control **10** (1972), 93–98.
8. R. R. Phelps, *Metric projections and the gradient projection method in Banach spaces*, SIAM J. Control and Optim. **23** (1985), 973–977.
9. R. T. Rockafellar, *Convex analysis*, Princeton Univ. Press, Princeton, N.J., 1970.
10. A. Shapiro, *Asymptotic distribution of test statistics in the analysis of moment structures under inequality constraints*, Biometrika **72** (1985), 133–144.
11. ———, *Second order sensitivity analysis and asymptotic theory of parametrized nonlinear programs*, Mathematical Programming **33** (1985), 280–299.
12. E. H. Zarantonello, *Projections on convex sets in Hilbert space and spectral theory*, Contributions to Nonlinear Functional Analysis, Academic Press, New York, 1971, pp. 237–424.

DEPARTMENT OF MATHEMATICS AND APPLIED MATHEMATICS, UNIVERSITY OF SOUTH AFRICA, P. O. BOX 392, PRETORIA 0001, SOUTH AFRICA